# Platform Competition under Dispersed Information\*

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#### Abstract

We study monopolistic and competitive pricing in a two-sided market where agents have incomplete information about the quality of the product provided by each platform. The analysis is carried out within a global-game framework that offers the convenience of equilibrium uniqueness while permitting the outcome of such equilibrium to depend on the pricing strategies of the competing platforms. We first show how the dispersion of information interacts with the network effects in determining the elasticity of demand on each side and thereby the equilibrium prices. We then study "informative" advertising campaigns that increase the agents' ability to estimate their own valuations and/or the distribution of valuations on the other side of the market.

JEL classification: D82

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### 1 Introduction

Many markets feature platforms mediating the interactions among the various sides of the market. Examples include media outlets mediating the interactions between readers/viewers on one side and content providers and advertisers on the other side, video-game consoles mediating the interactions between gamers and video-game developers, operating systems mediating the interactions between end-users and software developers, employment agencies mediating the interactions between employers and job seekers, and dating agencies mediating the search of partner-seekers.

Following the initial work of Caillaud and Jullien (2001,2003), Rochet and Tirole (2003), and Armstrong (2006), the two-sided market literature has studied the role of prices in implementing such mediated interactions (See Rysman (2009) and Weyl (2010) for excellent overviews and for recent developments).

The assumption that is commonly made in this literature is that preferences on each side of the market are common knowledge. This assumption implies that, given the prices set by the platforms, each agent can perfectly predict the participation decisions of any other agent. In equilibrium, such predictions are accurate and coincide with the platforms' predictions.

While a convenient modelling shortcut, the assumption that preferences are common knowledge does not square well with most markets. Preferences over the products and services of different platforms typically reflect personal traits, making it difficult for an agent to predict the behavior of other agents. Due to network externalities, predicting how many agents from the opposite side will choose a given platform is key to an agent's own decision about which platform to join. Furthermore, because preferences are typically positively correlated among agents from the same side (albeit, possibly negatively correlated with the preferences of those agents from the opposite side), agents may experience difficulty in predicting not just individual actions but also the entire distribution of actions in the cross-section of the population. In other words, agents from each side face nontrivial uncertainty about how many agents from the opposite side will choose one platform over the other.

Because such uncertainty impacts the elasticity of the demand that the platform faces on each side, it is bound to impact the equilibrium prices and thereby the allocations they induce. In addition, the platforms themselves may face uncertainty about the distribution of preferences in the cross-section of the population and hence about the demand they face on each side, which also contributes to their pricing strategies.

In this paper, we develop a tractable, yet rich, model of platform competition under dispersed information, where the distribution of preferences in the cross-section of the population is unknown to both the platforms and to each individual agent, and where each agent has private information both about his own preferences as well as about the distribution of preferences in the cross-section of the population. Part of the contribution is in showing how such dispersion of information interacts with the network effects that are typical of multi-sided markets in determining the elasticity of

demand on each side. We then use such a characterization to examine the effects of the dispersion of information on the equilibrium prices and on the allocations they induce. Finally, we examine the platforms' incentives to change the information available to each side of the market through informative advertising campaigns, as well as their incentive to innovate by changing the way their product is likely to be perceived relative to those of the competitors.<sup>1</sup>

Model preview. Two platforms compete on two sides of a market populated by a continuum of agents on each side. Each agent derives a direct utility from each platform's product or services, hereafter referred to as the agent's "stand-alone valuation" (other terms favored in the literature include "intrinsic benefit" or "membership benefit"). In addition, each agent derives an indirect utility from interacting with the other side that is proportional to the number of agents from the other side who join the same platform; hereafter, we will refer to this component of the agent's payoff as "network effect" (other expressions favored in the literature include "usage valuation", "cross-side externality" and "interaction benefit"). Each agent is uncertain about the distribution of stand-alone valuations in the cross section of the population. In addition, we allow for the possibility that each agent faces uncertainty about his own stand-alone valuations for the two platforms, reflecting the idea that agents need not know which products and services serve best their needs (think of an agent choosing over competing technologies).

For simplicity, we assume that all agents from the same side attach the same value to interacting with the opposite side. However, because agents differ in their expectation about how many agents from the opposite side will join, de facto, agents are heterogeneous not only in their true (and estimated) stand-alone valuations, but also in their estimation of the network effects from joining each of the two platforms.<sup>2</sup>

We allow for the possibility that the network effects be negative on one side but assume that there is always one side where they are positive (for example, in the case of a media outlet competing for readers, or viewers, on one side and for advertisers on the other side, it is reasonable to assume that network effects are negative on the readers' side—most readers dislike advertisement—but positive on the advertisers' side). We also assume that stand-alone valuations are positively correlated between any two agents from the same side but possibly negatively correlated between two agents from opposite sides (Think of the market for operating systems; a system that appeals to software developers need not necessarily appeal to end-users, for the latter typically value the various features of the operating system differently from the developers—e.g., they may value the simplicity of the key tasks more than the flexibility and sophistication of the code).

We build on the global-game literature (Carlsson and Van Damme (1993), Morris and Shin (2003)) by assuming that the cross-sectional distribution of the stand-alone valuations can be parametrized by a bivariate state-variable drawn from a known distribution which constitutes the

<sup>&</sup>lt;sup>1</sup>See Anderson and Renault (2006, 2009) for recent models of advertising along this line.

<sup>&</sup>lt;sup>2</sup>We do not expect any significant change to the nature of the results coming from the introduction of heterogeneity in the importance the agents assign to the network effects.

common prior. Each agent then receives a noisy signal of his own stand-alone valuations for the products of the two platforms which he uses to decide which platform to join. Because of network effects, agents use their signal not only to estimate their own stand-alone valuations, but also to predict the distribution of stand-alone valuations on the other side of the market. In other words, agents use their appreciation of each platform's product and services to form an opinion about the preferences on the other side of the market. This inference problem creates new subtle effects that are missing under complete information and that are reflected in the determination of the equilibrium allocations (prices and participation decisions).

Implications for equilibrium prices. As in most of the literature, we abstract from price discrimination and assume that platforms compete by setting access fees to each side of the market. By paying the fee, an agent is granted access to the platform's product and thereby also obtains access to the other side of the market. To isolate the effects mentioned above, we assume that platforms do not possess any private information relative to the rest of the market. This permits us to abstract from the signaling role of prices and instead focus on how prices respond to the agents' extrapolation from their own preferences to the distribution of preferences in the cross-section of the population.<sup>3</sup>

The advantage of casting the analysis within a global-game framework is twofold: (i) it permits us to investigate the implications of the dispersion of information on equilibrium prices, and (ii) it guarantees that the equilibrium demand functions are unique (thus avoiding the usual "chicken and egg" problem of many models of competition in two-sided markets—e.g., Caillaud and Jullien (2003)): For any given vector of prices there is a unique continuation equilibrium in the subgame where agents choose which platform to join (note that this is true despite the fact that platforms in our model compete in simple access fees—as they do in many markets—that do not condition on participation rates from the opposite side).<sup>4</sup>

One difference relative to the complete-information case is that the beliefs of the "marginal agent" on each side about the participation decisions on the opposite side depend on the marginal agent's own estimated stand-alone valuation (the marginal agent is the one who is indifferent between joining one platform or the other). As the platform changes its price on one side, the marginal agent's beliefs also change (note that, under complete information, the marginal agent's expectation of the participation rate from the opposite side always coincides with the platform's expectation).

This observation has important implications for equilibrium prices. Suppose, for example, that network effects are positive on each side (meaning that all agents benefit from a higher participation

<sup>&</sup>lt;sup>3</sup>We also abstract from within-side externalities and heterogeneity in users' attractiveness. See Damiano and Li (2007), Gomes and Pavan (2011), and Vega and Weyl (2012) for models that accommodate a certain form of price discrimination and heterogeneity in attractiveness.

<sup>&</sup>lt;sup>4</sup>In the baseline version of the model we do not allow agents to multi-home (that is, to join both platforms). Later in the paper, however, we relaxed this assumption and show that multihoming does not obtain under reasonable parameter configurations if one assumes that platforms cannot set negative prices.

rate on the opposite side) and that tastes are positively correlated between the two sides (so that a high perceived stand-alone valuation is "good news" about participation from the opposite side). Suppose then that a platform were to raise its price on, say, side 1. Because the marginal agent who is excluded is the most "pessimistic" about side 2's participation, among those who are joining the platform, the drop in expected demand is smaller than in a world where all agents share the same beliefs about the other side's participation (as in complete-information models). In other words, when preferences are positively correlated between the two sides and network effects are positive on both sides, this new effect contributes to a reduction in the own-price elasticity of the demand functions. As a result of this new effect, the equilibrium price on each side increases with the intensity of the network effects on that side when preferences are positively correlated between the two sides, and decreases otherwise. This is in contrast to the complete-information case where the equilibrium price on each side i decreases with the intensity of the network effects on side i, but is independent of the intensity of the network effects on side i (see, e.g., Armstrong, 2006, and Rochet and Tirole, 2006).<sup>5,6</sup>

A second insight is that, holding fixed the ex-ante distribution of estimated stand-alone valuations, the equilibrium prices depend on the distribution of information over the two sides only through the coefficient of mutual forecastability, which is an increasing transformation of the correlation coefficient between the signals of any two agents from opposite sides. Indeed what matters for the impact of network effects on equilibrium prices is the ability of each side to predict the change of demand on the other side triggered by a variation in prices. Suppose that the quality of information is very high on one side but not on the other. Then, the less-informed side will not respond much to variations in the distribution of stand-alone valuations, making the information of the other side of limited value. Fixing the ex-ante distribution of estimated stand-alone valuations (which amounts to fixing the ex-ante degree of horizontal differentiation between the two platforms), what matters for equilibrium prices is not so much the ability of each side to forecast the distribution of true stand-alone valuations on the opposite side but its ability to predict how the distribution of estimated stand-alone valuations on the opposite side changes with the underlying "state". As a result, equilibrium prices respond to variations in the information structure only through the impact that these variations have on the two sides mutual ability to forecast each other, as captured by the coefficient of mutual forecastability. In the special case of a market that is

<sup>&</sup>lt;sup>5</sup>The pricing formulae obtained in these papers can be understood as monopoly pricing adjusted for the fact that a platform can leverage an increase of demand on one side by increasing the price it charges to the other side. This means that the relevant opportunity cost for losing the marginal agent on one side must incorporate the revenue loss stemming from the lower utility enjoyed by all agents who participate on the opposite side, thus explaining why prices depend negatively on the intensity of the other-side network effects.

<sup>&</sup>lt;sup>6</sup>Note that if agents differed in the importance they assign to network effects, then equilibrium prices would depend on the intensity of own-side network effects also under complete information. The effect of own-side network effects on equilibrium prices would then depend on the correlation between network effects and stand-alone valuations. Surprisingly, the effect of such correlation on equilibrium prices has received little attention in the literature.

perfectly symmetric under complete information (meaning that the intensity of the network effects is the same on the two sides and so is the ex-ante distribution of stand-alone valuations), the fact that prices depend on the information structure only through the coefficient of mutual forecastability implies that the equilibrium prices remain perfectly symmetric under dispersed information, despite possible asymmetries in the distribution of information over the two sides.

Implications for advertising campaigns and product selection. Having characterized how equilibrium prices depend on the prior distribution of stand-alone valuations and on the dispersion of information, we then proceed by investigating the platforms' incentives to engage in advertising campaigns that provide information (possibly in an asymmetric way) to each of the two sides. In our model, firms are uninformed about the true distribution of stand-alone valuations. This implies that advertising campaigns cannot change the mean of the distribution of estimated stand-alone valuations but only the agents' ability to predict their own stand-alone valuations as well as their ability to predict the distribution of (true and estimated) stand-alone valuations in the cross-section of the population.<sup>7</sup>

We show that campaigns that increase the agents' ability to estimate their own stand alone valuations (holding fixed the agents' ability to predict the behavior of the other side) always raise profits by increasing the sensitivity of individual demands to information which amounts to increasing the ex-ante degree of differentiation between the two platforms (equivalently, reducing the elasticity of the residual demand functions), thus softening competition.

On the other hand, campaigns that help the agents predict the behavior of the other side, without affecting the agents' ability to estimate their own stand-alone valuations, increase profits if and only if the correlation of tastes between the two sides is of the same sign as the sum of the intensity of the network effects. In particular, this means that such campaigns increase profits when network effects are positive on both sides and tastes are positively correlated (as is probably the case for most video-games consoles). On the contrary, they decrease profits when either tastes are negatively correlated and network effects are positive (as is possibly the case for some operating systems), or tastes are (weakly) positively correlated but one side suffers from the presence of the other side more than the other side benefits from its presence.

To understand this last result, assume that preferences are positively correlated between the two sides and consider a campaign that increases the ability of, say, side-1's agents to forecast side-2's preferences. An increase in such ability reduces the own-price elasticity of demand on side 1 by making the marginal agent's beliefs more sensitive to his private information (Recall that, as explained above, a higher sensitivity to private information implies a lower drop in demand in response to an increase in price due to the fact that the marginal agent is less optimistic about participation from the opposite side than any of the infra-marginal agents). Interestingly, when

<sup>&</sup>lt;sup>7</sup>The reason why advertising campaigns cannot change the mean of the distribution of estimated stand-alone valuations is the same that makes hidden actions ineffective in the signal-jamming literature (see e.g., Fudenberg and Tirole (1986)).

preferences are positively correlated, an increase in the precision of information on side 1 about side-2's preferences also reduces the own-price elasticity of the side-2 demand by making the behavior of side-1's agents more predictable in the eyes of side-2's agents. These effects unambiguously contribute to a higher equilibrium price on each side.

At the same time, more precise information on side 1 also implies a higher sensitivity of both demands to variations in prices on the opposite side, which contributes negatively to the equilibrium prices. While the net effect on the equilibrium price on each side then depends on the relative importance that the two sides attach to interacting with one another, the net effect on total profits is always unambiguously positive when the sum of the network effects is positive (more generally, of the same sign as the correlation of preferences between the two sides). This is because, holding constant the ex-ante distributions of estimated stand-alone valuations, the equilibrium price on each side depends on the dispersion of information only through the index of mutual forecastability, which measures the two sides' ability to forecast each other and which is increasing in the quality of information on each of the two sides. When the sum of the network effects is positive, then any possible loss of revenues on one side must necessarily be more than compensated by an increase in revenues on the opposite side, making the equilibrium total profits unambiguously increase with each side's ability to forecast the distribution of preferences on the other side.

We conclude by investigating how equilibrium profits change with variations in the prior distribution from which stand-alone valuations are drawn. These comparative statics, contrary to the ones pertaining the quality of information, are meant to shed light on a platform's incentives to differentiate its product and services from the competitor's, without knowing the exact distribution of preferences on either side of the market. For instance, we show that raising the similarity with the opponent's product (which amounts to a prior that concentrates more measure around zero) always reduces the equilibrium profits by intensifying competition. On the other hand, aligning the preferences of the two sides by favoring dimensions that are appealing to both sides increases profits for positive network effects but reduces them when the sum of the network effects is negative (that is, when one side suffers from the presence of the other side more than the other side benefits from its presence).

Outline. The rest of the paper is organized as follows. Section 2 presents the model and introduces some preliminary results concerning the ability of each side to forecast its own preferences and the cross-sectional distribution of preferences on the other side of the market. Section 3 then characterizes optimal prices for a monopolistic platform. Section 4 contains the main results for the duopoly case. Section 5 contains implications for product positioning and advertising campaigns. Section 6 contains a few concluding remarks. All proofs are in the Appendix.

## 2 Model

**Players.** Two platforms, indexed by k = A, B, compete on two sides, i = 1, 2. Each side is populated by a measure-one continuum of agents, indexed by  $l \in [0, 1]$ .

Actions and payoffs. Each agent  $l \in [0,1]$  from each side i = 1, 2 must choose which platform to join, if any.<sup>8</sup>. The payoff  $U_{il}^k$  that agent l from side i derives from joining platform k is given by

$$U_{il}^k = u_{il}^k + \gamma_i m_j^k - p_i^k$$

where  $u_{il}^k$  is the idiosyncratic stand-alone valuation<sup>9</sup> of joining platform k,  $m_j^k \in [0,1]$  is the mass of agents from side  $j \neq i$  that join platform k,  $\gamma_i \in \mathbb{R}$  is a parameter that controls for the intensity of the network effects<sup>10</sup> on side i and  $p_i^k$  is the access fee (equivalently, the uniform price) charged by platform k to side i.

We assume the network effects are positive on at least one of the two sides but allow them to be negative on the opposite side; that is, we assume that there is  $i \in \{1, 2\}$  such that  $\gamma_i > 0$ .

The payoff that each agent  $l \in [0,1]$  from each side i = 1,2 obtains from not joining any platform is assumed to be equal to zero.

Each platform's payoff  $\Pi^k$  is the total revenue from collecting the access fees from the two sides:<sup>11</sup>

$$\Pi^k = p_1^k m_1^k + p_2^k m_2^k.$$

All players are risk-neutral expected-utility maximizers.

Horizontal differentiation and information. We assume that the stand-alone valuations are given by

$$u_{il}^A = s_i - \frac{1}{2} z_i [\theta_i + \varepsilon_{il}]$$
  
 $u_{il}^B = s_i + \frac{1}{2} z_i [\theta_i + \varepsilon_{il}]$ 

 $i=1,2,\ k=A,B,\ l\in[0,1]$ , where  $s_i\in\mathbb{R}$  and  $z_i\in\mathbb{R}_+$  are scalars whose role is to control for the agents' payoff relative to their outside options and for the importance the agents assign to the quality of the products and services provided by the two platforms,  $\theta_i$  is the realization of a Normal random variable with mean zero and variance  $\alpha_i^{-1}$  ( $\alpha_i$  is thus the precision of the distribution) and  $\varepsilon_{il}$  is the realization of a Normal random variable with mean zero and variance  $(\beta_i^u)^{-1}$ . The above specification is chosen so that the difference in stand-alone valuations

$$v_{il} \equiv u_{il}^B - u_{il}^A = z_i [\theta_i + \varepsilon_{il}]$$

<sup>&</sup>lt;sup>8</sup>Below we will also discuss the possibility that the agents may choose to join both platforms (multihoming).

<sup>&</sup>lt;sup>9</sup>Also referred to in the literature as "intrinsic benefit" — see, e.g., Armstrong and Wright (2007) — and "membership benefit" — see e.g., Weyl (2010).

<sup>&</sup>lt;sup>10</sup> Also referred to in the literature as "usage value" (e.g., Rochet and Tirole (2006)), "cross-side externality" (e.g., Armstrong (2006)) and "interaction benefit" (e.g., Weyl, (2010)).

<sup>&</sup>lt;sup>11</sup>All results extend to the case where the platforms incur costs to provide access to the users. Because these costs do not play any role, we disregard them to facilitate the exposition.

is Normally distributed with mean zero and variance  $z_i^2(\alpha_i + \beta_i^u)/\alpha_i\beta_i^u$ .

The random variable  $\tilde{\theta}_i$  captures the common-value component in the stand-alone valuations whereas the random variable  $\tilde{\varepsilon}_{il}$  captures the idiosyncratic component.<sup>12</sup>

We assume that agent l does not necessarily know his own stand-alone valuations. Instead, he receives a signal

$$x_{il} = \theta_i + \eta_{il}$$

where  $\eta_{il}$  is drawn from a Normal distribution with zero mean and variance  $(\beta_i^x)^{-1}$ .

Nature first draws  $\theta \equiv (\theta_1, \theta_2)$  from a bivariate Normal distribution with zero mean and variance-covariance matrix

$$\Sigma_{\theta} = \left[ \begin{array}{cc} \alpha_1^{-1} & \frac{\rho_{\theta}}{\sqrt{\alpha_1 \alpha_2}} \\ \frac{\rho_{\theta}}{\sqrt{\alpha_1 \alpha_2}} & \alpha_2^{-1} \end{array} \right]$$

where the parameter  $\rho_{\theta}$  is the coefficient of linear correlation between  $\tilde{\theta}_1$  and  $\tilde{\theta}_2$ .

For each agent  $l \in [0,1]$  from each side i = 1, 2, Nature then draws a pair  $(\varepsilon_{il}, \eta_{il})$  from a bivariate Normal distribution with zero mean and variance-covariance matrix

$$\Sigma_i = \begin{bmatrix} (\beta_i^u)^{-1} & \frac{\rho_i}{\sqrt{\beta_i^u \cdot \beta_i^x}} \\ \frac{\rho_i}{\sqrt{\beta_i^u \cdot \beta_i^x}} & (\beta_i^x)^{-1} \end{bmatrix}$$

where the parameter  $\rho_i \geq 0$  is the coefficient of linear correlation between  $\tilde{\varepsilon}$  and  $\tilde{\eta}$ . The pairs  $(\varepsilon_{il}, \eta_{il})_{l \in [0,1]}$  are drawn independently across agents from the above distribution and independently from  $(\tilde{\theta}_1, \tilde{\theta}_2)$ .

Platforms are assumed not to possess any private information.

#### Timing.

- At stage 1, platforms simultaneously set access prices on each side.
- At stage 2, after observing the prices  $(p_i^k)_{i=1,2}^{k=A,B}$ , and after receiving the private signal  $x_{il}$ , each agent  $l \in [0,1]$  from each side i=1,2, chooses which platform to join, if any.
- Finally, at stage 3, payoffs are realized.

**Remark**. The above specification has the advantage of being tractable, while at the same time rich enough to capture a variety of situations. Thanks to Normality, the "aggregate state" (i.e., the cross-sectional distribution of preferences and information) is uniquely pinned down by the bivariate variable  $\theta = (\theta_1, \theta_2)$ . The information about  $\theta$  is dispersed so that different agents have different beliefs about  $\theta$ . The pure common-value case where agents on side i have identical preferences over the two platforms but different information about the quality differential  $\theta_i$  is captured as the limit in which  $\beta_i^u \to \infty$  in which case  $v_{il} = z_i \theta_i$  all l. The parameter  $\alpha_i$  is then

 $<sup>^{12}</sup>$  Throughout, we will use tildes " $^{\sim}$ " to denote random variables.

a measure of uncertainty over the degree of horizontal differentiation between the two platforms, as perceived by side i. Letting  $\alpha_1 = \alpha_2$  and  $\rho_{\theta} = 1$  while allowing  $\beta_1^x \neq \beta_2^x$  then permits us to capture situations where the quality differential between the two platforms is the same on each side but the two sides have different information. Letting  $z_i = 0$  on one of the two sides then permits us to capture situations where agents on side i do not care about the intrinsic quality differential between the two platforms but nonetheless have information about the distribution of preferences on the opposite side (as in the case of advertisers who choose which media platform to place ads on entirely on the basis of their expectation of the platform's ability to attract readers and viewers from the opposite side).

More generally, allowing the correlation coefficient  $\rho_{\theta}$  to be different from one permits us to capture situations where the quality differential between the two platforms differs across the two sides (including situations where it is potentially negatively correlated), as well as situations where one side may be able to perfectly predict the behavior of each agent from that side but not the behavior of agents from the opposite side (which corresponds to the limit where,  $\beta_i^x = \infty$ ).

The model can also capture situations in which different users from the same side have different preferences for the two platforms. This amounts to letting the variance of  $\varepsilon_{il}$  be strictly positive, or equivalently,  $\beta_i^u < \infty$ . Depending on the degree of correlation  $\rho_i$  between  $\varepsilon_{il}$  and  $\eta_{il}$  users may then possess more or less accurate information about their own preferences. For example, the case where each agent perfectly knows his own preferences but is imperfectly informed about the preferences of other users (from either side) is captured as the limit in which  $\rho_i \to 1$ . Lastly, the case of independent private values in which users' valuations are independent of one another is captured as the limit in which  $\alpha_i \to \infty$  and  $\beta_i^u < \infty$ .

**Useful estimations**. Having illustrated the flexibility of the model, we now show how its key parameters determine the agents' ability to forecast their own stand-alone valuations, as well as the distribution of such valuations on the other side of the market.

Let

$$V_{il} \equiv \mathbb{E} \left[ \tilde{v}_{il} \mid x_{il} \right]$$

denote the estimated differential in stand-alone valuations for an agent l from side i with information  $x_{il}$ . Observing that  $\mathbb{E}[\tilde{x}_{il} \mid \theta_i] = \theta_i$ ,  $var[\tilde{x}_{il} \mid \theta_i] = 1/\beta_i^x$ ,  $\mathbb{E}[\tilde{x}_{il}] = 0$  and  $var[\tilde{x}_{il}] = (\alpha_i + \beta_i^x)/\alpha_i\beta_i^x$ , from standard projection results, we then have that

$$V_{il} = \kappa_i x_{il} \text{ where } \kappa_i \equiv \frac{cov[\tilde{v}_{il}, \tilde{x}_{il}]}{var[\tilde{x}_{il}]} = z_i \frac{\beta_i^x + \rho_i \alpha_i \sqrt{\beta_i^x/\beta_i^u}}{\alpha_i + \beta_i^x}$$
(1)

Because  $V_{il} = \mathbb{E}[z_i(\tilde{\theta}_i + \tilde{\epsilon}_{il}) \mid x_{il}]$  uniquely pins down not only the differential but also the agent's estimated stand-alone valuation for each of the two platforms, throughout we will refer to  $V_{il}$  as to the estimated stand-alone valuation.

Given the aggregate state  $\theta$ , the cross-sectional distribution of estimated stand-alone valuations

 $V_{il}$  on side i is then given by a Normal distribution with mean  $\kappa_i \theta_i$  and variance

$$var[\tilde{V}_{il} \mid \theta_i] = \frac{\kappa_i^2}{\beta_i^x}$$

Because  $(\kappa_i, \alpha_i, \beta_i^x, \beta_i^u)_{i=1,2}$  are known, we then have that the entire cross-sectional distributions of true stand-alone valuations  $v_{il}$ , as well as the entire cross-sectional distribution of estimated stand-alone valuations  $V_{il}$ , i=1,2, are entirely pinned down by the "aggregate state"  $\theta=(\theta_1,\theta_2)$ . Also note that, from an ex-ante perspective (and hence also from the perspective of the two platforms), the distribution of estimated stand alone valuations  $\tilde{V}_{il}$  on each side i is Normal with mean zero and variance

$$var[\tilde{V}_i] = \kappa_i^2 var[\tilde{x}_{il}] = z_i^2 \frac{(\beta_i^x + \rho_i \alpha_i \sqrt{\beta_i^x/\beta_i^u})^2}{(\alpha_i + \beta_i^x)\alpha_i \beta_i^x}$$
(2)

As one can expect, this variance will play a prominent role in determining the elasticity of the demand functions (and hence the level of the equilibrium prices).

Hereafter, we will measure the agents' ability to forecast their own stand-alone valuations by the variance of their forecast errors about  $\tilde{v}_{il}$ , which is given by

$$var[\tilde{v}_{il} - \tilde{V}_{il}] = var[\tilde{v}_{il} | x_{il}] = \left(1 - \frac{cov(\tilde{v}_{il}, \tilde{x}_{il})^{2}}{var(\tilde{v}_{il}) var(\tilde{x}_{il})}\right) var(\tilde{v}_{il})$$

$$= z_{i}^{2} \frac{\alpha_{i} + \beta_{i}^{u}}{\alpha_{i}\beta_{i}^{u}} - \kappa_{i}^{2} \frac{[\alpha_{i} + \beta_{i}^{x}]}{\alpha_{i}\beta_{i}^{x}} = z_{i}^{2} \frac{\alpha_{i} + \beta_{i}^{u}}{\alpha_{i}\beta_{i}^{u}} - z_{i}^{2} \frac{(\beta_{i}^{x} + \rho_{i}\alpha_{i}\sqrt{\beta_{i}^{x}/\beta_{i}^{u}})^{2}}{(\alpha_{i} + \beta_{i}^{x}) \alpha_{i}\beta_{i}^{x}}$$

$$(3)$$

Not surprisingly, the agents' ability to forecast their own stand-alone valuations increases with  $\rho_i$  and decreases with  $z_i$  and is not necessarily monotone in  $\alpha_i$ ,  $\beta_i^x$  and  $\beta_i^u$ . This is because, holding constant the correlation coefficient  $\rho_i$  between  $\tilde{\varepsilon}_{il}$  and  $\tilde{\eta}_{il}$ , an increase in  $\beta_i^x$  or in  $\beta_i^u$  implies a reduction in the covariance between  $\tilde{\varepsilon}_{il}$  and  $\tilde{\eta}_{il}$ .

Next, consider the agents' ability to forecast the distribution of preferences on the other side of the market. Observe that each agent l from side i observing a signal  $x_{il}$  believes that  $\tilde{\theta}_j$  is Normally distributed with mean

$$\mathbb{E}[\tilde{\theta}_j \mid x_{il}] = \chi_i x_{il} \text{ where } \chi_i \equiv \frac{cov(\tilde{\theta}_j, \tilde{x}_{il})}{var(\tilde{x}_{il})} = \rho_\theta \frac{\beta_i^x}{\alpha_i + \beta_i^x} \sqrt{\frac{\alpha_i}{\alpha_j}}$$

and variance

$$var[\tilde{\theta}_j \mid x_{il}] = \left(1 - \frac{cov(\tilde{\theta}_j, \tilde{x}_{il})^2}{var(\tilde{\theta}_j)var(\tilde{x}_{il})}\right)var(\tilde{\theta}_j) = \left(1 - \rho_\theta^2 \frac{\beta_i^x}{\alpha_i + \beta_i^x}\right) \frac{1}{\alpha_j}$$
(4)

Because the entire cross-sectional distribution of true and stand-alone valuations on side  $j \neq i$  is pinned down by  $\theta_j$ , hereafter, we will measure the ability of each agent l (from side i) to forecast the cross-sectional distribution of preferences on side  $j \neq i$  (without distinguishing between the true preferences  $v_{il}$  and the estimated ones  $V_{il}$ ) with the variance of the agent's forecast errors about  $\tilde{\theta}_j$ , which is given by

$$var[\tilde{\theta}_j - \mathbb{E}[\tilde{\theta}_j | \tilde{x}_{il}]] = var(\tilde{\theta}_j | x_{il}) = \left(1 - \rho_\theta^2 \frac{\beta_i^x}{\alpha_i + \beta_i^x}\right) \frac{1}{\alpha_j}$$
 (5)

Not surprisingly, the agent's ability to forecast  $\tilde{\theta}_j$  increases with  $|\rho_{\theta}|$ ,  $\beta_i^x$ , and  $\alpha_j$ , and decreases with  $\alpha_i$ .

Now observe that, fixing the prior distribution of stand-alone valuations as parametrized by  $(\alpha_1, \alpha_2, \rho_\theta, \beta_1^u, \beta_2^u, z_1, z_2, s_1, s_2)$ , and holding constant the parameters  $(\beta_1^x, \beta_2^x)$  while varying the parameters  $(\rho_1, \rho_2)$ , one can then capture variations in the agents' ability to estimate their own stand-alone valuations, holding constant the agents' ability to forecast the distribution of true (and estimated) stand-alone valuations on the other side of the market. Conversely, by varying the parameters  $(\beta_1^x, \beta_2^x)$  and adjusting the parameters  $(\rho_1, \rho_2)$  accordingly, one can capture variations in the agents' ability to forecast the distribution of true (and estimated) valuations on the other side of the market, while holding constant their ability to estimate their own stand-alone valuations.

Summarizing, the key parameters of the model are  $(\alpha_1, \alpha_2, \rho_\theta, \beta_1^u, \beta_2^u)$  which, together with  $(z_1, z_2, s_1, s_2)$  define the prior distribution of the individual stand-alone valuations, and the scalars  $(\beta_1^x, \beta_2^x, \rho_1, \rho_2)$ , which parametrize the agents' information.

**Remark**. The scalars  $(z_1, z_2)$  only serve the purpose of parametrizing the quality of the agents' information about their own stand-alone valuations relative to the quality of their information about the distribution of stand-alone valuations on the other side of the market. These parameters are not crucial and could have been dispensed with by introducing two separate signals for each agent, one for  $\tilde{\theta}_1$ , the other for  $\tilde{\theta}_2$ . This, however, would have made the subsequent analysis significantly more complicated by essentially requiring that we describe the equilibrium strategies in terms of semi-planes as opposed to simple cut-off rules. The remaining parameters  $(s_1, s_2)$  play a role only for the agent' decision to opt out of the market by not joining any platform.

# 3 Monopoly

As a useful step towards the characterization of the equilibrium in the game with competing platforms, we start by considering the case of a monopolistic market, in which only platform A is active. Given the prices  $(p_1^A, p_2^A)$ , each agent l on each side i then finds it optimal to join the platform only if

$$\mathbb{E}[\tilde{u}_{il}^A \mid x_{il}] + \gamma_i \mathbb{E}[\tilde{m}_i^A \mid x_{il}] - p_i^A \ge 0.$$

Now let  $\gamma_i^- \equiv \min\{\gamma_i; 0\}$ ,  $\gamma_i^+ \equiv \max\{\gamma_i; 0\}$ . It is immediate to see that any agent whose expected stand-alone valuation  $\mathbb{E}[\tilde{u}_{il}^A \mid x_{il}]$  is less than  $(p_i^A - \gamma_i^+)$  finds it dominant not to join, whereas any agent whose expected stand-alone valuation  $\mathbb{E}[\tilde{u}_{il}^A \mid x_{il}]$  is greater than  $p_i^A - \gamma_i^-$  finds it dominant to join. Using  $\mathbb{E}[\tilde{u}_{il}^A \mid x_{il}] = s_i - \kappa_i x_{il}/2$ , we then have that iterated deletion of strictly dominated strategies leads to a pair of thresholds  $\underline{x}_i = \underline{x}_i(p_1^A, p_2^A)$  and  $\bar{x}_i = \bar{x}_i(p_1^A, p_2^A)$  on each side i = 1, 2 such that it is iteratively dominant for each agent l from each side i to join for  $x_{il} < \underline{x}_i$  and not to join for  $x_{il} > \bar{x}_i$ . These observations also suggest existence of a continuation equilibrium in thresholds strategies whereby each agent l from each side i joins if and only if  $x_{il} \leq \hat{x}_i$ . In any such

continuation equilibrium, the measure of agents from side  $j \neq i$  who join is given by

$$m_j^A(\theta_j) = \Pr(\tilde{x}_{jl} \le \hat{x}_j \mid \theta_j) = \Phi(\sqrt{\beta_j^x}(\hat{x}_j - \theta_j)),$$

where  $\Phi$  denotes the c.d.f. of the standard Normal distribution and  $\phi$  its density. This implies that the thresholds  $(\hat{x}_1, \hat{x}_2)$  must jointly solve the following system of conditions

$$G_i(\hat{x}_1, \hat{x}_2) = p_i^A \quad i = 1, 2$$
 (6)

where

$$G_i(x_1, x_2) \equiv s_i - \kappa_i x_i / 2 + \gamma_i \mathbb{E} \left[ \Phi(\sqrt{\beta_j^x} (x_j - \tilde{\theta}_j)) \mid x_i \right]. \tag{7}$$

Note that the function  $G_i(x_1, x_2)$  represents the payoff, gross of payments, of joining platform A for an agent on side i whose signal is equal to the threshold signal  $x_i$  when he expects all users on side  $j \neq i$  to join if and only if their signal is smaller than  $x_j$ .

Next, note that

$$\mathbb{E}\left[\Phi(\sqrt{\beta_j^x}(x_j - \tilde{\theta}_j)) \mid x_i\right] = \Pr\left(\tilde{\eta}_j < x_j - \tilde{\theta}_j \mid x_i\right)$$

$$= \Pr\left(\tilde{\eta}_j + \tilde{\theta}_j < x_j \mid x_i\right)$$

$$= \Phi\left(\sqrt{\frac{\beta_j^x \alpha_j(\alpha_i + \beta_i^x)}{\alpha_1 \alpha_2 + \beta_1^x \alpha_2 + \beta_2^x \alpha_1 + (1 - \rho_\theta^2)\beta_1^x \beta_2^x}} (x_j - \chi_i x_i)\right)$$

where we used the fact that, given  $x_i$ ,  $\tilde{\eta}_j + \tilde{\theta}_j$  follows a Normal distribution with mean  $\chi_i x_i$  and variance

$$\frac{1}{\beta_j^x} + \left(1 - \rho_\theta^2 \frac{\beta_i^x}{\alpha_i + \beta_i^x}\right) \frac{1}{\alpha_j} = \frac{\alpha_1 \alpha_2 + \beta_1^x \alpha_2 + \beta_2^x \alpha_1 + (1 - \rho_\theta^2) \beta_1^x \beta_2^x}{\beta_j^x \alpha_j (\alpha_i + \beta_i^x)}.$$

Now, for any  $i, j = 1, 2, i \neq j$ , let

$$X_{ji}(x_1, x_2) \equiv \sqrt{\frac{\beta_j^x \alpha_j(\alpha_i + \beta_i^x)}{\alpha_1 \alpha_2 + \beta_1^x \alpha_2 + \beta_2^x \alpha_1 + (1 - \rho_\theta^2) \beta_1^x \beta_2^x}} (x_j - \chi_i x_i)$$

$$= \Omega \left( \frac{1}{\rho_\theta} \sqrt{\frac{\alpha_j (\alpha_i + \beta_i^x)}{\beta_i^x}} x_j - \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} x_i \right)$$

where

$$\Omega \equiv \rho_{\theta} \sqrt{\frac{\beta_1^x \beta_2^x}{\alpha_1 \alpha_2 + \beta_1^x \alpha_2 + \beta_2^x \alpha_1 + (1 - \rho_{\theta}^2) \beta_1^x \beta_2^x}}.$$

Denote by  $\rho_x \equiv corr(\tilde{x}_1, \tilde{x}_2)$  the correlation coefficient between any pair of signals from the two sides. Then we have

$$\Omega \equiv \frac{\rho_x}{\sqrt{1 - \rho_x^2}}$$

Hereafter, we will refer to the term  $\Omega$  as to the coefficient of mutual forecastability, for one can show that this term is increasing in each side's ability to forecast the distribution of signals on the

opposite side. As one can expect, this term will play an important role in determining both the monopolist's and the competitive prices.

With the notation introduced above, the function  $G_i(x_1, x_2)$  can then be rewritten as

$$G_i(x_1, x_2) = s_i - \kappa_i x_i / 2 + \gamma_i \Phi(X_{ji}(x_1, x_2)).$$

To ensure that, for any vector of prices, a continuation equilibrium in threshold strategies exists, we assume that the function  $G_i$  is decreasing in  $x_i$ . This is the case, for all  $x_i$ , if and only if the following condition holds, which we assume throughout:

Condition M: The parameters of the model are such that

$$\kappa_i/2 + \gamma_i \Omega \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \phi(0) > 0.$$

Note that the above condition imposes that, when side i values interacting with the other side—namely, when  $\gamma_i > 0$ , the preferences between the two sides be not too negatively correlated. Symmetrically, the condition requires the correlation between  $\tilde{\theta}_1$  and  $\tilde{\theta}_2$  to be sufficiently small when side i dislikes the presence of the other side, that is when  $\gamma_i < 0$ . This is intuitive. Consider the case where  $\gamma_i > 0$ ; if  $\tilde{\theta}_1$  and  $\tilde{\theta}_2$  were strongly negatively correlated, then an increase in the appreciation of agent l from side i of platform A's product could make the agent less willing to join if he expects a significant drop in the participation by agents from side j due to the negative correlation in the preferences of the two sides.

We then have the following preliminary result:

**Lemma 1** For any vector of prices  $p = (p_1^A, p_2^A)$ , there exists at least one solution to the system of conditions given by (6), which implies that a threshold continuation equilibrium always exists.

Now, to guarantee that the continuation equilibrium is unique, for all possible prices, we assume that the strength of the network effects is not too large, given the distribution of the stand-alone valuations, in the sense of Condition Q below, which we assume throughout the rest of the analysis.

Condition Q. The parameters of the model are such that

$$\gamma_{1}\gamma_{2}\sqrt{\frac{\alpha_{1}\alpha_{2}\beta_{1}^{x}\beta_{2}^{x}}{\left(\alpha_{1}+\beta_{1}^{x}\right)\left(\alpha_{2}+\beta_{2}^{x}\right)}}\phi\left(0\right)^{2}-\frac{1}{2}\left(\kappa_{1}\gamma_{2}\sqrt{\frac{\alpha_{2}\beta_{2}^{x}}{\alpha_{2}+\beta_{2}^{x}}}+\kappa_{2}\gamma_{1}\sqrt{\frac{\alpha_{1}\beta_{1}^{x}}{\alpha_{1}+\beta_{1}^{x}}}\right)\Omega\phi\left(0\right)-\frac{\kappa_{1}\kappa_{2}}{4}<0.$$

We then have the following result:

**Lemma 2** For any vector of prices  $(p_1^A, p_2^A)$ , the continuation equilibrium is unique.

The proof in the Appendix first shows that, when conditions M and Q hold, then, for any vector of prices, there exists a unique pair of thresholds  $\hat{x}_i = \hat{x}_i(p_1^A, p_2^A)$ , i = 1, 2, that solve the system of equations defined by the indifference conditions (6). Standard arguments from the

global-games literature based on iterated deletion of strictly dominated strategies then imply that the unique monotone equilibrium defined by the thresholds  $\hat{x}_i$ , i = 1, 2, is the unique equilibrium of the continuation game.

The above result implies that there exists a unique pair of demand functions. For any vector of prices  $(p_1^A, p_2^A)$ , the demand on side i in state  $\theta = (\theta_1, \theta_2)$  is given by  $m_i^A(\theta) = \Phi(\sqrt{\beta_i^x}(\hat{x}_i - \theta_i))$ , while the (unconditional) demand is  $\Phi\left(\sqrt{\frac{\alpha_i\beta_i^x}{\alpha_i+\beta_i^x}}\hat{x}_i\right)$ , where the thresholds  $\hat{x}_i = \hat{x}_i(p_1^A, p_2^A)$ , i = 1, 2, are the unique solution to the system of equations given by (6).

Then consider the choice of prices by the monopolist. For any pair of prices  $(p_1^A, p_2^A)$ , the monopolist's profits are equal to

$$\Pi^{A}(p_{1}^{A}, p_{2}^{A}) = \sum_{i=1,2} p_{i}^{A} \Phi\left(\sqrt{\frac{\alpha_{i} \beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \hat{x}_{i}(p_{1}^{A}, p_{2}^{A})\right).$$

Notice that the system of demand equations (6) defines a bijective relationship between  $(p_1^A, p_2^A)$  and  $(\hat{x}_1, \hat{x}_2)$ . The monopolist's problem can thus also be seen as choosing a pair of thresholds  $(\hat{x}_1, \hat{x}_2)$  so as to maximize

$$\hat{\Pi}^{A}(\hat{x}_{1}, \hat{x}_{2}) \equiv \sum_{i=1,2} G_{i}(\hat{x}_{1}, \hat{x}_{2}) \Phi\left(\sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \hat{x}_{i}\right)$$
(8)

where

$$G_{i}(\hat{x}_{1}, \hat{x}_{2}) = s_{i} - \kappa_{i} \hat{x}_{i} / 2 + \gamma_{i} \Phi \left( X_{ji}(\hat{x}_{1}, \hat{x}_{2}) \right)$$

$$= s_{i} - \kappa_{i} \hat{x}_{i} / 2 + \gamma_{i} \Phi \left( \Omega \left( \frac{1}{\rho_{\theta}} \sqrt{\frac{\alpha_{j} \left( \alpha_{i} + \beta_{i}^{x} \right)}{\beta_{i}^{x}}} \hat{x}_{j} - \sqrt{\frac{\alpha_{i} \beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \hat{x}_{i} \right) \right)$$

denotes the gross surplus of the marginal agent on side i, whose signal is equal to the threshold  $\hat{x}_i$ . Next, for i = 1, 2, let

$$g_i(x) \equiv \left[s_i - \frac{\kappa_i}{2}x + \gamma_i^-\right] \Phi\left(\sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}}x\right)$$

where recall that  $\gamma_i^- \equiv \min\{\gamma_i; 0\}$ ,  $\gamma_i^+ \equiv \max\{\gamma_i; 0\}$ . Throughout, we will assume that the following condition also holds, which guarantees that the optimal prices will be interior.

Condition (W). The parameters of the model are such that, for any  $i, j = 1, 2, j \neq i$ , <sup>14</sup>

$$\max_{x \in \mathbb{R}} g_i(x) > |\gamma_j|.$$

There we used again the property that  $\mathbb{E}\left[\Phi(\sqrt{\beta_i^x}(\hat{x}_i - \tilde{\theta}_i))\right] = \Pr\left(\tilde{\eta}_i < \hat{x}_i - \tilde{\theta}_i\right)$  where  $\tilde{\eta}_i$  and  $\tilde{\theta}_i$  are independent Normal random variables both with zero mean and precisions  $\beta_i^x$  and  $\alpha_i$  respectively.

<sup>&</sup>lt;sup>14</sup>That the function  $g_i$  has a maximum follows from the fact that it is continuous, positive for  $\hat{x}_i < 2(s_i + \gamma_i^-)/\kappa_i$ , negative for  $\hat{x}_i > 2(s_i + \gamma_i^-)/\kappa_i$  and such that  $\lim_{\hat{x}_i \to -\infty} g_i(\hat{x}_i) = 0$ .

Note that Condition (W) is trivially satisfied when  $s_i$  are large enough. The condition simply guarantees that it is always optimal to induce a strictly positive participation rate on both sides, despite the possibility that one side may suffer from the presence of the other side. We then have the following result:

**Lemma 3** There exists a unique vector of prices  $(p_1^A, p_2^A)$  that maximize firm A's profits. These prices are given by  $p_i^A = G_i(\hat{x}_1, \hat{x}_2)$ , i = 1, 2, where  $(\hat{x}_1, \hat{x}_2)$  is the unique solution to the system of conditions given by

$$G_{i}(\hat{x}_{1}, \hat{x}_{2}) \sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \phi \left( \sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \hat{x}_{i} \right) + \frac{\partial G_{i}(\hat{x}_{1}, \hat{x}_{2})}{\partial x_{i}} \Phi \left( \sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \hat{x}_{i} \right) + \frac{\partial G_{j}(\hat{x}_{1}, \hat{x}_{2})}{\partial x_{i}} \Phi \left( \sqrt{\frac{\alpha_{j}\beta_{j}^{x}}{\alpha_{j} + \beta_{j}^{x}}} \hat{x}_{j} \right) = 0.$$

$$(9)$$

To shed light on what lies underneath the first-order conditions for the monopolist's profitmaximizing prices, note that the latter are equivalent to

$$p_{i}^{A} \cdot \left. \frac{dQ_{i}^{A} \left( p_{1}^{A}, p_{2}^{A} \right)}{dp_{i}^{A}} \right|_{Q_{j}^{A} = cont} + Q_{i}^{A} \left( p_{1}^{A}, p_{2}^{A} \right) + \left. \frac{dp_{j}^{A}}{dp_{i}^{A}} \right|_{Q_{j}^{A} = cont} Q_{j}^{A} \left( p_{1}^{A}, p_{2}^{A} \right) = 0$$

$$(10)$$

where  $Q_i^A\left(p_1^A,p_2^A\right)=\mathbb{E}[\tilde{m}_i^A]$  is the demand on side i, as expected by the platform. These first-order conditions are the incomplete-information analogs of the familiar complete-information optimality conditions according to which, at the optimum, profits must not vary when the monopolist changes infinitesimally the price on side i and then adjusts the price on side j so as to maintain the demand on side j constant.

What is interesting here is how incomplete information affects the slope of the demand functions on the two sides and thereby the prices. In particular, while, with complete information, these slopes are the same irrespective of whether they are computed by the platform or by any other agent, this is not the case with dispersed information. To see this, observe that

$$\left. \frac{dQ_i^A \left( p_1^A, p_2^A \right)}{dp_i^A} \right|_{Q_i^A = cont} = \frac{d\mathbb{E}[\tilde{m}_i^A]}{d\hat{x}_i} \left. \frac{d\hat{x}_i}{dp_i^A} \right|_{\hat{x}_j = cont}$$

where

$$\frac{d\mathbb{E}[\tilde{m}_i^A]}{\partial \hat{x}_i} = \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \phi\left(\sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \hat{x}_i\right)$$
(11)

and where

$$\frac{d\hat{x}_i}{dp_i^A}\bigg|_{\hat{x}_j=cont} = \frac{1}{\frac{\partial G_i(\hat{x}_1,\hat{x}_2)}{\partial x_i}} = \frac{1}{-\kappa_i/2 + \gamma_i \frac{\partial \mathbb{E}\left[\tilde{m}_j^A | \hat{x}_i\right]}{\partial \hat{x}_i}}$$

with

$$\frac{d\mathbb{E}\left[\tilde{m}_{j}^{A} \mid \hat{x}_{i}\right]}{d\hat{x}_{i}} = -\Omega\sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}}\phi\left(X_{ji}(\hat{x}_{1}, \hat{x}_{2})\right)$$
(12)

The conditions above highlight a key difference with respect to complete information. Even if the platform adjusts the price on side j in response to a variation in the price on side i so as to maintain the expected demand from side j constant, the slope of the side-i's demand curve naturally depends on the intensity of the side-i's network effects  $\gamma_i$ . The reason is that, when changing  $p_i^A$ , the platform changes the value of the marginal agent  $\hat{x}_i$ . Because of dispersed information, the marginal agent's expectation of the participation rate on side j then also changes, despite the fact that, from the platform's perspective, participation on side j has not changed. This effect, of course, will play an important role for equilibrium prices.

There is a second difference with respect to complete information. The variation in the side-i demand that the platform expects to trigger by changing the price  $p_i^A$  and then adjusting the price  $p_j^A$  to keep the expected side-j demand constant need not coincide with the variation expected by the marginal agent from side j. That is,

$$\frac{\partial \mathbb{E}\left[\tilde{m}_{i}^{A}\right]}{\partial \hat{x}_{i}} \neq \frac{\partial \mathbb{E}\left[\tilde{m}_{i}^{A} \mid \hat{x}_{j}\right]}{\partial \hat{x}_{i}}.$$

This effect in turn impacts the adjustment in the side-j price that the platform must undertake to compensate for the variation in the side-i's demand, as it can be observed from the following decomposition:

$$\left. \frac{dp_{j}^{A}}{dp_{i}^{A}} \right|_{Q_{j}^{A}=cont} = \frac{\partial G_{j}\left(\hat{x}_{1},\hat{x}_{2}\right)}{\partial x_{i}} \left. \frac{d\hat{x}_{i}}{dp_{i}^{A}} \right|_{\hat{x}_{j}=cont}$$

where

$$\frac{\partial G_j\left(\hat{x}_1, \hat{x}_2\right)}{\partial x_i} = \gamma_j \frac{d\mathbb{E}\left[\tilde{m}_i^A \mid \hat{x}_j\right]}{d\hat{x}_i}$$

with

$$\frac{d\mathbb{E}\left[\tilde{m}_{i}^{A} \mid \hat{x}_{j}\right]}{d\hat{x}_{i}} = \frac{1}{\rho_{\theta}} \Omega \sqrt{\frac{\alpha_{i}\left(\alpha_{j} + \beta_{j}^{x}\right)}{\beta_{j}^{x}}} \phi\left(X_{ij}(\hat{x}_{1}, \hat{x}_{2})\right)$$
(13)

The following proposition combines the above observations into a formula for the equilibrium prices that will turn useful when considering competition between the two platforms (the proof follows from the arguments above):

**Proposition 1** The monopolist's profit-maximizing prices, expressed as a function of the demand thresholds they induce, satisfy the following conditions:

$$p_i^A = \left(\frac{\kappa_i}{2} \frac{\mathbb{E}[\tilde{m}_i^A]}{\frac{d\mathbb{E}[\tilde{m}_i^A]}{d\hat{x}_i}}\right) - \gamma_i \left(\frac{d\mathbb{E}\left[\tilde{m}_j^A \mid \hat{x}_i\right]}{d\hat{x}_i} \frac{\mathbb{E}[\tilde{m}_i^A]}{\frac{d\mathbb{E}[\tilde{m}_i^A]}{d\hat{x}_i}}\right) - \gamma_j \left(\frac{\frac{d\mathbb{E}[\tilde{m}_i^A \mid \hat{x}_j]}{d\hat{x}_i}}{\frac{d\mathbb{E}[\tilde{m}_i^A]}{d\hat{x}_i}} \mathbb{E}[\tilde{m}_j^A]\right) \qquad i = 1, 2$$
 (14)

with  $\hat{x}_1$  and  $\hat{x}_2$  implicitly defined by the system of equations given by (6).

One way one can relate the price formula in (14) to its familiar complete-information counterpart is by rewriting it as

$$p_i^A = \frac{\left(\frac{\kappa_i}{2} - \gamma_i \frac{d\mathbb{E}\left[\tilde{m}_j^A | \hat{x}_i\right]}{d\hat{x}_i}\right)}{\frac{d\mathbb{E}\left[\tilde{m}_i^A\right]}{\partial \hat{x}_i}} \mathbb{E}\left[\tilde{m}_i^A\right] - \left(\frac{\gamma_j \frac{d\mathbb{E}\left[\tilde{m}_i^A | \hat{x}_j\right]}{d\hat{x}_i}}{\frac{d\mathbb{E}\left[\tilde{m}_i^A\right]}{d\hat{x}_i}}\right) \mathbb{E}\left[\tilde{m}_j^A\right]$$

and then noticing that the latter is equivalent to

$$p_i^A = -\left(\frac{\frac{dp_i^A}{d\hat{x}_i}\Big|_{\hat{x}_j = cont}}{\frac{d\mathbb{E}[\tilde{m}_i^A]}{\partial \hat{x}_i}}\right) \mathbb{E}[\tilde{m}_i^A] - \left(\frac{\frac{dp_j^A}{d\hat{x}_i}}{\frac{d\mathbb{E}[\tilde{m}_i^A]}{\partial \hat{x}_i}}\right) \mathbb{E}[\tilde{m}_j^A]$$

which is the familiar two-sided-market optimality condition

$$p_{i}^{A} = -\left. \frac{dp_{i}^{A}}{dQ_{i}^{A}} \right|_{Q_{j}^{A} = cont} Q_{i}^{A} - \left. \frac{dp_{j}^{A}}{dQ_{i}^{A}} \right|_{Q_{j}^{A} = cont} Q_{j}^{A}$$
(15)

according to which the monopolist's price is equal to the usual one-sided-market inverse semielasticity (the first term in (15)) adjusted by the effect of a variation in the side-i's participation on side-j's revenues (the second term in (15))—see, for example, Weyl (2010).

The first term in (14) is thus the inverse semi-elasticity of demand in the absence of network effects, expressed in terms of thresholds as opposed to prices. This term is completely standard and entirely driven by the distribution of the estimated stand-alone valuation. In our model it depends on the information structure only because the latter also affects the distribution of the estimated stand-alone valuations.

The third-term in (14) captures the familiar extra cost of raising prices in a two-sided market due to a reduction of demand (or equivalently of price) on the other side. When side j benefits from the presence of side i, that is, when  $\gamma_j > 0$ , this term is known to contribute negatively to the price charged by the monopolist (see e.g., Armstrong, 2006). As discussed above, the novelty relative to complete information comes from the fact that the variation in the side-i demand that the platform expects to trigger by raising  $p_i^A$  now differs from the variation expected by the marginal agent on side j.

The second term in (14) is the most interesting one, for it is absent under complete information. As explained above, this term originates in the fact that a variation in the side-i demand now implies a variation in side-i's expectation about side-j's participation, despite the fact that, from the platform's perspective, the side-j's expected demand does not change, given the adjustment in the side-j price. Whether this new term contributes positively or negatively to the side-i's own price elasticity (and thus ultimately to on the monopolist's profit-maximizing price) depends on the interaction between the sign of the side-i network effect,  $\gamma_i$ , and the sign of the correlation between the two sides' preferences. To understand this, recall that, by lowering the price  $p_i^A$ , the monopolist raises the threshold  $\hat{x}_i$ . Equivalently, it lowers the expected stand-alone valuation of the marginal

agent who is just indifferent between joining and staying home. When valuations are positively correlated between the two sides, this means that the new marginal agent will also expects that fewer agents from the opposite side will like the platform's product and thus join. When side i values participation from side j, this new effect thus reduces the elasticity of the demand on side i and thus contributes to a higher optimal price.

# 4 Competition

We now reintroduce platform B and examine the outcome of the duopoly game where platforms simultaneously compete in prices on each side. Consider the continuation game starting in stage 2 given the prices  $(p_1^A, p_2^A, p_1^B, p_2^B)$ . Assuming full participation (that is, each agent who does not choose platform A chooses platform B), we have that each agent l from each side i = 1, 2 chooses platform A when

$$-V_{il} + \gamma_i \mathbb{E}[\tilde{m}_i^A - \tilde{m}_i^B \mid x_{il}] - p_i^A + p_i^B > 0$$
 (16)

and platform B when the inequality is reversed. Using  $m_i^A + m_i^B = 1$ , i = 1, 2, and (1), Condition (16) can be rewritten as

$$-\kappa_i x_{il} + 2\gamma_i \mathbb{E}[\tilde{m}_i^A \mid x_{il}] - \gamma_i - p_i^A + p_i^B > 0.$$

Now suppose that each agent l from side  $j \neq i$  follows a threshold strategy according to which he chooses A if  $x_{jl} < \hat{x}_j$  and B if  $x_{jl} > \hat{x}_j$ . When this is the case, the measure of agents from side j on platform A is a decreasing function of  $\theta_j$  and is given by

$$m_j^A(\theta) = \Pr(\tilde{x}_{jl} \le \hat{x}_j | \theta_j) = \Phi(\sqrt{\beta_j^x} (\hat{x}_j - \theta_j)).$$

Notice that  $m_j^A(\theta_j)$  decreases with  $\theta_j$ , since a higher  $\theta_j$  means more users with higher stand-alone values for platform B than for platform A. Given the expectation that each agent from side  $j \neq i$  follows such a strategy, each agent l from side i chooses platform A if

$$-\kappa_i x_{il} + 2\gamma_i \mathbb{E}[\Phi(\sqrt{\beta_j^x}(\hat{x}_j - \tilde{\theta}_j)) \mid x_{il}] - \gamma_i - p_i^A + p_i^B > 0$$
(17)

Under Condition (M), the left hand side in (17) is decreasing in  $x_{il}$ . Applying the same logic to each side, we then conclude that a monotone continuation equilibrium is characterized by a pair of thresholds  $(\hat{x}_1, \hat{x}_2)$  that jointly solve

$$-\kappa_i \hat{x}_i + 2\gamma_i \mathbb{E}\left[\Phi(\sqrt{\beta_j^x}(\hat{x}_j - \tilde{\theta}_j)) \mid \hat{x}_i\right] - \gamma_i = p_i^A - p_i^B \qquad i, j = 1, 2, \ j \neq i.$$
(18)

Note that the left-hand side of (18) is the gross payoff differential of joining platform A relative to platform B for the marginal agent  $\hat{x}_i$  on side i when users on both sides follow threshold strategies with respective cutoffs  $\hat{x}_1$  and  $\hat{x}_2$ .

<sup>&</sup>lt;sup>15</sup>As it will become clear below, full participation can be justified by assuming that the stand-alone valuations are sufficiently high—see Proposition 3.

Recognizing that

$$-\kappa_i \hat{x}_i + 2\gamma_i \mathbb{E}\left[\Phi(\sqrt{\beta_j^x}(\hat{x}_j - \tilde{\theta}_j)) \mid \hat{x}_i\right] - \gamma_i = 2G_i\left(\hat{x}_1, \hat{x}_2\right) - 2s_i - \gamma_i$$

where  $G_i$  are the functions defined above for the monopolist case, we then have that many of the properties identified above for the monopolist case carry over to the duopoly case. In particular, for any vector of prices  $p = (p_1^A, p_2^A, p_1^B, p_2^B)$ , there always exists a solution to the system of conditions given by (18), which implies that a threshold continuation equilibrium always exists. Furthermore, under Condition (Q), this continuation equilibrium is the unique continuation equilibrium, which implies that we can associate to any vector of prices a unique system of demands given, in each state  $\theta = (\theta_1, \theta_2)$  by

$$m_i^A(\theta) = \Phi(\sqrt{\beta_i^x} (\hat{x}_i - \theta_i)) = 1 - m_i^B(\theta_i) \quad i = 1, 2.$$

Thus consider the choice of prices by the two platforms. For any  $p = (p_1^A, p_2^A, p_1^B, p_2^B)$ , the two platforms' profits are equal to

$$\Pi^{A}(p_{1}^{A}, p_{2}^{A}, p_{1}^{B}, p_{2}^{B}) = \sum_{i=1,2} p_{i}^{A} \Phi\left(\sqrt{\frac{\alpha_{i} \beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \hat{x}_{i}\right)$$

and

$$\Pi^{B}(p_{1}^{A}, p_{2}^{A}, p_{1}^{B}, p_{2}^{B}) = \sum_{i=1,2} p_{i}^{B} \left( 1 - \Phi\left(\sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \hat{x}_{i}\right) \right)$$

with the thresholds  $(\hat{x}_1, \hat{x}_2)$  uniquely defined by the system (18).

Now fix  $(p_1^B, p_2^B)$  and consider the choice of prices by platform A. Given the bijective relationship between  $(p_1^A, p_2^A)$  and  $(\hat{x}_1, \hat{x}_2)$  given by

$$p_i^A = p_i^B - \kappa_i \hat{x}_i + 2\gamma_i \mathbb{E}\left[\Phi(\sqrt{\beta_j^x}(\hat{x}_j - \tilde{\theta}_j)) \mid \hat{x}_i\right] - \gamma_i$$
$$= p_i^B + 2G_i(\hat{x}_1, \hat{x}_2) - 2s_i - \gamma_i$$

we have that the prices  $(p_1^A, p_2^A)$  constitute a best-response for platform A if and only if the corresponding thresholds  $(\hat{x}_1, \hat{x}_2)$  solve the following problem:

$$\max_{(\hat{x}_1, \hat{x}_2)} \hat{\Pi}^A (\hat{x}_1, \hat{x}_2) \equiv \sum_{i=1,2} \left[ p_i^B + 2G_i (\hat{x}_1, \hat{x}_2) - 2s_i - \gamma_i \right] \Phi \left( \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \hat{x}_i \right)$$
(19)

Arguments similar to those for the monopolist case then easily permit us to verify that, under Condition (Q), for any vector of prices  $(p_1^B, p_2^B)$  there exists a unique vector of prices  $(p_1^A, p_2^A)$  that maximizes platform A's profits. These prices are the unique solution to the system of first-order conditions given by

$$\left[p_i^B + 2G_i\left(\hat{x}_1, \hat{x}_2\right) - 2s_i - \gamma_i\right] \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \phi\left(\sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \hat{x}_i\right) + 2\frac{\partial G_i\left(\hat{x}_1, \hat{x}_2\right)}{\partial x_i} \Phi\left(\sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \hat{x}_i\right) \right) + 2\frac{\partial G_i\left(\hat{x}_1, \hat{x}_2\right)}{\partial x_i} \Phi\left(\sqrt{\frac{\alpha_j \beta_j^x}{\alpha_j + \beta_j^x}} \hat{x}_j\right) = 0.$$

The above conditions are the duopoly analogs of the optimality conditions (10) for the monopoly case; they describe the relation between the profit-maximizing thresholds and the corresponding prices. Following steps similar to those in the previous section, we can then show that the combination of optimal prices and thresholds for platform A must satisfy the following conditions

$$p_i^A = \kappa_i \frac{\mathbb{E}[\tilde{m}_i^A]}{\frac{d\mathbb{E}[\tilde{m}_i^A]}{d\hat{x}_i}} - 2\gamma_i \left( \frac{d\mathbb{E}\left[\tilde{m}_j^A \mid \hat{x}_i\right]}{d\hat{x}_i} \frac{\mathbb{E}[\tilde{m}_i^A]}{\frac{d\mathbb{E}[\tilde{m}_i^A]}{d\hat{x}_i}} \right) + 2\gamma_j \left( \frac{\frac{d\mathbb{E}[\tilde{m}_i^A \mid \hat{x}_j]}{d\hat{x}_i}}{\frac{d\mathbb{E}[\tilde{m}_i^A]}{d\hat{x}_i}} \mathbb{E}[\tilde{m}_j^A] \right)$$
(21)

along with  $p_i^A = p_i^B + 2G_i(\hat{x}_1, \hat{x}_2) - 2s_i - \gamma_i$ , i = 1, 2. The advantage of the above representation is that it highlights the analogy with the monopolist's case (the only difference is that the optimality conditions now apply to the residual demands). It also permits us to identify the unique equilibrium prices that are sustained in equilibrium.

**Proposition 2** There exists a unique symmetric equilibrium. In this equilibrium, the prices that both platforms charge on each side i = 1, 2 are given by

$$p_i^* = \frac{\sqrt{var[\tilde{V}_i]}}{2\phi(0)} + \gamma_i \Omega - \gamma_j \sqrt{1 + \Omega^2}$$
(22)

where

$$var[\tilde{V}_i] = \frac{(\beta_i^x + \rho_i \alpha_i \sqrt{\beta_i^x / \beta_i^u})^2}{(\alpha_i + \beta_i^x) \alpha_i \beta_i^x}$$

is the ex-ante dispersion of estimated stand-alone valuations and where

$$\Omega \equiv \rho_{\theta} \sqrt{\frac{\beta_1^x \beta_2^x}{\alpha_1 \alpha_2 + \beta_1^x \alpha_2 + \beta_2^x \alpha_1 + (1 - \rho_{\theta}^2) \beta_1^x \beta_2^x}} = \frac{\rho_x}{\sqrt{1 - \rho_x^2}}$$

is the coefficient of mutual forecastability between the two sides, with  $\rho_x = corr(\tilde{x}_1, \tilde{x}_2)$ .

As in the monopolist's case, the first term in (22) is the inverse semi-elasticity of the component of the demand on side i that comes from the stand-alone valuations, accounting for the relation between information and estimated valuations. The last two terms in (22) capture the interaction between the network effects and the dispersion of information. In particular, the term  $\gamma_i\Omega$ , which is absent under complete information, captures the effects of dispersed information on side-i own-price elasticity. As in the monopolist's case, whether this term contributes positively or negatively to the equilibrium prices depends on the sign of the network effects  $\gamma_i$  on side i and on the correlation  $\rho_{\theta}$  between the preferences on the two sides (recall that  $sign(\Omega) = sign(\rho_{\theta})$ ). Finally, the third term in (22) captures the cost of increasing the price on side i due to the effect that this has on the platform's profits on the other side of the market. As in the case of complete-information, this effect contributes to a lower equilibrium price when side j benefits from the presence of side i, i.e., when  $\gamma_j > 0$ , and to a higher price when  $\gamma_j < 0$ .

We summarize the above findings in the following corollary.

Corollary 1 As in the complete-information case, equilibrium prices under platform competition (i) increase with the inverse-semi-elasticity of the component of the demand that comes from the estimated stand-alone valuations and (ii) decrease with the intensity of the network effect from the opposite side. However, contrary to the complete-information case, equilibrium prices under dispersed information (a) increase with the intensity of the own-side network effects when preferences between the two sides are positively correlated, and (b) decrease when they are negatively correlated.

A second important observation is that, holding constant the ex-ante distribution of the estimated stand-alone valuations (that is, the first term of the price equation (22)), the equilibrium price on each side depends on the properties of the information structure only through the *coefficient* of mutual forecastability

$$\Omega \equiv \frac{\rho_x}{\sqrt{1 - \rho_x^2}}.$$

This property suggests that equilibrium prices need not be too sensitive to the specific way the information is distributed across the two sides. Fixing the distribution of the estimated stand-alone valuations, any two information structures resulting in the same coefficient of mutual forecastability result in the same equilibrium prices.

This observation is particularly sharp in the case of a market whose primitives are perfectly symmetric under complete information. That is, consider a market where both the intensity of the network effect and the distribution of the stand alone valuations is the same between the two sides, i.e.,  $\gamma_1 = \gamma_2 = \gamma$  and  $var[\tilde{V}_i] = var[\tilde{V}]$ , i = 1, 2. The complete-information equilibrium prices are then given by <sup>16</sup>

$$p_i^c = \frac{\sqrt{var[\tilde{V}]}}{2\phi(0)} - \gamma, \qquad i = 1, 2,$$

which, not surprising, are the same across the two sides.

Perhaps more surprising, under dispersed information, the equilibrium prices continue to be the same across the two sides, even when the distribution of information is not symmetric. This is because, holding constant the distribution of the estimated stand-alone valuations, and assuming that the intensity of the network effect is the same across the two sides, a variation in the quality of information on side i has an identical effect on the elasticity of demand on each of the two sides. To gauge some intuition, consider the case where preferences are perfectly correlated between the two sides so that  $\tilde{\theta}_1 = \tilde{\theta}_2$  almost surely (in which case  $\alpha_1 = \alpha_2$  and  $\rho_{\theta} = 1$ ). Now suppose that information is very precise on side 1, while very imprecise on side 2, so that  $\beta_1^x \to \infty$  while  $\beta_2^x \to 0$ . Because the agents' behavior on side 2 does not vary much with the state  $\theta_2$ , the value of the information held by the side-1 agents is pretty much the same as if side-1 was uninformed about the distribution of the side-2 valuations.

More generally, the result in Proposition 2 implies that shocks that affect the agents's ability to forecast the cross-sectional distribution of valuations in an asymmetric way across the two sides have

<sup>&</sup>lt;sup>16</sup>The formula can be obtained from (22) by taking the limit where  $\alpha_1, \alpha_2 \to \infty$ .

nonetheless a symmetric effect on the equilibrium prices, as long as the intensity of the network effect is the same across the two sides. This is because, holding fixed the ex-ante distribution of estimated stand-alone valuations, the value that each side assigns to being able to predict the distribution of preferences on the opposite side comes entirely from its ability to predict the participation decisions on the opposite side. When the importance that the two sides assign to the presence of the opposite side is the same (that is, when  $\gamma_1 = \gamma_2$ ), in equilibrium the two platforms then equalize the prices over the two sides, despite possibly asymmetries in the distribution of information.

We summarize the above observations, and combine them with a few other comparative statics results for the symmetric case, in the following corollary.

Corollary 2 Consider a market that is perfectly symmetric under complete information in the sense that the distribution of the stand-alone valuations and the importance of the network effects are the same across the two sides  $(var[\tilde{V}_1] = var[\tilde{V}_2] \text{ and } \gamma_1 = \gamma_2 > 0)$ . Then, holding constant the ex-ante distributions of estimated stand-alone valuations (equivalently, the degree of horizontal-differentiation between the two platforms),

- 1. the equilibrium prices are the same on the two sides, despite possible asymmetries in the distribution of information;
- 2. the equilibrium prices are increasing in each side's ability to forecast the distribution of valuations on the opposite side if the correlation of preferences over the two sides is positive and are decreasing otherwise (That is,  $\partial p^*/\partial \beta_i^x > 0$  if  $\rho_{\theta} > 0$  and  $\partial p^*/\partial \beta_i^x < 0$  if  $\rho_{\theta} < 0$ , i = 1, 2).

Turning to the comparative statics with respect to the ex-ante distribution of estimated standalone valuations, observe that the dispersion of estimated stand alone valuations on each side i = 1, 2 is given by

$$var[\tilde{V}_i] = \kappa_i^2 var[\tilde{x}_i] = z_i^2 \frac{(\beta_i^x + \rho_i \alpha_i \sqrt{\beta_i^x / \beta_i^u})^2}{(\alpha_i + \beta_i^x) \alpha_i \beta_i^x}$$

Because equilibrium prices are increasing in  $var[\tilde{V}_i]$ , it is immediate to see that shocks that increase the agents' ability to estimate their own valuations (through an increase in  $\rho_i$ ) or shocks that increase the cross-sectional dispersion of true valuations (through a reduction in  $\beta_i^u$ ) always contribute to higher equilibrium prices. This makes sense, for such shocks contribute to making the two platforms more horizontally differentiated from an ex-ante standpoint, thus softening competition.

We conclude this section with two results that show that, under plausible additional assumptions, the equilibrium prices characterized above (along with the participation decisions they induce) continue to remain equilibrium outcomes when agents can choose to "opt out" of the market, or to multihome by joining both platforms. These results should be interpreted as (minimal) robustness checks aimed at showing that the above characterization results are not unduly driven by the

choice of simplifying the analysis by abstracting from these possibilities. In future work, it would be interesting to extend the analysis to markets where multihoming and partial market-coverage occur in equilibrium.

We start with the following result that pertains our assumption of full market-coverage:

**Proposition 3** There exist finite  $(\underline{s}_i)_{i=1,2}$  such that, for any  $(s_i)_{i=1,2}$  with  $s_i > \underline{s}_i$ , i = 1, 2, the equilibrium in the game where agents must join one of the two platforms is also an equilibrium in the game where agents can "opt out" of the market by choosing not to join any platform.

The reason why the equilibrium prices in the game with compulsory participation need not remain equilibrium prices in the game where agents can opt out of the market is the following. First, when platforms set the prices at the level of Proposition 2, some agents may experience a negative equilibrium payoff and hence prefer to opt out. Because the equilibrium prices  $p_i^*$  in Proposition 2 are independent of  $s_1$  and  $s_2$ , this possibility can be ruled out by assuming that the marginal agents' equilibrium payoffs are positive, which amounts to assuming that  $s_i + \gamma_i/2 \geq p_i^*$ . Under this condition, no agent finds it optimal to opt out, for any agent's equilibrium payoff is at least as high as that of the marginal agents. This condition, however, does not suffice. In fact, platforms may have an incentive to raise one of their prices above the equilibrium levels of Proposition 2 if they expect that, by inducing some agents to opt out, their demand will fall less than that of the other platform, relative to the case where non-participation is not an option. Consider, for example, a deviation by platform A to a vector of prices  $(p_1^A, p_2^A)$  with  $p_1^A > p_1^*$ . Now suppose that, in the unique continuation equilibrium of the game where participation is compulsory, the payoff of the marginal agent  $\hat{x}_1(p_1^A, p_2^A, p_1^*, p_2^*)$  on side 1 is negative (that is, below his outside option). This means that, in the game where participation is voluntarily, some agents in a neighborhood of  $\hat{x}_1(p_1^A, p_2^A, p_1^*, p_2^*)$  may now decide to opt out. Note that some of these agents were joining platform B in the game with compulsory participation. When network effects are positive, this in turn implies that such a deviation may now be profitable for firm A if the measure of agents on side 1 who would have join platform B in the game with compulsory participation and now decide to opt out is larger than the measure of agents who would have joined platform A and now drop out. That is, when the platform expects a larger drop in the rival's demand than in its own (relative to the case where participation is compulsory), then a deviation that was not profitable in the game where participation is compulsory may now become profitable. For this to be the case, however, it must be that the intensity of the network effects is sufficiently strong to prevail on the direct effect coming from the stand-alone valuations. The proof in the Appendix shows that this is never the case when  $s_i$  are sufficiently large.

Next, consider the possibility that agents multihome by choosing to join both platforms. We assume that, by doing so, each agent l from each side i obtains a gross payoff equal to  $2s_i + \gamma_i(m_j^A + \mu_j^B)$ , where  $2s_i = u_{il}^A + u_{il}^B$  is the sum of the stand-alone valuations and where  $\mu_j^B$  is the measure of agents from side j who join platform B without joining platform A (to avoid double counting).

We then have the following result:

**Proposition 4** Consider the variant of the game where agents from each side of the market can multihome, as described above. For any vector of prices  $(p_1^A, p_2^A, p_1^B, p_2^B)$  such that  $p_i^A + p_i^B \ge 2s_i + \gamma_i$  i = 1, 2, there exists a continuation equilibrium where each agent from each side singlehomes. Conversely, such a continuation equilibrium fails to exist for any vector of prices for which  $p_i^A + p_i^B < 2s_i + \gamma_i$ , for some  $i \in \{1, 2\}$ .

Note that the condition in the Proposition simply says that an agent who expects all other agents to singlehome and who decides to multihome experiences a negative payoff. The proof in the Appendix then shows that, when this is the case, then no agent from either side finds multihoming optimal. Note that the result is not trivial, for it also applies to price profiles for which some agents do experience a negative payoff in the unique continuation equilibrium where all agents singlehome. Conversely, when the net utility  $2s_i + \gamma_i - p_i^A - p_i^B$  that an agent obtains by multihoming is positive, then there is no continuation equilibrium where all agents singlehome. Again, this is true even when the payoffs in the continuation game where all agents singlehome may well be strictly positive for all agents.

The following corollary is then an immediate implication of the above result:

Corollary 3 Let  $(p_1^*, p_2^*)$  be the equilibrium prices in the game where multihoming is not possible, as defined in (22), and assume that  $p_i^* \geq \gamma_i + 2s_i$ , i = 1, 2. Assuming that platforms cannot set negative prices, we then have that the equilibrium in the game where agents are not allowed to multihome continues to be an equilibrium in the game where multihoming is possible.

Because equilibrium prices are increasing in the ex-ante dispersion of the estimated stand-alone valuations and because the latter measures the degree of horizontal differentiation between the two platforms, the result in Corollary 3 is consistent with the finding in Armstrong and Wright (2007) that strong product differentiation on both sides of the market implies that agents have no incentive to multihome when prices are restricted to be non-negative (As argued in that paper, and in other contexts as well, the assumption that prices must be non-negative can be justified by the fact that negative prices can create moral hazard and adverse selection problems).

Together, the results in Proposition 3 and Corollary 3 imply that, when the stand-alone valuations of the marginal agents are neither too high nor too low (intermediate  $s_i$ ) and when the two platforms are seen as sufficiently differentiated on both sides of the market (the ex-ante distribution of estimated stand-alone valuations is sufficiently diffuse), then the unique symmetric equilibrium of the baseline game is also an equilibrium in the more general game where agents can multihome and can opt out of the market.

# 5 Implications for advertising and product selection

We now turn to the effects on equilibrium prices of variations in (i) the prior distribution from which stand-alone valuations are drawn and (ii) the quality of the agents' information. The results in this section have implications for various advertising campaigns, as well as for the platforms' incentives to differentiate their products from those of the competitors.

### 5.1 Advertising campaigns

Think of a software firm entering the market with a new operating system. The firm must decide how much information to disclose to the public about the various features of its operating system. We think of these disclosures as affecting both the developers' and the end-users' ability to estimate their own stand-alone valuations (both in absolute value and relative to the operating system produced by the rival incumbent firm), as well as their ability to forecast the distribution of valuations on the other side of the market.

Formally, we think of these disclosure and advertising campaigns as affecting the information available to the two sides of the market, for fixed distribution of true stand-alone valuations. That is, fix the parameters  $(\alpha_1, \alpha_2, \rho_\theta, \beta_1^u, \beta_2^u)$  defining the prior distribution of the individual stand-alone valuations and consider the effects on profits of variations in (i) the agents' ability to estimate their own stand-alone valuations (as captured by variations in  $\rho_i$ ) and (ii) in their ability to forecast the distribution of true (and estimated) stand-alone valuations on the other side of the market (as captured by joint variations in  $(\beta_i^x, \rho_i)_{i=1,2}$  that leave  $var[\tilde{V}_i]$  unchanged but that affect the variance of the forecast errors of  $\tilde{\theta}_i$ , as measured by (5)).

We then have the following result:

**Proposition 5** Informative advertising campaigns that increase the agents' ability to estimate their own stand-alone valuations (as measured by (3)) without affecting their ability to forecast the distribution of such valuations on the other side of the market (as measured by (5)) always increase profits.

Conversely, campaigns that increase the agents' ability to forecast the distribution of (true or estimated) stand-alone valuations on the other side of the market without affecting their ability to estimate their own valuations increase profits if  $\rho_{\theta}(\gamma_1 + \gamma_2) > 0$  and reduce profits otherwise.

The result is quite intuitive. Consider first campaigns that increase the agents' ability to understand their own needs and preferences, without affecting their ability to forecast other agents' preferences. By making agents more responsive to their own idiosyncrasies, such campaigns increase the ex-ante dispersion of estimated stand-alone valuations, thus reducing the semi-price elasticity of the part of the demand on each side that comes from the stand-alone valuations. These campaigns are thus similar to those that increase the degree of horizontal differentiation between the two platforms under complete information. By reducing the intensity of the competition between the

two platforms, such campaigns thus unambiguously contribute to higher prices and hence to higher profits.

Next, consider campaigns whose primary effect is to make agents more informed about what is "hip". That is, while these campaigns do not help agents understand their own needs, they help them better predict the distribution of preferences on the other side of the market. As we show in the Appendix, mathematically, these campaigns impact the coefficient of mutual forecastability  $\Omega$ , without affecting the ex-ante distribution of estimated stand-alone valuations  $var[\tilde{V}_i]$ . From the equilibrium price equation (22), one can then see that, depending on the intensity of the network effects, such campaigns may either increase or decrease the equilibrium prices. Their total effect on equilibrium profits, which in a symmetric equilibrium are given by

$$\Pi^* = \frac{1}{2}(p_1^* + p_2^*) = \frac{1}{2} \left\{ \frac{\sqrt{var[\tilde{V}_1]} + \sqrt{var[\tilde{V}_2]}}{2\phi(0)} + (\gamma_1 + \gamma_2) \left(\Omega - \sqrt{1 + \Omega^2}\right) \right\},$$
(23)

is then determined by (i) the sign of the total network effects  $\gamma_1 + \gamma_2$  and (ii) by whether increasing the agents' ability to forecast the distribution of preferences on the other side (which, by (5), corresponds to an increase in the precision  $\beta_i^x$  of the agents' information) increases or decreases the coefficient of mutual forecastability  $\Omega$ . Because the latter is increasing in the quality of the agents' information  $\beta_1^x$  and  $\beta_2^x$  if and only if preferences are positively correlated between the two sides (that is, if and only if  $\rho_{\theta} > 0$ ), we then have that the effect of such campaigns on profits is positive if and only if the correlation of tastes between the two sides of the market is of the same sign as the sum of the intensity of the network effects (that is if and only if  $\rho_{\theta}(\gamma_1 + \gamma_2) > 0$ ).

To better understand this result, recall that the term  $\gamma_i\Omega$  in the price equation captures the effect of the dispersion of information on side-i's own-price elasticity. From the discussion in the previous section, when network effects are positive and preferences are positively correlated between the two sides, then  $\gamma_i\Omega$  increases in either of the two sides' quality of information (that is in either  $\beta_1^x$  and  $\beta_2^x$ ). This effect comes from the fact that more precise information on side i makes the marginal agent on both sides more responsive to his private information. When preferences are positively correlated and network effects are positive, this effect in turn contributes to a higher equilibrium price on each side by making each side's demand function less elastic.

At the same time, more precise information also implies a higher sensitivity of both demand functions to variations in prices on the opposite side. These effects, which are captured by the terms  $\gamma_j \sqrt{1 + \Omega^2}$  in the price equations, contribute negatively to the equilibrium prices. While the net effect on the equilibrium prices on each side then depends on the relative strengths of the network effects  $\gamma_1$  and  $\gamma_2$ , the net effect on total profits is unambiguously positive when the sum of the network effects is positive (more generally, when it is of the same sign as the correlation of preferences between the two sides). This is because any loss of profits on one side is more than compensated by an increase in profits on the opposite side, as one can see from (23).

What is interesting about the results in the proposition is that they identify two fairly general

channels through which information affects profits, without specifying the particular mechanics by which the campaigns operate. In reality, most campaigns operate through both channels. That is, they impact both the agents' ability to understand their own preferences and their ability to understand what other agents are likely to find attractive. The results in the proposition then indicate that such campaigns unambiguously increase profits in markets where (i) preferences are positively correlated between the two sides and (ii) the sum of the network effects is positive (which is always the case when each side benefits from the presence of the other side). In contrast, in markets where the sum of the network effects is positive but where preferences are negatively correlated between the two sides (or, vice versa), profits may decrease with the agents' ability to forecast other agents' preferences and platforms may find it optimal to conceal part of the information they have.

Note that the above results refer to informative campaigns. They do not apply to campaigns that distorts the average perception the agents have about the quality differential between the two products. These campaigns could be modelled in our framework by allowing the platforms to manipulate the mean of the distributions from which the signals  $x_{il}$  are drawn. However, note that, in our environment where platforms do not possess any private information and where agents are fully rational, the effect of such campaigns on profits is unambiguously negative, for each agent can always "undo" the manipulation by adjusting the interpretation of the information he receives. As discussed in the "signal-jamming" literature (e.g., Fudenberg and Tirole (1986)), platforms may then be trapped into a situation where they have to invest resources in such campaigns, despite the fact that, in equilibrium, such campaigns have no effect on the agents' decisions.

### 5.2 Product selection

We conclude by considering campaigns that impact the distribution from which the true stand-alone valuations are drawn. As anticipated in the Introduction, such campaigns—formally captured by a change in the parameters  $(\alpha_1, \alpha_2, \rho_\theta, \beta_1^u, \beta_2^u)$ —should be interpreted as the choice of how to position a product relative to the one offered by the other platform. For example, an increase in  $\alpha_1$  and  $\alpha_2$  should be interpreted as the choice to enter the market with a product similar to the one provided by the incumbent platform. We then have the following result:

**Proposition 6** Fix the quality of the information on each side of the market (that is, fix  $(\beta_i^x, \rho_i)$ , i = 1, 2). An increase in the similarity between the two products (as captured by an increase in  $(\alpha_1, \alpha_2)$ ) always reduces the equilibrium profits. The same is true for a reduction in the cross-sectional heterogeneity of individual preferences (as captured by an increase in  $(\beta_1^u, \beta_2^u)$ ).

Conversely, an increase in the alignment of preferences between the two sides (as captured by an increase in  $\rho_{\theta}$ ) increases profits if  $\gamma_1 + \gamma_2 > 0$  and reduces them if  $\gamma_1 + \gamma_2 < 0$ .

That both a higher similarity in the products and a smaller relevance of dimensions that are responsible for idiosyncratic appreciations contribute negatively to profits is immediate, for they both contribute to fiercer competition on prices.

The result about the effect of aligning the preferences of the two sides is less obvious. Observe from the price equation (22) that an increase in the alignment of preferences (which amounts to an increase in the coefficient  $\Omega$  of mutual forecastability) may increase prices on one side while decreasing prices on the other side. This is true even if each side benefits from the participation of the other side. The net effect of profits is however always positive if the sum of the network effects is positive, while it is negative otherwise. For example, in a market for media outlets, more alignment in the preferences of viewers and advertisers over the features of competing outlets can be profitenhancing if the viewers' tolerance towards advertising is high, while it may be profit-reducing otherwise.

## 6 Conclusions

We examined the effects of dispersed information on prices and equilibrium profits in a simple, yet flexible, model of platform competition with horizontally differentiated products. The analysis identified a novel channel through which the dispersion of information interacts with the network effects in determining the elasticity of the demands functions. We then showed how equilibrium profits are affected by variations in the prior distribution from which valuations are drawn and by the quality of information available to the two sides. We used these results to shed light on the platforms' incentives to align the preferences of the two sides and to engage in advertising campaigns that affect the agents' ability to predict their own preferences and/or the distribution of preferences on the other side of the market.

In future work, it would be interesting to extend the analysis to accommodate the possibility of price discrimination, whereby each platform grants differential access to the participating population from the opposite side. It would also be interesting to extend the analysis to a dynamic setting with switching costs and investigate the platforms' incentives to price aggressively at the early stages so as to build a user base as a barrier to entry and to future competition. It would also be interesting to investigate how the platforms' pricing strategies affect the dynamics of learning and the speed of technology adoption.

# 7 Appendix

**Proof of Lemma 1.** Fix  $(p_1^A, p_2^A)$ . Under Assumption M,  $G_i(x_1, x_2)$  is a continuous decreasing  $^{17}$  function onto  $\mathbb{R}$  of  $\hat{x}_i$ . Thus for any  $x_2$  there exists a unique value  $x_1 = \xi_1(x_2)$  that solves

To see this note that  $\frac{\partial G_i\left(x_1,x_2\right)}{\partial x_i} = -\kappa_i/2 - \gamma_i \Omega \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \phi\left(X_{ji}(x_1,x_2)\right).$ 

 $G_1(\xi_1(x_2), x_2) = p_1^A$ . Thus consider the function

$$F(x_2) \equiv G_2(\xi_1(x_2), x_2) - p_2^A.$$

This is a continuous function, positive for  $x_2$  small enough and negative for  $x_2$  large enough. Thus a solution to  $F(x_2) = 0$  always exists, which establishes the result.

**Proof of Lemma 2.** To fix ideas, we assume here that  $\gamma_1 \geq 0$ . The proof for the case where  $\gamma_1 < 0 \leq \gamma_2$  is symmetric to the one for the case where  $\gamma_2 < 0 \leq \gamma_1$  which is covered below. Consider again the function  $F(x_2) \equiv G_2(\xi_1(x_2), x_2)$  introduced in the proof of Lemma 1, where  $\xi_1(x_2)$  is the unique solution to  $G_1(\xi_1(x_2), x_2) = p_1^A$ . From the implicit function theorem, and given that  $\partial G_i(x_1, x_2)/\partial x_i < 0$ , we have that

$$sign\left(\frac{dF\left(x_{2}\right)}{dx_{2}}\right)$$

$$= sign\left(\frac{\partial G_{2}\left(\xi_{1}\left(x_{2}\right), x_{2}\right)}{\partial x_{1}} \frac{\partial G_{1}\left(\xi_{1}\left(x_{2}\right), x_{2}\right)}{\partial x_{2}} - \frac{\partial G_{2}\left(\xi_{1}\left(x_{2}\right), x_{2}\right)}{\partial x_{2}} \frac{\partial G_{1}\left(\xi_{1}\left(x_{2}\right), x_{2}\right)}{\partial x_{1}}\right).$$

Using

$$\frac{\partial G_i(x_1, x_2)}{\partial x_i} = -\kappa_i/2 - \gamma_i \Omega \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \phi\left(X_{ji}(x_1, x_2)\right)$$
$$\frac{\partial G_i(x_1, x_2)}{\partial x_j} = \gamma_i \frac{1}{\rho_\theta} \Omega \sqrt{\frac{\alpha_j (\alpha_i + \beta_i^x)}{\beta_i^x}} \phi\left(X_{ji}(x_1, x_2)\right)$$

after some algebra, we obtain that

$$\frac{\partial G_{2}(x_{1}, x_{2})}{\partial x_{1}} \frac{\partial G_{1}(x_{1}, x_{2})}{\partial x_{2}} - \frac{\partial G_{2}(x_{1}, x_{2})}{\partial x_{2}} \frac{\partial G_{1}(x_{1}, x_{2})}{\partial x_{1}} 
= \left(\gamma_{1} \gamma_{2} \sqrt{\frac{\alpha_{1} \alpha_{2} \beta_{1}^{x} \beta_{2}^{x}}{(\alpha_{1} + \beta_{1}^{x})(\alpha_{2} + \beta_{2}^{x})}} \phi\left(X_{12}(x_{1}, x_{2})\right) - \frac{\kappa_{2}}{2} \gamma_{1} \Omega \sqrt{\frac{\alpha_{1} \beta_{1}^{x}}{\alpha_{1} + \beta_{1}^{x}}}\right) \phi\left(X_{21}(x_{1}, x_{2})\right) 
- \frac{\kappa_{1}}{2} \gamma_{2} \Omega \sqrt{\frac{\alpha_{2} \beta_{2}^{x}}{\alpha_{2} + \beta_{2}^{x}}} \phi\left(X_{12}(x_{1}, x_{2})\right) - \frac{\kappa_{1} \kappa_{2}}{4}.$$
(24)

Now we claim that, under Condition Q, the expression in (24) is strictly negative for any  $(x_1, x_2)$ . To see this, suppose, on the contrary, that there exists  $(x_1, x_2)$  for which the sign of the expression in (24) is nonnegative. Consider first the case where  $\gamma_1, \gamma_2, \Omega \geq 0$ . Then for the expression in (24) to be nonnegative, it must be that

$$\gamma_{1}\gamma_{2}\sqrt{\frac{\alpha_{1}\alpha_{2}\beta_{1}^{x}\beta_{2}^{x}}{(\alpha_{1}+\beta_{1}^{x})(\alpha_{2}+\beta_{2}^{x})}}\phi\left(X_{12}(x_{1},x_{2})\right) - \frac{\kappa_{2}}{2}\gamma_{1}\Omega\sqrt{\frac{\alpha_{1}\beta_{1}^{x}}{\alpha_{1}+\beta_{1}^{x}}} > 0$$

Hence, when  $\gamma_i \Omega \geq 0$ ,  $\frac{\partial G_i(x_1, x_2)}{\partial x_i} < 0$  while for  $\gamma_i \Omega < 0$ ,

$$\frac{\partial G_i\left(x_1, x_2\right)}{\partial x_i} \le -\kappa_i/2 - \gamma_i \Omega \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \phi\left(0\right)$$

which is again negative by Assumption M.

which in turn implies that

$$\frac{\partial G_2(x_1, x_2)}{\partial x_1} \frac{\partial G_1(x_1, x_2)}{\partial x_2} - \frac{\partial G_2(x_1, x_2)}{\partial x_2} \frac{\partial G_1(x_1, x_2)}{\partial x_1} \\
\leq \left( \gamma_1 \gamma_2 \sqrt{\frac{\alpha_1 \alpha_2 \beta_1^x \beta_2^x}{(\alpha_1 + \beta_1^x)(\alpha_2 + \beta_2^x)}} \phi\left(X_{12}(x_1, x_2)\right) - \frac{\kappa_2}{2} \gamma_1 \Omega \sqrt{\frac{\alpha_1 \beta_1^x}{\alpha_1 + \beta_1^x}} \right) \phi\left(0\right) \\
- \frac{\kappa_1}{2} \gamma_2 \Omega \sqrt{\frac{\alpha_2 \beta_2^x}{\alpha_2 + \beta_2^x}} \phi\left(X_{12}(x_1, x_2)\right) - \frac{\kappa_1 \kappa_2}{4}.$$
(25)

Because the right-hand side of (25) can also be rewritten as

$$\left(\gamma_{1}\gamma_{2}\sqrt{\frac{\alpha_{1}\alpha_{2}\beta_{1}^{x}\beta_{2}^{x}}{(\alpha_{1}+\beta_{1}^{x})(\alpha_{2}+\beta_{2}^{x})}}\phi\left(0\right) - \frac{\kappa_{1}}{2}\gamma_{2}\Omega\sqrt{\frac{\alpha_{2}\beta_{2}^{x}}{\alpha_{2}+\beta_{2}^{x}}}\right)\phi\left(X_{12}(x_{1},x_{2})\right) - \frac{\kappa_{2}}{2}\gamma_{1}\Omega\sqrt{\frac{\alpha_{1}\beta_{1}^{x}}{\alpha_{1}+\beta_{1}^{x}}}\phi\left(0\right) - \frac{\kappa_{1}\kappa_{2}}{4}$$
(26)

for the sign of the expression in (26) to be nonnegative, by the same reasoning as above, it must be that the sign of the first term in (26) is also strictly positive. It must then be that

$$\left(\gamma_{1}\gamma_{2}\sqrt{\frac{\alpha_{1}\alpha_{2}\beta_{1}^{x}\beta_{2}^{x}}{(\alpha_{1}+\beta_{1}^{x})(\alpha_{2}+\beta_{2}^{x})}}\phi\left(0\right) - \frac{\kappa_{1}}{2}\gamma_{2}\Omega\sqrt{\frac{\alpha_{2}\beta_{2}^{x}}{\alpha_{2}+\beta_{2}^{x}}}\right)\phi\left(0\right) - \frac{\kappa_{2}}{2}\gamma_{1}\Omega\sqrt{\frac{\alpha_{1}\beta_{1}^{x}}{\alpha_{1}+\beta_{1}^{x}}}\phi\left(0\right) - \frac{\kappa_{1}\kappa_{2}}{4} \ge 0$$
(27)

which is impossible when Condition Q holds.

Next assume that  $\gamma_1, \gamma_2 \geq 0 > \Omega$ . Then, by the same arguments as above, the existence of a pair  $(\hat{x}_1, \hat{x}_2)$  for which the sign of the expression in (24) is nonnegative contradicts the assumption that Condition Q holds.

Next, assume that  $\gamma_1, \Omega \geq 0 > \gamma_2$ . It follows that

$$\frac{\partial G_2(x_1, x_2)}{\partial x_1} \frac{\partial G_1(x_1, x_2)}{\partial x_2} - \frac{\partial G_2(x_1, x_2)}{\partial x_2} \frac{\partial G_1(x_1, x_2)}{\partial x_1} \le$$

$$- \frac{\kappa_1}{2} \gamma_2 \Omega \sqrt{\frac{\alpha_2 \beta_2^x}{\alpha_2 + \beta_2^x}} \phi(X_{12}(x_1, x_2)) - \frac{\kappa_1 \kappa_2}{4}$$
(28)

For the expression in the right-hand-side of (28) to be nonnegative, it must then be that

$$-\gamma_2 \Omega \sqrt{\frac{\alpha_2 \beta_2^x}{\alpha_2 + \beta_2^x}} \phi(0) - \frac{\kappa_2}{2} \ge 0$$

which is impossible under condition (M).

Next consider the case where  $\gamma_1 \geq 0 > \Omega, \gamma_2$ . We then have that

$$\frac{\partial G_2(x_1, x_2)}{\partial x_1} \frac{\partial G_1(x_1, x_2)}{\partial x_2} - \frac{\partial G_2(x_1, x_2)}{\partial x_2} \frac{\partial G_1(x_1, x_2)}{\partial x_1} \\
\leq -\frac{\kappa_2}{2} \gamma_1 \Omega \sqrt{\frac{\alpha_1 \beta_1^x}{\alpha_1 + \beta_1^x}} \phi(0) - \frac{\kappa_1 \kappa_2}{4} < 0$$
(29)

where the last inequality is again by Condition (M).

We conclude that the function  $F(\cdot)$  is strictly decreasing which implies that the threshold continuation equilibrium of Lemma 1 is unique. Standard global-game arguments then imply that there do not exist continuation equilibria other than the threshold one, which establishes the result.

**Proof of Lemma 3. Existence.** Because of the bijective relation between  $(p_1^A, p_2^A)$  and  $(\hat{x}_1, \hat{x}_2)$  it suffices to show that there exists a vector of thresholds  $(\hat{x}_1, \hat{x}_2)$  that maximize (8). To see this, note that, for any pair  $(\hat{x}_1, \hat{x}_2)$ ,

$$\hat{\Pi}^{A}(\hat{x}_{1}, \hat{x}_{2}) \equiv \sum_{i=1,2} \left[ s_{i} - \frac{\kappa_{i}}{2} \hat{x}_{i} + \gamma_{i} \Phi\left(X_{ji}(\hat{x}_{1}, \hat{x}_{2})\right) \right] \Phi\left(\sqrt{\frac{\alpha_{i} \beta_{i}^{x}}{\alpha_{i} + \beta_{i}^{x}}} \hat{x}_{i}\right)$$

which means that

$$\sum_{i=1,2} \left[ s_i - \frac{\kappa_i}{2} \hat{x}_i + \gamma_i^- \right] \Phi\left( \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \hat{x}_i \right) \leq \hat{\Pi}^A \left( \hat{x}_1, \hat{x}_2 \right) \leq \sum_{i=1,2} \left[ s_i - \frac{\kappa_i}{2} \hat{x}_i + \gamma_i^+ \right] \Phi\left( \sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}} \hat{x}_i \right)$$

$$(30)$$

Next, consider the function

$$f_i(x_i) \equiv \left[s_i - \frac{\kappa_i}{2}x_i + \gamma_i^+\right] \Phi\left(\sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}}x_i\right)$$

and note that this function is bounded from above but not from below.<sup>18</sup> By looking at the right-hand side of (30), it is then immediate that, for any i=1,2, there exists a finite  $\bar{x}_i$  such that  $\hat{\Pi}^A(\hat{x}_1,\hat{x}_2)<0$  for any  $(\hat{x}_1,\hat{x}_2)$  such that  $\hat{x}_i\geq\bar{x}_i$ . Because the platform can always guarantee itself zero profits by setting prices equal to zero, this means that, to find a maximizer of  $\hat{\Pi}^A(\hat{x}_1,\hat{x}_2)$ , one can restrict attention to pairs  $(\hat{x}_1,\hat{x}_2)$  such that  $\hat{x}_i\leq\bar{x}_i$ , i=1,2.

Next, note that  $\lim_{x_i\to-\infty} f_i(x_i)=0$ . This means that for any  $i=1,2,\ j\neq i$  and  $\varepsilon>0$  arbitrarily small, there exists a finite  $\underline{x}_i$  such that, for any  $(\hat{x}_1,\hat{x}_2)$  with  $\hat{x}_i\leq\underline{x}_i$ ,

$$\hat{\Pi}^{A}(\hat{x}_{1}, \hat{x}_{2}) \leq \varepsilon + \left[s_{j} - \frac{\kappa_{j}}{2}\hat{x}_{j} + \gamma_{j}^{+}\right] \Phi\left(\sqrt{\frac{\alpha_{j}\beta_{j}^{x}}{\alpha_{j} + \beta_{j}^{x}}}\hat{x}_{j}\right)$$
(31)

Now take any  $\hat{x}_i^{\#} \in \arg \max_x g_i(x)$  and note that any such  $\hat{x}_i^{\#}$  is such that  $\hat{x}_i^{\#} > \underline{x}_i$ . This means, for any  $(\hat{x}_1, \hat{x}_2)$  with  $\hat{x}_i \leq \underline{x}_i$ , the inequality in (31) holds whereas the following inequality

$$\hat{\Pi}^{A}(\hat{x}_{1}, \hat{x}_{2}) > g_{i}(\hat{x}_{i}^{\#}) + \left[s_{j} - \frac{\kappa_{j}}{2}\hat{x}_{j} + \gamma_{j}^{-}\right] \Phi\left(\sqrt{\frac{\alpha_{j}\beta_{j}^{x}}{\alpha_{j} + \beta_{j}^{x}}}\hat{x}_{j}\right)$$

$$(32)$$

holds for  $(\hat{x}_i^{\#}, \hat{x}_j)$ . By Condition (W), we then have that, for any i = 1, 2, any pair  $(\hat{x}_1, \hat{x}_2)$  with  $\hat{x}_i \leq \underline{x}_i$ , there exists a pair  $(\hat{x}_1', \hat{x}_2')$  with  $\hat{x}_i' = \hat{x}_i^{\#}$  and  $\hat{x}_j' = \hat{x}_j$  such that

$$\hat{\Pi}^{A}(\hat{x}'_{1},\hat{x}'_{2}) > \hat{\Pi}^{A}(\hat{x}_{1},\hat{x}_{2}).$$

<sup>&</sup>lt;sup>18</sup>This follows from the fact that the standard Normal distribution satisfies the property that  $\lim_{x\to-\infty} x\Phi(x) = 0$ .

Together with the result above, this means that, when looking for maximizers of  $\hat{\Pi}^A(\hat{x}_1, \hat{x}_2)$  one can restrict attention to pairs  $(\hat{x}_1, \hat{x}_2)$  such that  $\underline{x}_i \leq \hat{x}_i$ , i = 1, 2. Because the above is a compact set, and because the function  $\hat{\Pi}^A(\hat{x}_1, \hat{x}_2)$  is continuous and differentiable, this proves that a maximizer to  $\hat{\Pi}^A(\hat{x}_1, \hat{x}_2)$  always exists. Moreover, by construction of the intervals  $[\underline{x}_i, \bar{x}_i]$ , any maximizer is necessarily interior and thus satisfies the first-order conditions in (9).

Uniqueness. To be added.

**Proof of Proposition 2.** By definition, in a symmetric equilibrium,  $p_i^A = p_i^B$ , i = 1, 2. Under Conditions (M), (Q) and (W), the unique continuation equilibrium is then a threshold equilibrium with thresholds  $\hat{x}_1 = \hat{x}_2 = 0$  and expected demands  $\mathbb{E}[\tilde{m}_i^A] = 1/2$ , i = 1, 2. Substituting  $\hat{x}_i = 0$  and  $\mathbb{E}[\tilde{m}_i^A] = 1/2$ , i = 1, 2, into the the formulas for  $d\mathbb{E}[\tilde{m}_i^A]/d\hat{x}_i$ ,  $d\mathbb{E}[\tilde{m}_j^A \mid \hat{x}_i]/d\hat{x}_i$ , and  $d\mathbb{E}[\tilde{m}_i^A \mid \hat{x}_j]/d\hat{x}_i$  (as given by (11), (12) and (13), respectively) and replacing these formulas into the optimality conditions (21), we then have that the equilibrium prices are given by

$$p_{i}^{*} = \frac{\kappa_{i}}{2\sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i}+\beta_{i}^{x}}}\phi\left(0\right)} + \gamma_{i}\Omega - \gamma_{j}\left(\frac{\frac{1}{\rho_{\theta}}\Omega\sqrt{\frac{\alpha_{i}\left(\alpha_{j}+\beta_{j}^{x}\right)}{\beta_{j}^{x}}}}{\sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i}+\beta_{i}^{x}}}}\right)$$

Noticing that

$$\frac{\kappa_i}{\sqrt{\frac{\alpha_i \beta_i^x}{\alpha_i + \beta_i^x}}} = \sqrt{var[\tilde{V}_i]}$$

and that

$$\left(\frac{\frac{1}{\rho_{\theta}}\Omega\sqrt{\frac{\alpha_{i}(\alpha_{j}+\beta_{j}^{x})}{\beta_{j}^{x}}}}{\sqrt{\frac{\alpha_{i}\beta_{i}^{x}}{\alpha_{i}+\beta_{i}^{x}}}}\right) = \sqrt{1+\Omega^{2}}$$

then gives the result.

**Proof of Proposition 3.** First note that, when  $\underline{s}_i > p_i^* - \gamma_i^-$ , in the proposed equilibrium where participation to one of the two platforms is compulsory, each agent obtains more than his outside option (normalized to zero). Now suppose that platform B offers the equilibrium prices and consider the problem faced by platform A (the problem faced by platform B is symmetric). Clearly, for any deviation entailing a reduction in the price offered to each side, one can construct a continuation equilibrium where each agent behaves exactly as in the game where participation is compulsory, in which case the deviation is unprofitable. Next, for any i = 1, 2, let  $x_i^{\#}$  be implicitly defined by

$$s_i + \frac{1}{2}\kappa_i x_i^\# + \gamma_i^- = p_i^*$$

and observe that, no agent from side i receiving a signal  $x_i > x_i^{\#}$  will ever opt out, irrespective of the prices charged by platform A, for, irrespective of the other agents' decisions, he can obtain a positive surplus by joining platform B.

Now observe that the equilibrium prices  $p_i^*$ , i=1,2, are independent of  $s_i$  and that  $x_i^\#$  is strictly decreasing in  $s_i$ , going to  $-\infty$  as  $s_i$  goes to  $+\infty$ . Suppose now that there exists a vector of prices  $(p_1^A, p_2^A)$  such that, in any of the continuation equilibria that follow the selection of the prices  $(p_1^A, p_2^A, p_1^*, p_2^*)$ , platform A is strictly better off than under the monotone equilibrium that follows the selection of the equilibrium prices  $(p_1^*, p_2^*, p_1^*, p_2^*)$ . Clearly, for this to be possible, there must exist  $i \in \{1, 2\}$  such that  $\hat{x}_i(p_1^A, p_2^A, p_1^*, p_2^*) \leq x_i^\#$ , where  $\hat{x}_i(p_1^A, p_2^A, p_1^*, p_2^*)_{i=1,2}$  are the thresholds defined by (18) in the game where participation is compulsory. Finally, let  $x_i^+(p_1^A, p_2^A, p_1^*, p_2^*)$  be implicitly defined by

 $s_i - \frac{1}{2}\kappa_i x_i^+ + \gamma_i^+ = p_i^A$ 

and observe that no agent from side i with signal  $x_i > x_i^+(p_1^A, p_2^A, p_1^*, p_2^*)$  will ever join platform A, irrespective of his beliefs about the other agents' participation decisions. Now, letting side i be the one for which  $\hat{x}_i(p_1^A, p_2^A, p_1^*, p_2^*) \leq x_i^\#$ , observe that, necessarily,

$$x_i^+(p_1^A, p_2^A, p_1^*, p_2^*) < \hat{x}_i(p_1^A, p_2^A, p_1^*, p_2^*) + 2|\gamma_i|/\kappa_i.$$
(33)

To see this, let  $q(\cdot)$  and  $r(\cdot)$  be the function defined by

$$q(x_i) \equiv s_i - \frac{1}{2}\kappa_i x_i + \gamma_i^+ - p_i^A \text{ and}$$

$$r(x_i) \equiv s_i - \frac{1}{2}\kappa_i x_i + \gamma_i \mathbb{E}\left[\Phi(\sqrt{\beta_j^x}(\hat{x}_j - \tilde{\theta}_j)) \mid x_i\right] - p_i^A$$

where, again,  $\hat{x}_i(p_1^A, p_2^A, p_1^*, p_2^*)_{i=1,2}$  are the thresholds defined by (18) in the game where participation is compulsory. Note that, for any  $x_i$ ,

$$0 \le q(x_i) - r(x_i) \le |\gamma_i|.$$

Because  $r(\hat{x}_i) < 0$ , it follows that  $q(x_i) \le |\gamma_i|$ . Given the linearity of  $q(\cdot)$  in  $x_i$ , we then have that the unique solution  $x_i^+$  to  $q(x_i^+) = 0$  must necessarily satisfy (33).

Having established that  $x_i^\#, x_i^+, \hat{x}_i$  all converge (uniformly) to  $-\infty$  as  $s_i \to +\infty$ , we then have that, in the limit as  $s_i \to +\infty$ ,  $m_i^A(p_1^A, p_2^A, p_1^*, p_2^*) \to 0$  and  $m_i^B(p_1^A, p_2^A, p_1^*, p_2^*) \to 1$ , exactly as in the game where participation is compulsory. This means that, when  $s_i$  goes to infinity, i = 1, 2, platform A's payoff given the prices  $(p_1^A, p_2^A, p_1^*, p_2^*)$  under any continuation equilibrium in the game where participation is voluntarily must converge to its' payoff in the unique continuation equilibrium of the game where participation is compulsory. Because the latter is necessarily less then the platform's payoff under the equilibrium prices, and because, by quasiconcavity of payoffs, there exists K, M > 0 such that, in the game where participation is compulsory

$$\Pi^A(p_1^*,p_2^*,p_1^*,p_2^*) - \Pi^A(p_1^A,p_2^A,p_1^*,p_2^*) > K$$

for any  $(p_1^A, p_2^A, p_1^*, p_2^*)$  for which there exists  $i \in \{1, 2\}$  such that  $p_i^A > M$ , we conclude that, no matter the selected continuation equilibrium, any deviation resulting in partial participation is necessarily unprofitable. This completes the proof.

**Proof of Proposition 4.** Recall that each agent l from each side i prefers to join platform A to joining platform B if and only if

$$\mathbb{E}\left[z_i(\tilde{\theta}_i + \tilde{\varepsilon}_{il}) \mid x_{il}\right] + \gamma_i \mathbb{E}\left[\tilde{m}_j^B - \tilde{m}_j^A \mid x_{il}\right] \le p_i^B - p_i^A. \tag{34}$$

The same agent then prefers joining platform A to multihoming if and only if

$$s_i + \frac{1}{2}\mathbb{E}\left[z_i(\tilde{\theta}_i + \tilde{\varepsilon}_{il}) \mid x_{il}\right] + \gamma_i \mathbb{E}\left[\tilde{m}_j^B \mid x_{il}\right] - p_i^B < 0.$$
 (35)

Note that Condition (35) is implied by Condition (34) if and only if

$$2\gamma_i \mathbb{E}\left[\tilde{m}_i^B \mid x_{il}\right] - p_i^A - \gamma_i \mathbb{E}\left[\tilde{m}_i^B - \tilde{m}_i^A \mid x_{il}\right] \le p_i^B - 2s_i \tag{36}$$

In any continuation equilibrium where all agents singlehome  $m_j^B = 1 - m_j^A$ , in which case the inequality in (36) becomes equivalent to  $\gamma_i \leq p_i^A + p_i^B - 2s_i$ . The same conclusion applies to those agents that prefer platform B to platform A. From the results above, we know that the game where multihoming is not possible always admits a continuation equilibrium. We then conclude that, when  $p_i^A + p_i^B \geq \gamma_i + 2s_i$ , such a continuation equilibrium is also a continuation equilibrium in the game where agents can multihome.

Conversely, when  $p_i^A + p_i^B < \gamma_i + 2s_i$ , there exists no continuation equilibrium where all agents singlehome, for, if such equilibrium existed, then it would satisfy  $m_j^B + m_j^A = 1$ . Inverting the inequalities above, we would then have that some agent from side i would necessarily prefer to multihome.

**Proof of Proposition 5.** Recall that the agents' ability to forecast their own stand-alone valuations is measured by the variance of the forecast errors of  $\tilde{v}_{il}$ , which is given by

$$var[\tilde{v}_{il} - \tilde{V}_{il}] = z_i^2 \frac{\alpha_i + \beta_i^u}{\alpha_i \beta_i^u} - z_i^2 \frac{\left(\beta_i^x + \rho_i \alpha_i \sqrt{\beta_i^x / \beta_i^u}\right)^2}{\left(\alpha_i + \beta_i^x\right) \alpha_i \beta_i^x}$$
(37)

whereas their ability to forecast the distribution of true (as well as estimated) stand-alone valuations on the other side of the market is measured by the variance of the agents' forecast errors of  $\tilde{\theta}_j$ , which is given by

$$var[\tilde{\theta}_j - \mathbb{E}[\tilde{\theta}_j | \tilde{x}_{il}]] = \left(1 - \rho_{\theta}^2 \frac{\beta_i^x}{\alpha_i + \beta_i^x}\right) \frac{1}{\alpha_j}.$$

Finally, recall that the ex-ante distribution of estimated stand-alone valuations on each side of the market is Normal with zero mean and variance

$$var[\tilde{V}_i] = z_i^2 \frac{(\beta_i^x + \rho_i \alpha_i \sqrt{\beta_i^x/\beta_i^u})^2}{(\alpha_i + \beta_i^x)\alpha_i \beta_i^x}$$
(38)

Now observe that the equilibrium profits are given by

$$\Pi^A = \Pi^B = \Pi^* \equiv \frac{1}{2}(p_1^* + p_2^*)$$

with

$$p_i^* = \frac{\sqrt{var[\tilde{V}_i]}}{2\phi(0)} + \gamma_i \Omega - \gamma_j \sqrt{1 + \Omega^2}$$

where

$$\Omega \equiv \rho_{\theta} \sqrt{\frac{\beta_1^x \beta_2^x}{\alpha_1 \alpha_2 + \beta_1^x \alpha_2 + \beta_2^x \alpha_1 + (1 - \rho_{\theta}^2) \beta_1^x \beta_2^x}}$$

is the coefficient of mutual forecastability. Because the prior distribution is fixed, so are the parameters  $(\alpha_1, \alpha_2, \rho_\theta, \beta_1^u, \beta_2^u, z_1, z_2)$ . It is then immediate from (37) and (38) that campaigns that increase the agents' ability to forecast their own stand-alone valuations increase the ex-ante dispersion of estimated stand-alone valuations. From the formula for the equilibrium prices, it is then easy to see that, when such campaigns do not affect the agents' ability to forecast the distribution of true (and estimated) stand-alone valuations on the other side of the market (that is, when they leave  $\beta_1^x$  and  $\beta_2^x$  unchanged), they necessarily increase equilibrium prices and hence equilibrium profits.

Next consider campaigns that leave unchanged the agents' ability to forecast their own standalone valuations (and hence the ex-ante dispersion of estimated stand-alone valuations). Then such campaigns increase profits if and only if they increase

$$(\gamma_1 + \gamma_2) \left(\Omega - \sqrt{1 + \Omega^2}\right)$$

which is the case if and only if

$$\frac{\partial \Omega}{\partial \beta_i^x} (\gamma_1 + \gamma_2) \ge 0.$$

Using the fact that  $\Omega$  is increasing in  $\beta_1^x$  and  $\beta_2^x$  if and only if  $\rho_\theta \geq 0$ , we then have that such campaigns increase profits if and only if  $\rho_\theta(\gamma_1 + \gamma_2) \geq 0$ , thus establishing the result.

**Proof of Proposition 6.** The results concerning the comparative statics with respect to  $(\alpha_1, \alpha_2, \beta_1^u, \beta_2^u)$  follow directly from inspecting the formula for the equilibrium prices and observing that the ex-ante dispersion of estimated stand alone-valuations  $var[\tilde{V}_{il}]$  on each side i = 1, 2 decreases with  $(\alpha_i, \beta_i^u)$  and is independent of  $(\alpha_j, \beta_j^u)$ , whereas the coefficient of mutual forecastability  $\Omega$  is independent of  $(\alpha_1, \alpha_2, \beta_1^u, \beta_2^u)$ .

Next, consider the comparative statics with respect to the coefficient of correlation  $\rho_{\theta}$ . The result then follows from observing that

$$\frac{\partial \Pi^*}{\partial \rho_{\theta}} = \frac{1}{2} (\gamma_1 + \gamma_2) \frac{\partial \Omega}{\partial \rho_{\theta}} \left\{ 1 - \frac{\Omega}{\sqrt{1 + \Omega^2}} \right\}$$

which is positive if and only if  $\gamma_1 + \gamma_2 \ge 0$ .

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