

# The Utility Premium and The Risk Premium as Jensen's Gaps: A Unified Approach

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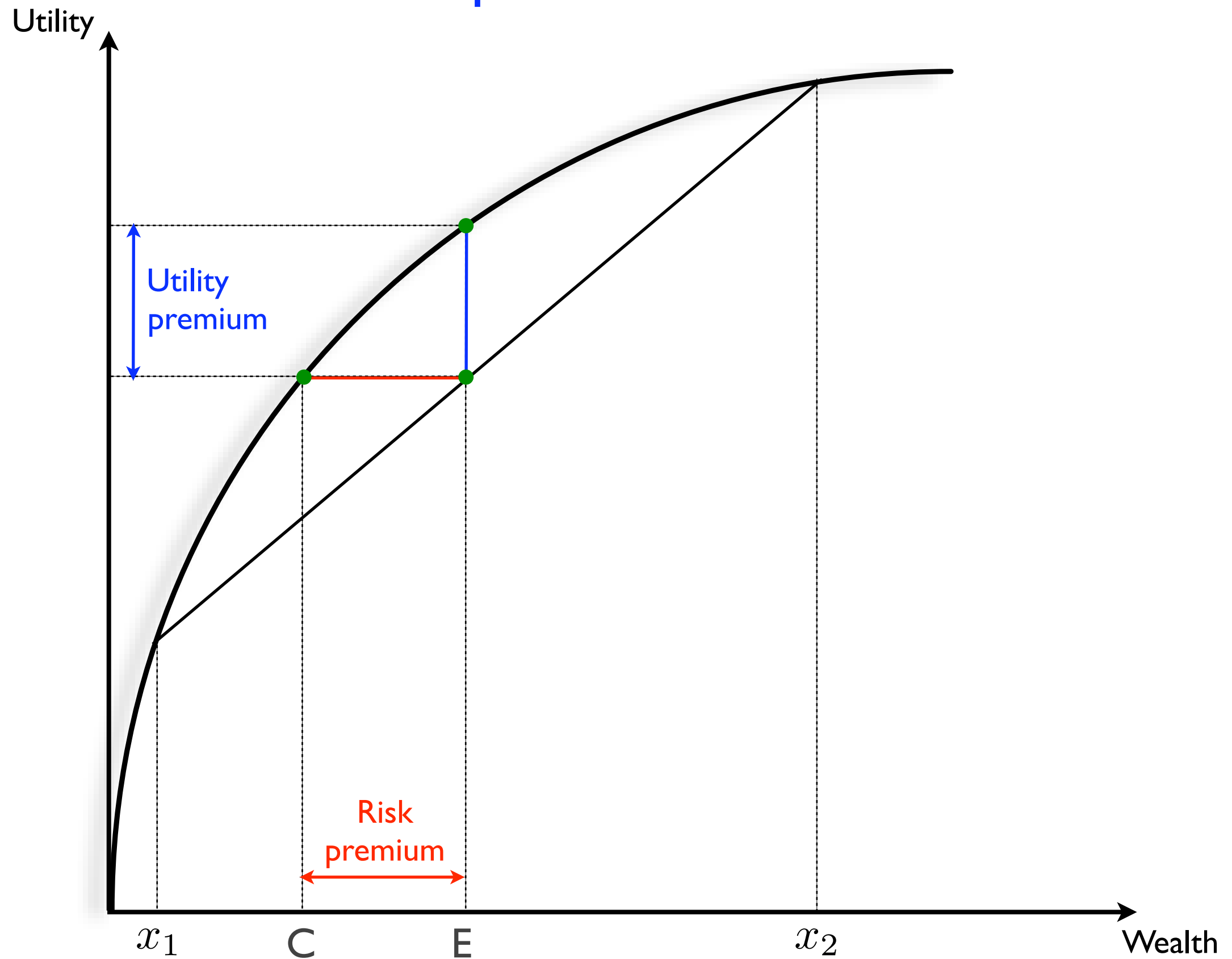
Risk and Choice: A Conference in Honor of Louis Eeckhoudt  
*Toulouse School of Economics*

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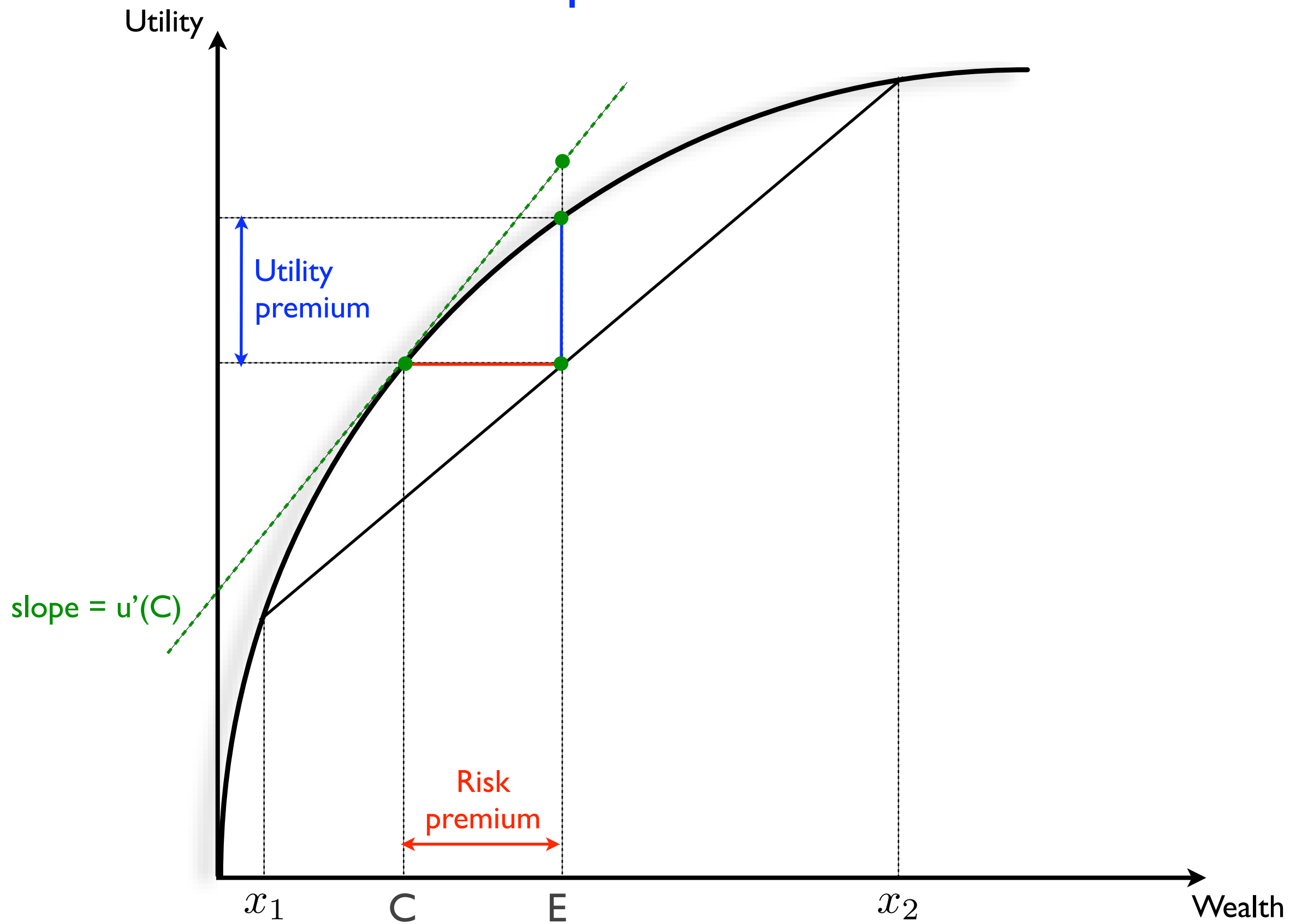
# Motivations

- **Louis (and his close friends)'s *coming out*** : « Pratt's paper has become too famous », one should focus on the utility premium.
- The **current economic crisis** has raised serious challenges to the ways choices under uncertainty are modelled.
  - Expected Utility, Non-Expected Utility, ... etc,
  - Distributional assumptions: Tail events, black swan, ... etc.
- Even without the current crisis, **far too many so-called puzzles remain**, which still beg for explanations:
  - Asset allocation between risk-free and risky assets,
  - The magnitude of the equity risk premium.
- Talking a lot with R. Nock made us discover very promising interactions with **computer science & information theory**

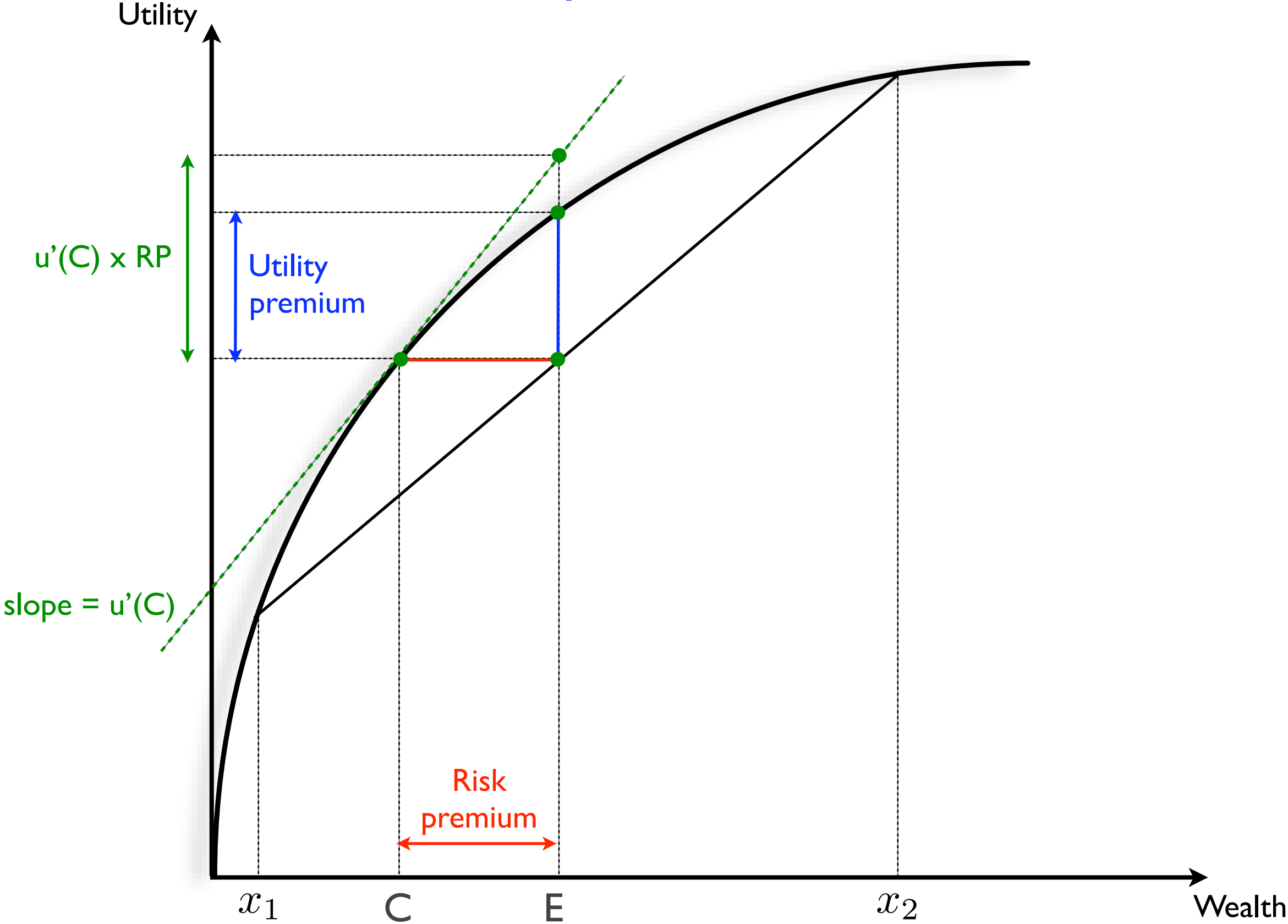
# Graphical Intuition



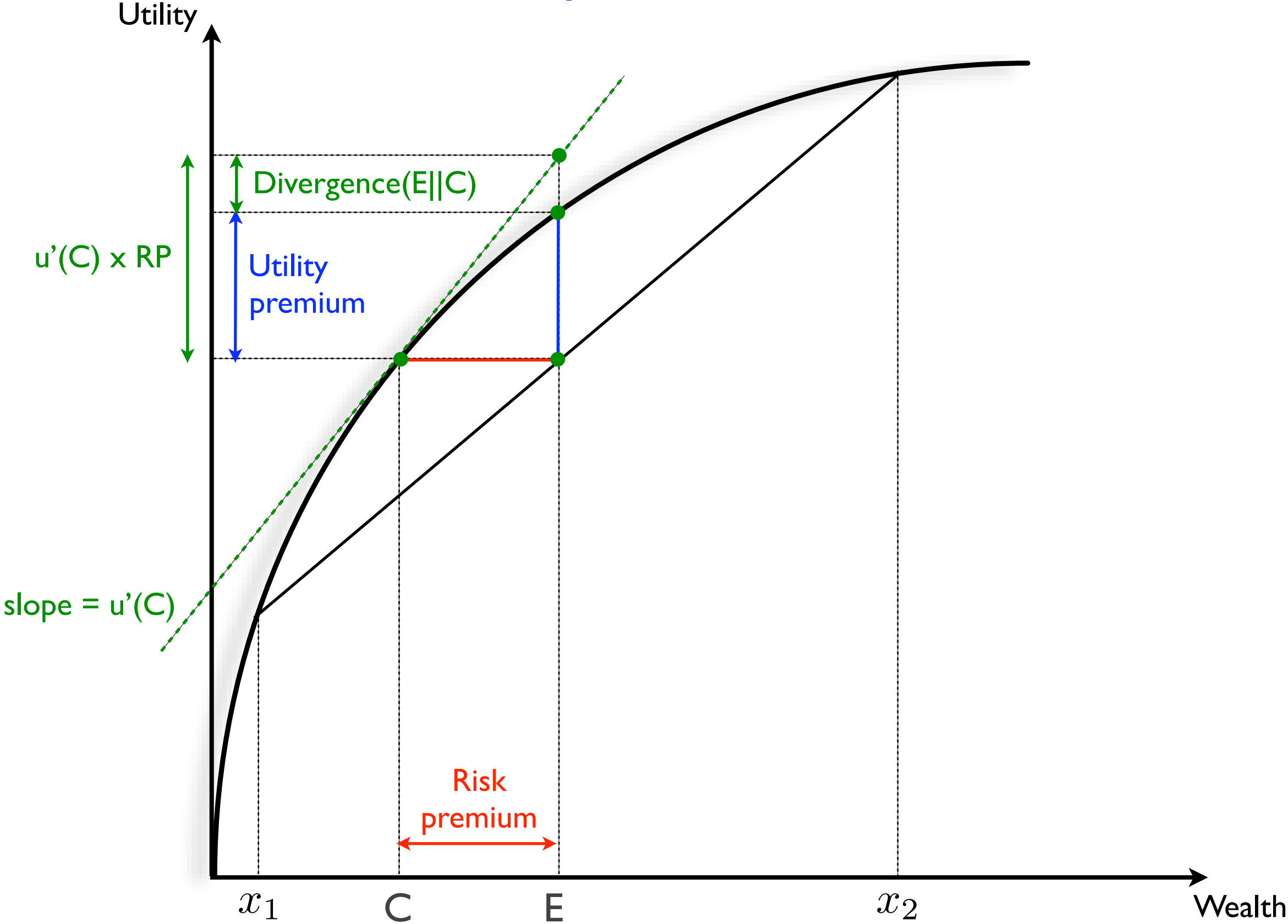
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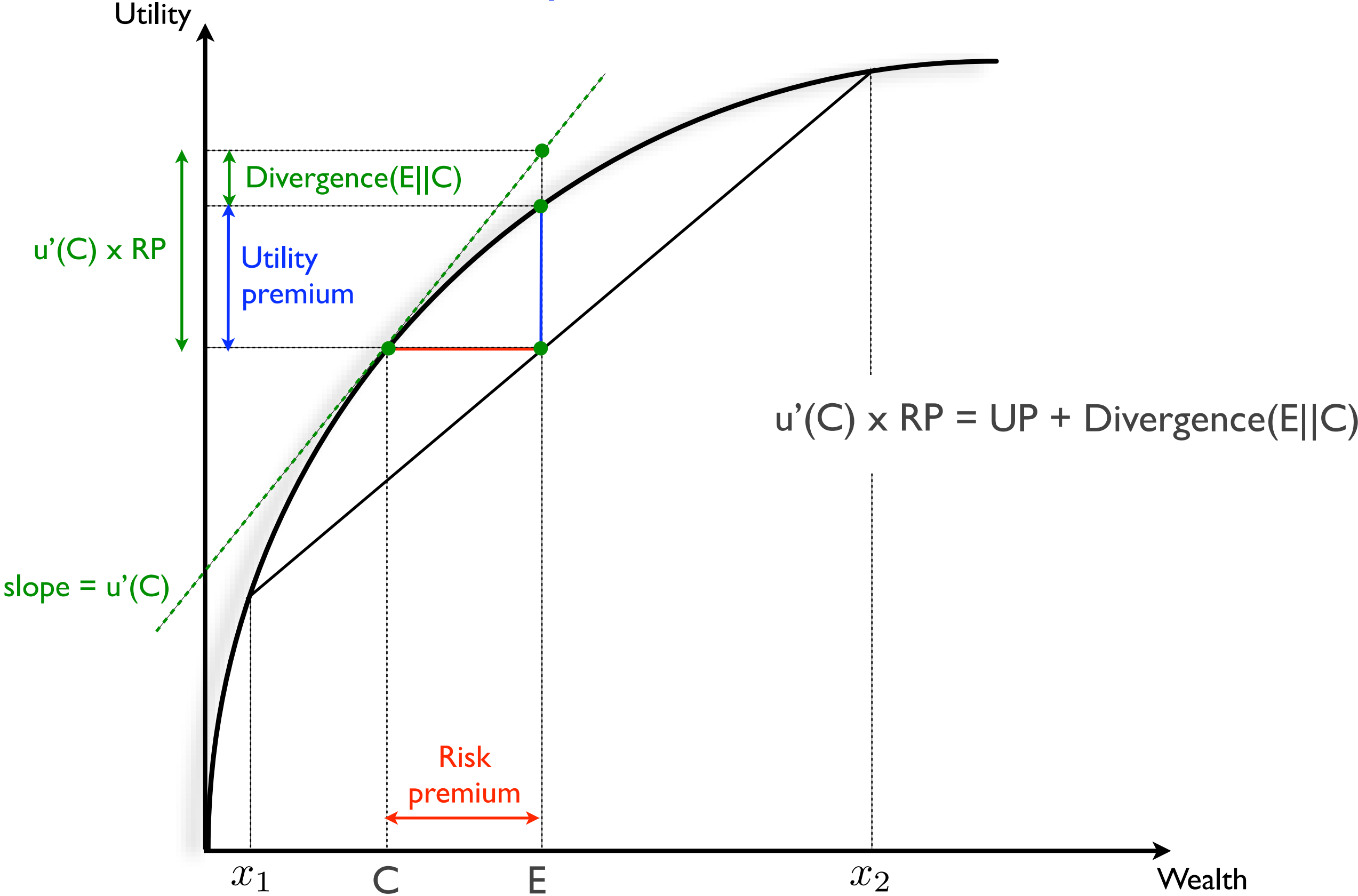
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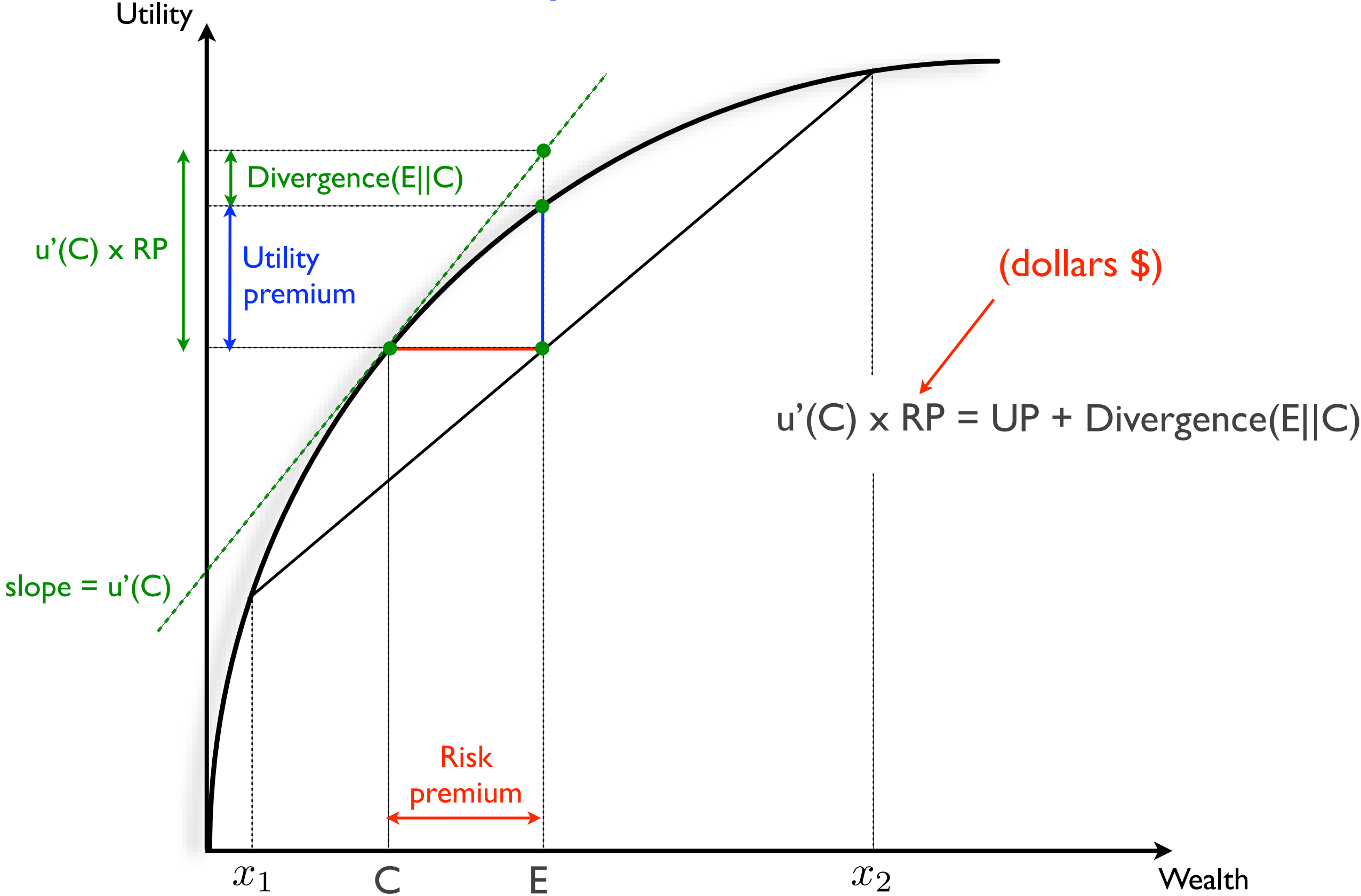
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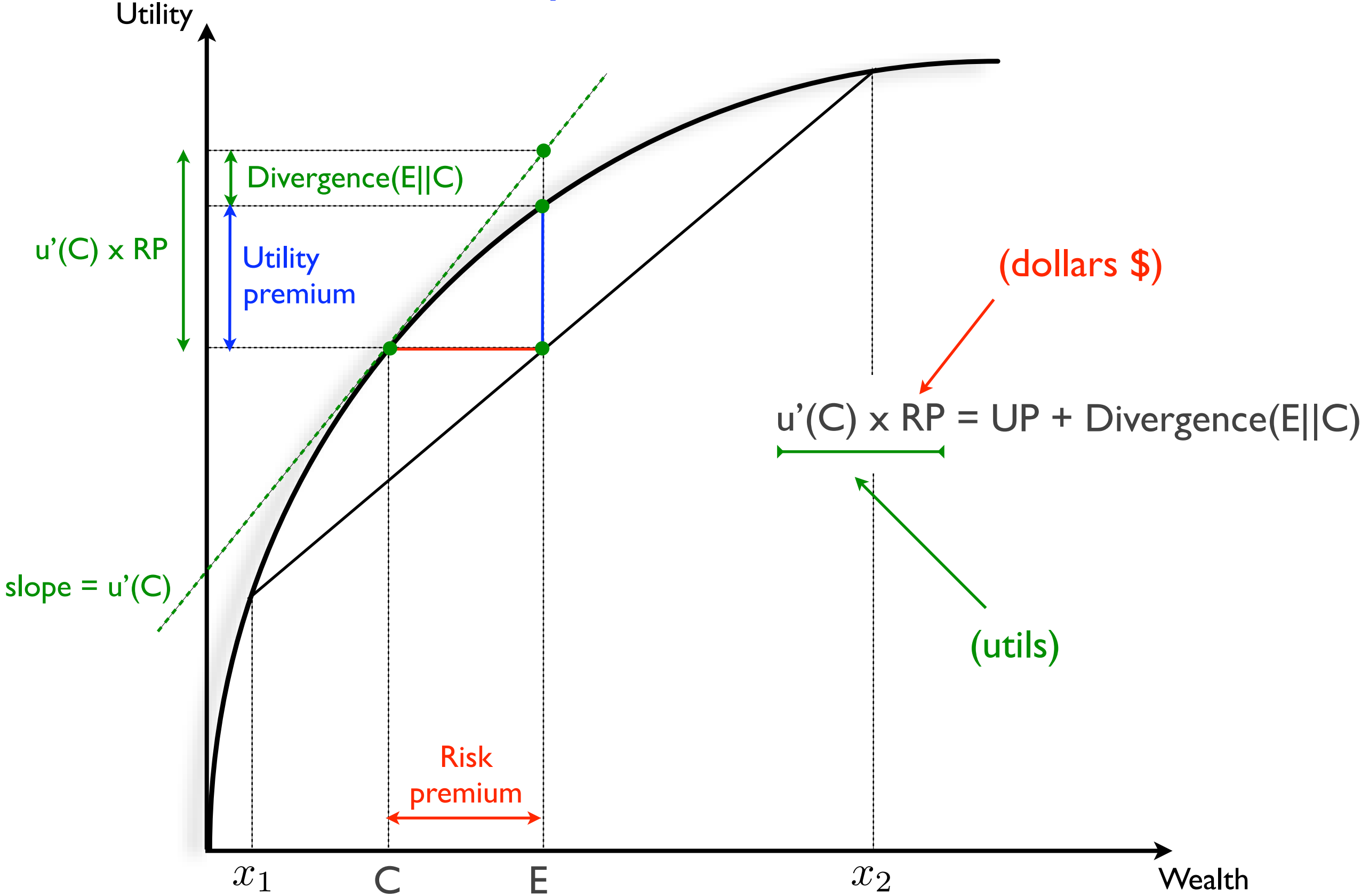


# Graphical Intuition





# Graphical Intuition



# The Framework

- **Notation:**

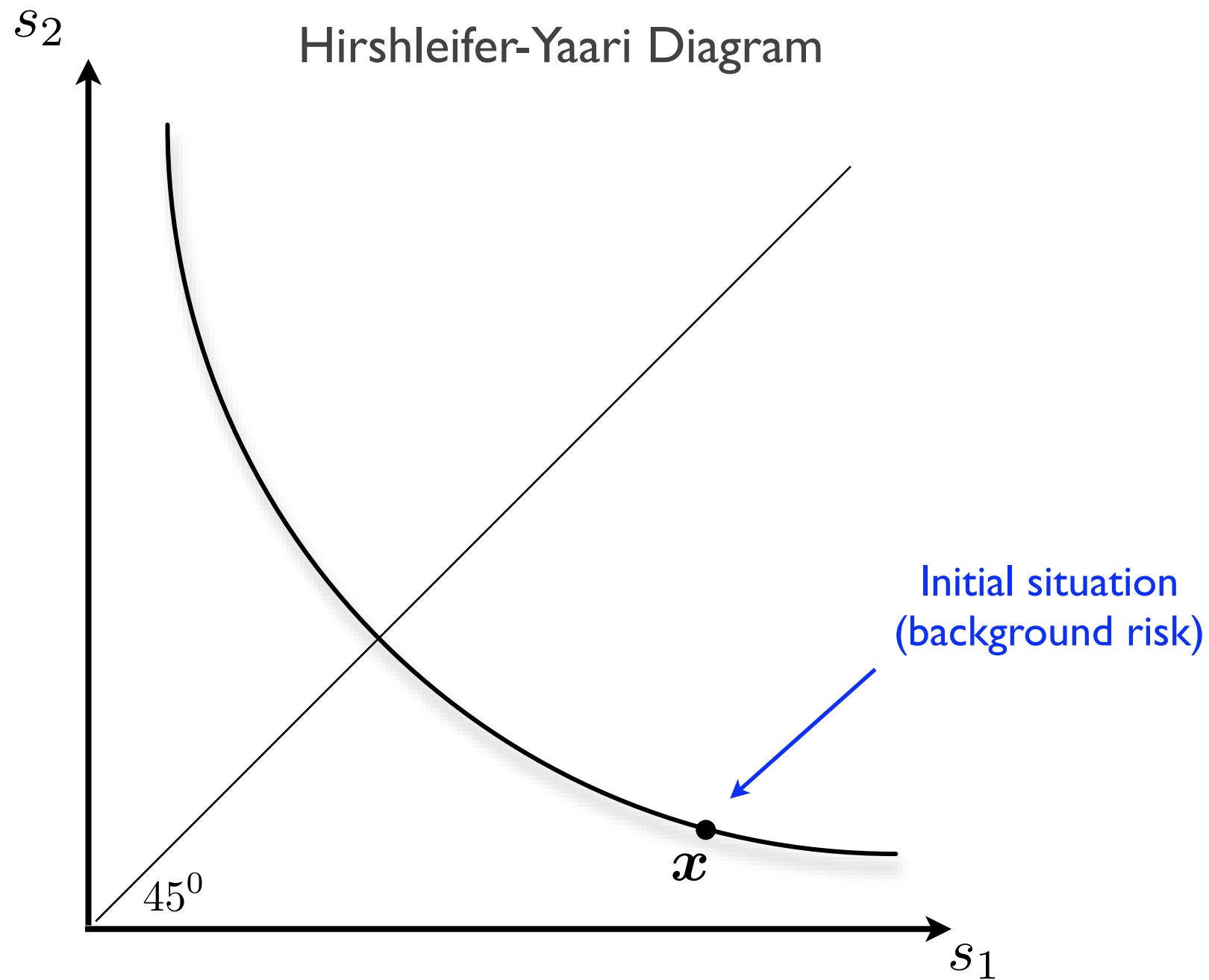
- A finite set of states of the world  $\mathcal{S} := \{s_1, s_2, \dots, s_n\}$
- A outcome  $x_i \in \mathcal{X}$  is associated to each state  $s_i \in \mathcal{S}$
- A risky situation  $\mathbf{x} = (x_1, \dots, x_n) \in \mathcal{X}^n$
- A riskless situation  $\mathbf{c} = (c, \dots, c) \in \mathcal{X}^n$

- **Smooth Preferences:**

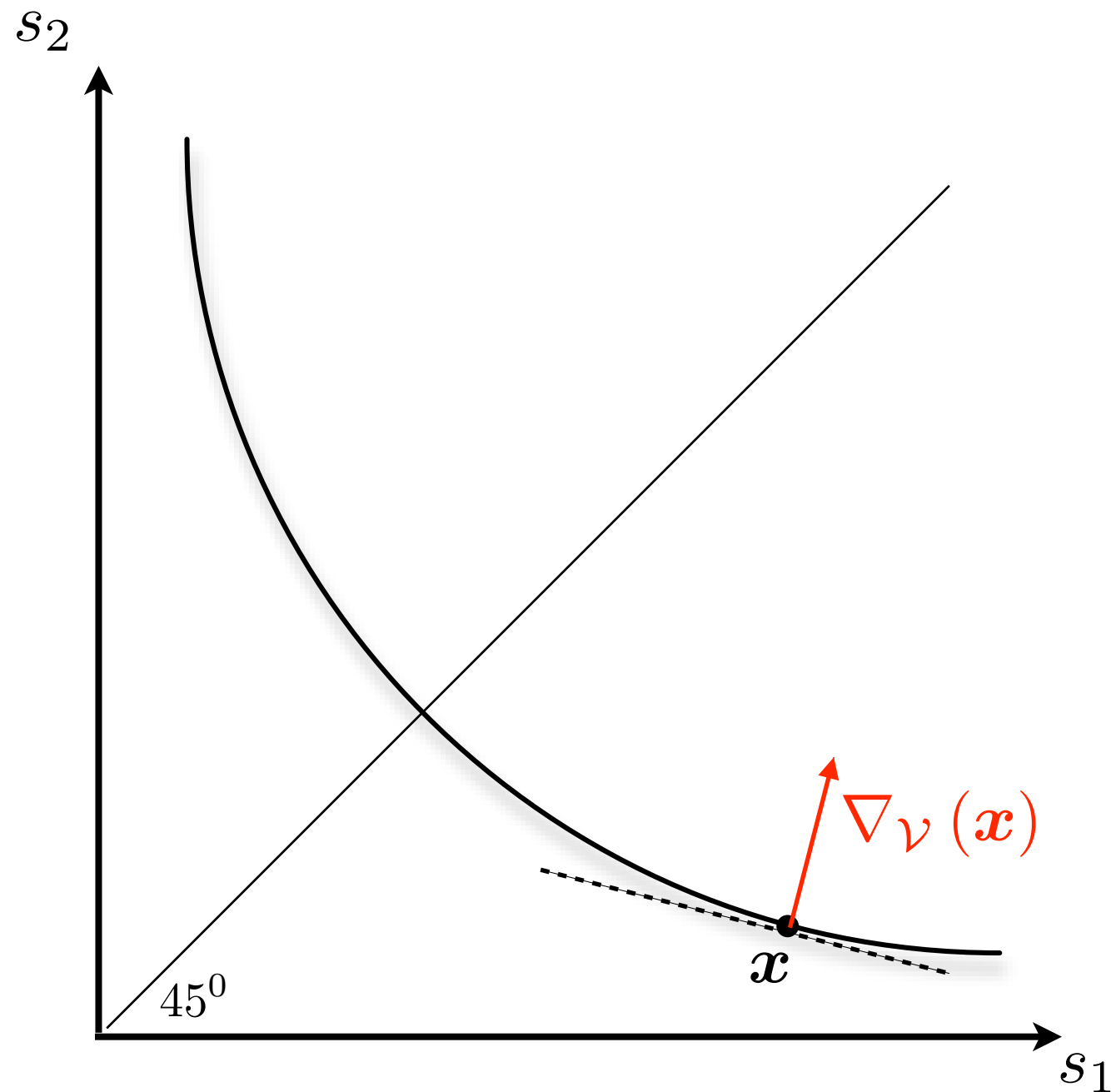
- Complete, transitive, continuous, monotone  
 $\exists \mathcal{V} : \mathcal{X}^n \longrightarrow \mathbb{R}$  so that  $\mathbf{x} \succeq \mathbf{y} \iff \mathcal{V}(\mathbf{x}) \geq \mathcal{V}(\mathbf{y}), \forall \mathbf{x}, \mathbf{y} \in \mathcal{X}^n$
- We assume that  $\mathcal{V}$  is differentiable

- **Expected Utility Model:**  $\mathcal{V}_{\text{EU}}(\mathbf{x}) = \sum_{i=1}^n p_i u(x_i), \forall \mathbf{x} \in \mathcal{X}^n$

# The Framework



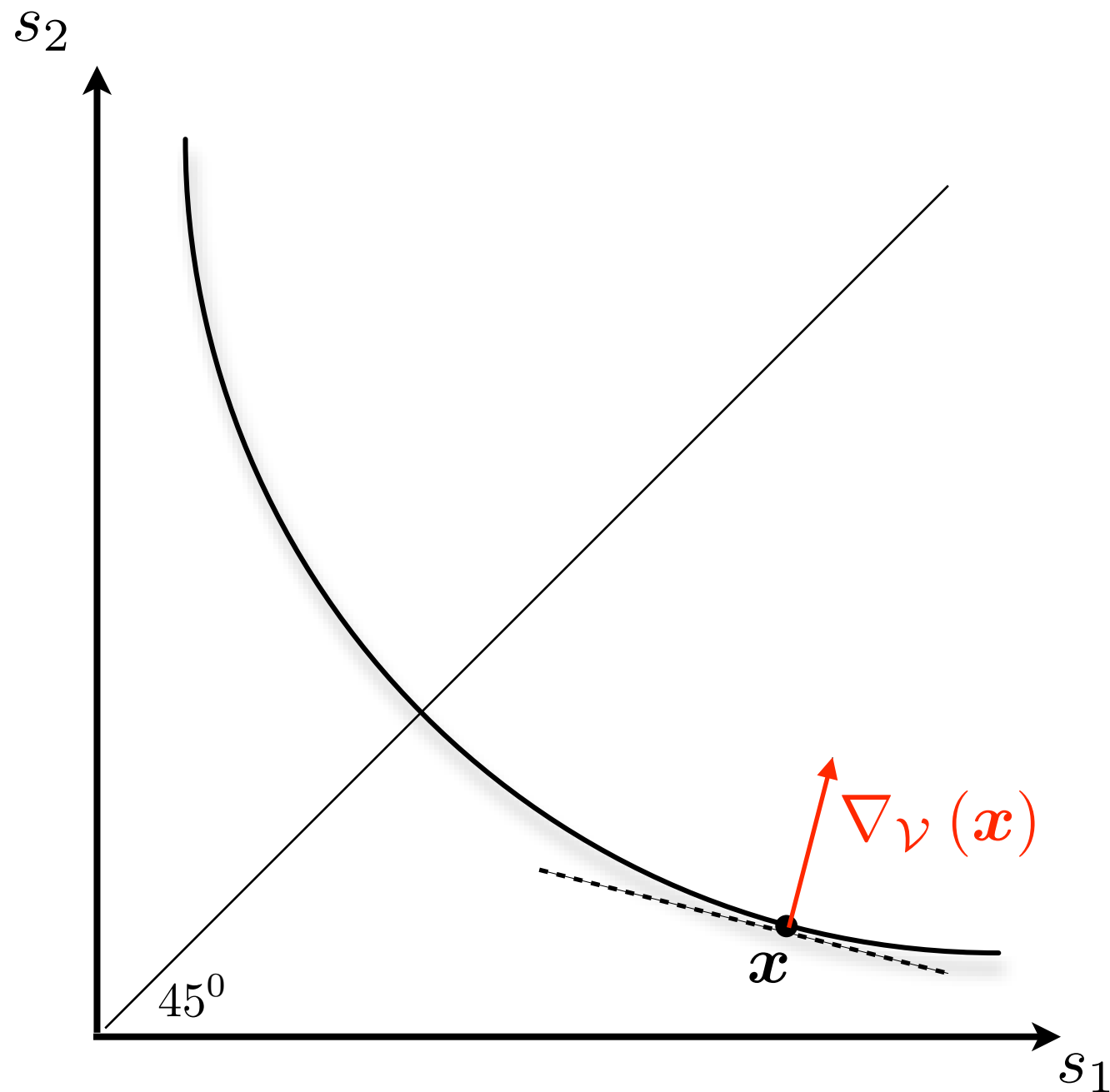
# The Framework



- The gradient of  $\mathcal{V}$  at  $x$ :

$$\nabla_{\mathcal{V}}(x) = \left( \frac{\partial \mathcal{V}(x)}{\partial x_1}, \frac{\partial \mathcal{V}(x)}{\partial x_2}, \dots, \frac{\partial \mathcal{V}(x)}{\partial x_n} \right)$$

# The Framework



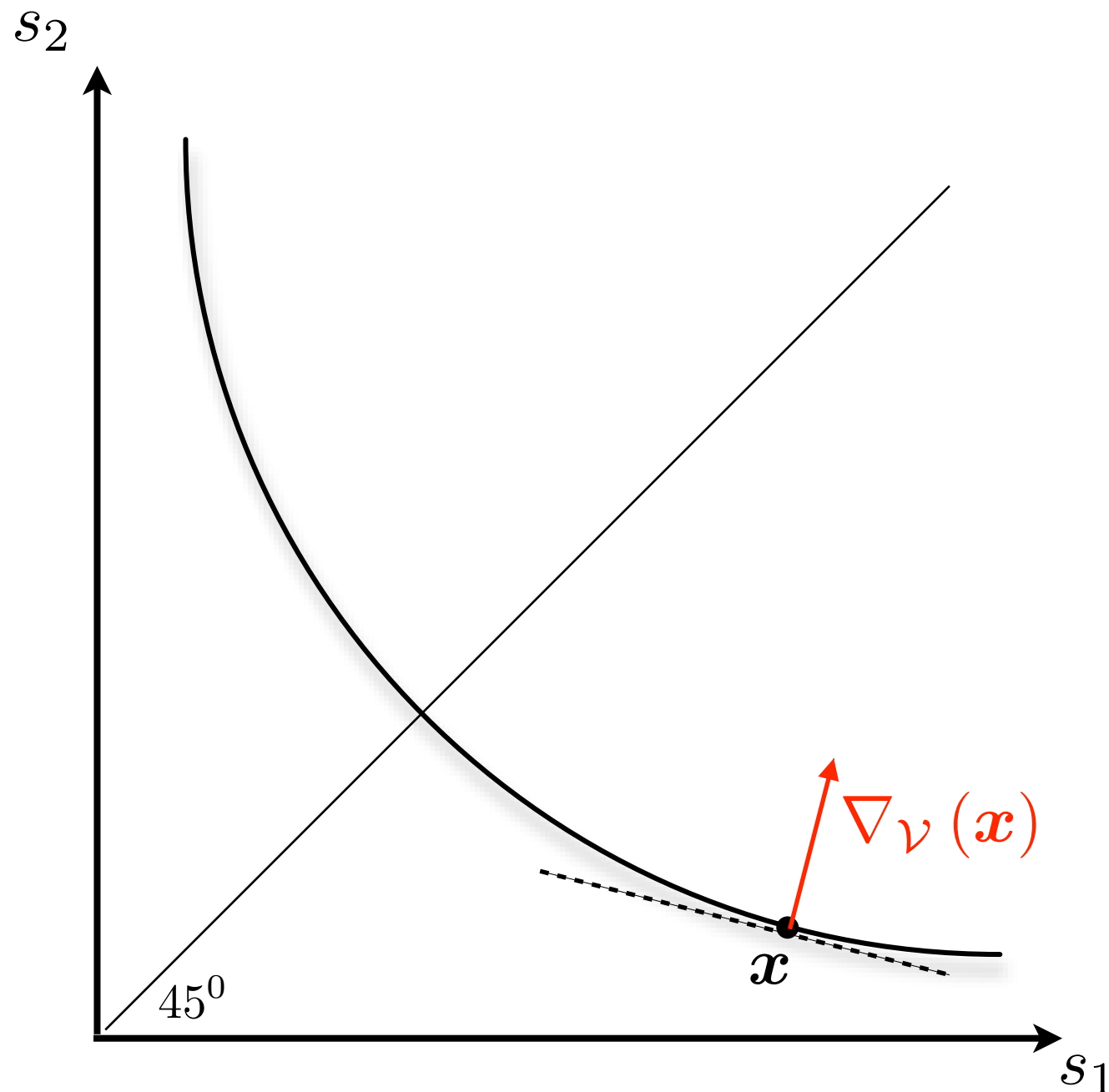
- The gradient of  $V$  at  $x$ :

$$\nabla_V(x) = \left( \frac{\partial V(x)}{\partial x_1}, \frac{\partial V(x)}{\partial x_2}, \dots, \frac{\partial V(x)}{\partial x_n} \right)$$

- The risk-neutral probabilities :

$$\pi(x) = \frac{\nabla_V(x)}{\nabla_V(x) \cdot \mathbf{1}}$$

# The Framework



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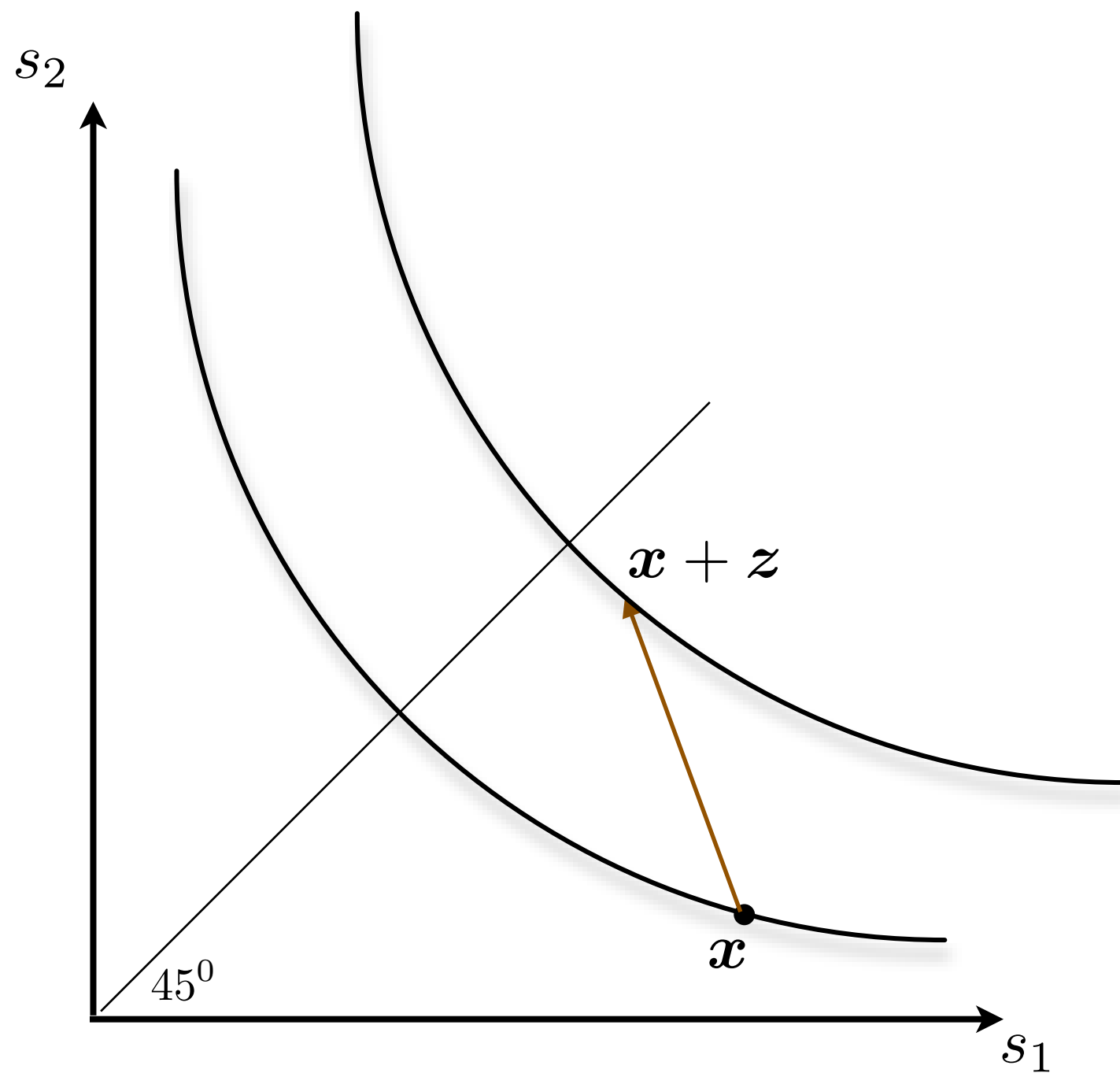
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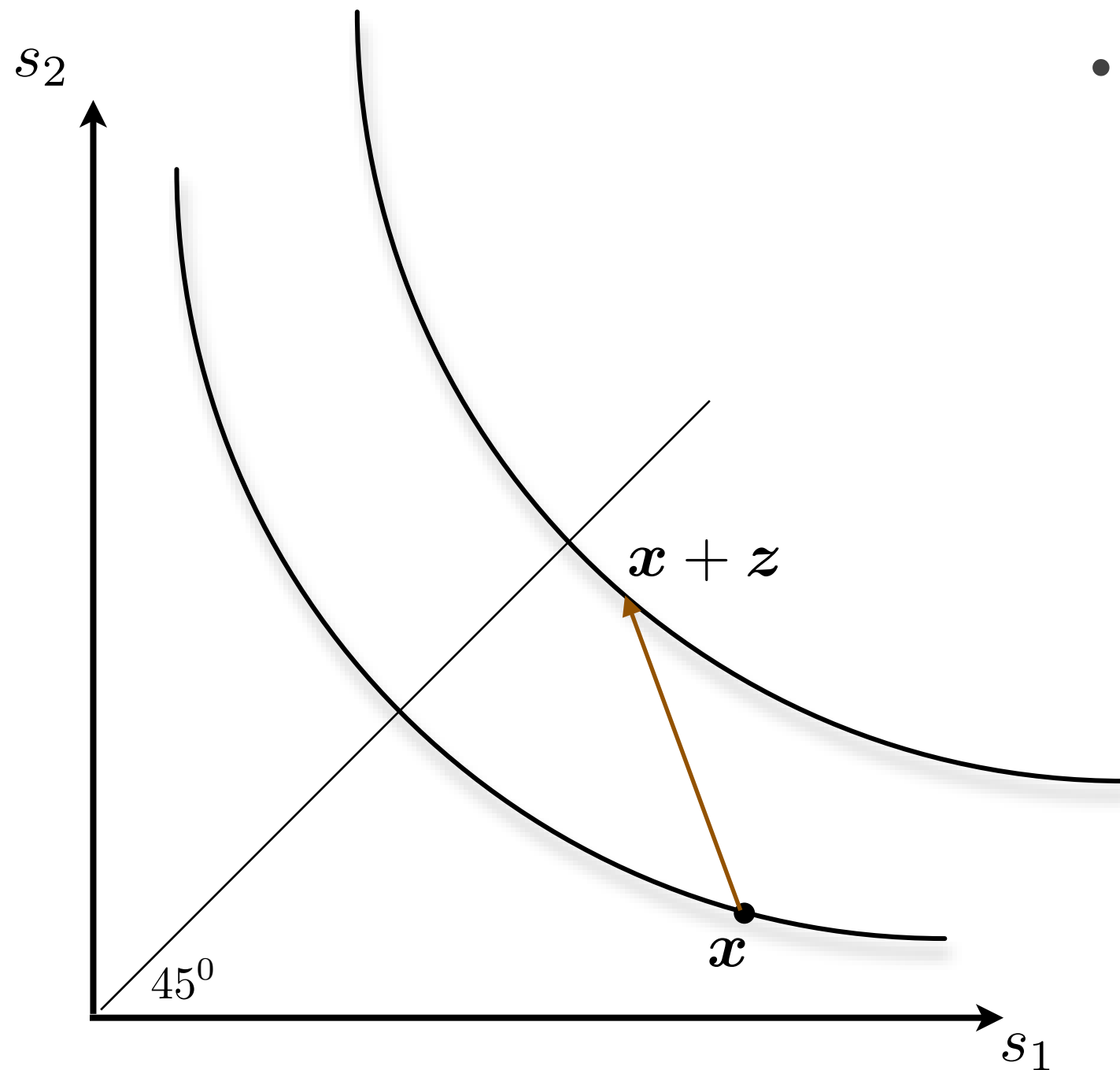
- If EU and riskless initial situation :

Risk-neutral prob. = True prob.

# The Framework



# The Framework

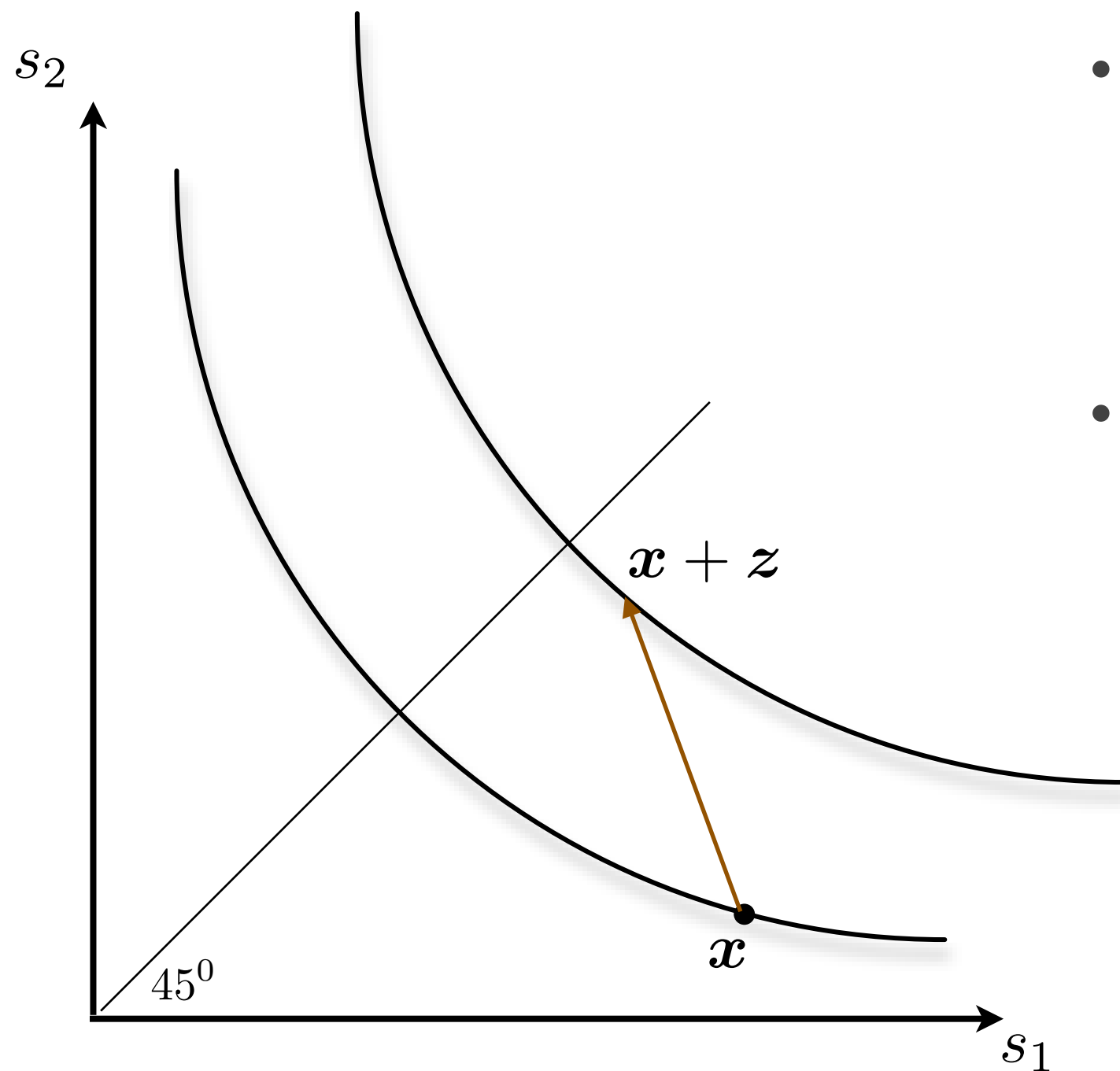


- The Utility Premium :

$$\mathcal{J}(z; x) = \mathcal{V}(x + z) - \mathcal{V}(x)$$



# The Framework



- The Utility Premium :

$$\mathcal{J}(z; x) = \mathcal{V}(x + z) - \mathcal{V}(x)$$

- The Risk Premium :

$$\mathcal{V}(x + z) = \mathcal{V}(x + \mathcal{C}(z; x) \cdot \mathbf{1})$$

where  $\mathcal{C}(z; x) = \mathbb{E}_{\pi(x)}(z) - \mathcal{R}(z; x)$

# Utility Premium for Small Risks

## Expected Utility Model

**Proposition 2.** *Consider a twice differentiable function  $u : \mathcal{X} \longrightarrow \mathbb{R}$ , an initial riskless situation  $\mathbf{c} \in \mathcal{X}^n$  and a small risk  $\mathbf{z} \in \mathcal{X}^n$ . We obtain the following approximation for the utility premium:*

$$\mathcal{J}_{\text{EU}}(\mathbf{z}; \mathbf{c}) \approx u'(\mathbf{c}) \left[ \mathbb{E}(\mathbf{z}) - \frac{1}{2} a(\mathbf{c}) \mathbb{E}(\mathbf{z}^2) \right] .$$

where  $a(\mathbf{c}) = -u''(\mathbf{c})/u'(\mathbf{c})$  be the Arrow-Pratt coefficient of absolute risk aversion, at  $\mathbf{c} \in \mathcal{X}$ .

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## Smooth Preferences

**Proposition 7.** Consider a function  $\mathcal{V} \in \mathcal{D}$  twice differentiable, an initial situation  $\mathbf{x} \in \mathcal{X}^n$  and a small risk  $\mathbf{z} \in \mathcal{X}^n$ . We obtain the following approximation for the utility premium:

$$\mathcal{J}(\mathbf{z}; \mathbf{x}) \approx (\nabla_{\mathcal{V}}(\mathbf{x}) \cdot \mathbf{1}) \left[ \mathbb{E}_{\pi(\mathbf{x})}(\mathbf{z}) - \frac{1}{2} \mathbf{z} \cdot \Sigma_{\pi(\mathbf{x})} \cdot \mathbf{z} \right],$$

where  $\Sigma_{\pi(\mathbf{x})} = -\nabla_{\mathcal{V}}^2(\mathbf{x}) / (\nabla_{\mathcal{V}}(\mathbf{x}) \cdot \mathbf{1})$ , which depends on the Hessian matrix  $\nabla_{\mathcal{V}}^2$  of  $\mathcal{V}$ .



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# Utility Premium & Divergence Measure

**Definition 2.** Consider a function  $\phi \in \mathcal{D}$  and two situations  $\mathbf{x}, \mathbf{y} \in \mathcal{X}^n$ . A Bregman divergence  $D_\phi : \mathcal{X}^n \times \mathcal{X}^n \longrightarrow \mathbb{R}$  is defined by:

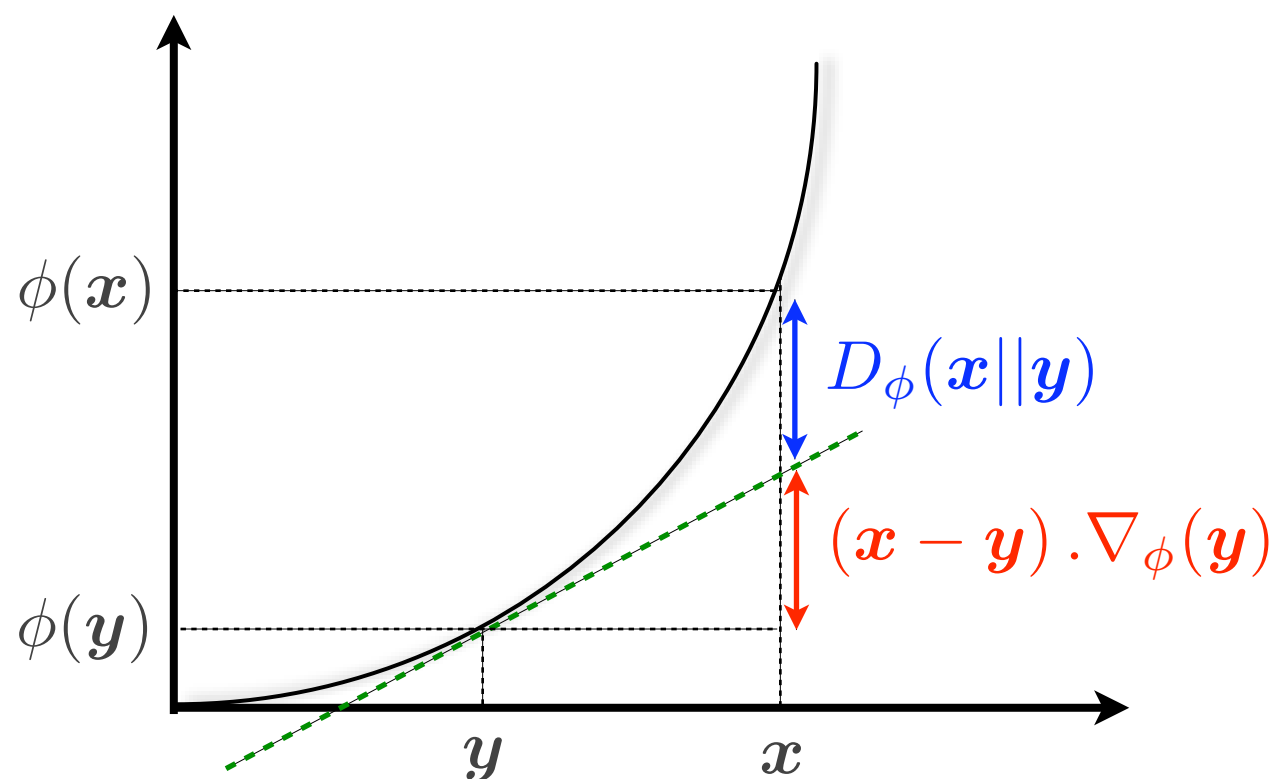
$$D_\phi(\mathbf{x} \parallel \mathbf{y}) = \phi(\mathbf{x}) - \phi(\mathbf{y}) - (\mathbf{x} - \mathbf{y}) \cdot \nabla_\phi(\mathbf{y}).$$

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## Geometric Interpretation

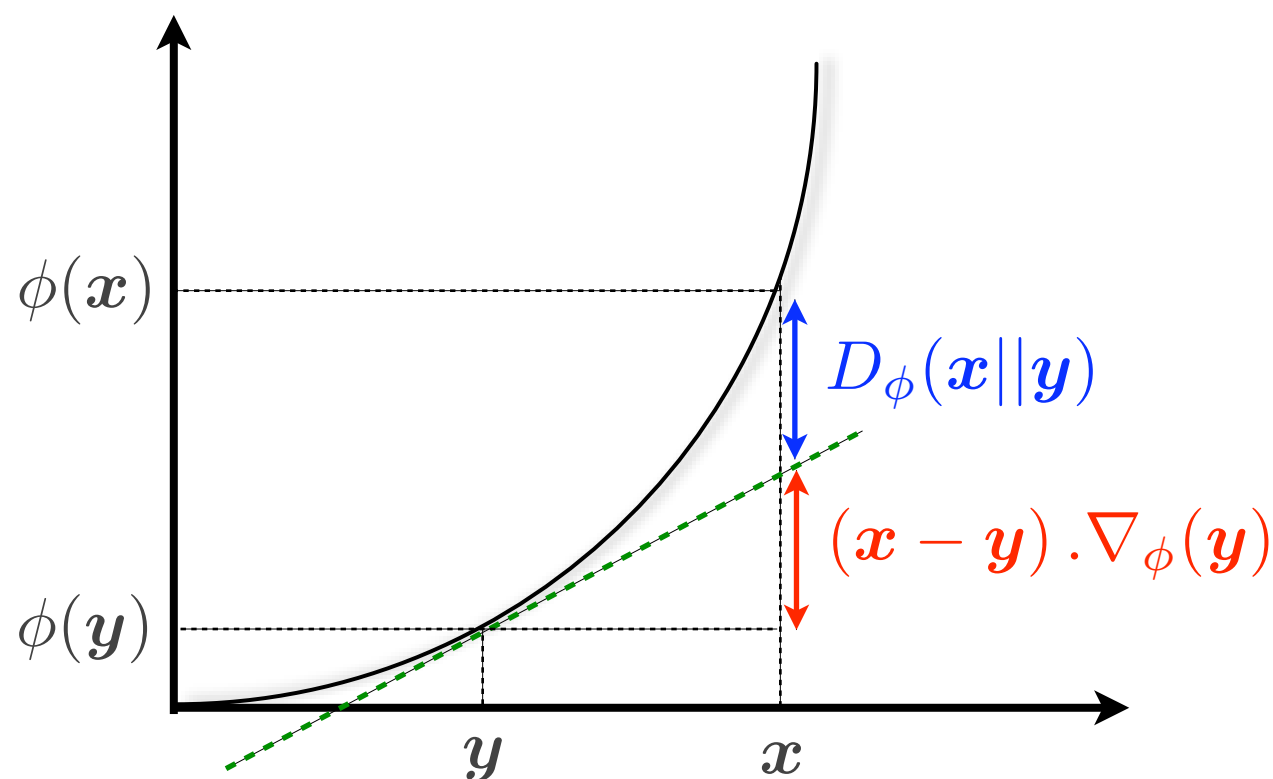


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Geometric Interpretation



(Differential) Calculus Interpretation

$$\phi(\mathbf{x}) = \phi(\mathbf{y}) + (\mathbf{x} - \mathbf{y}) \cdot \nabla_\phi(\mathbf{y}) + D_\phi(\mathbf{x}||\mathbf{y})$$

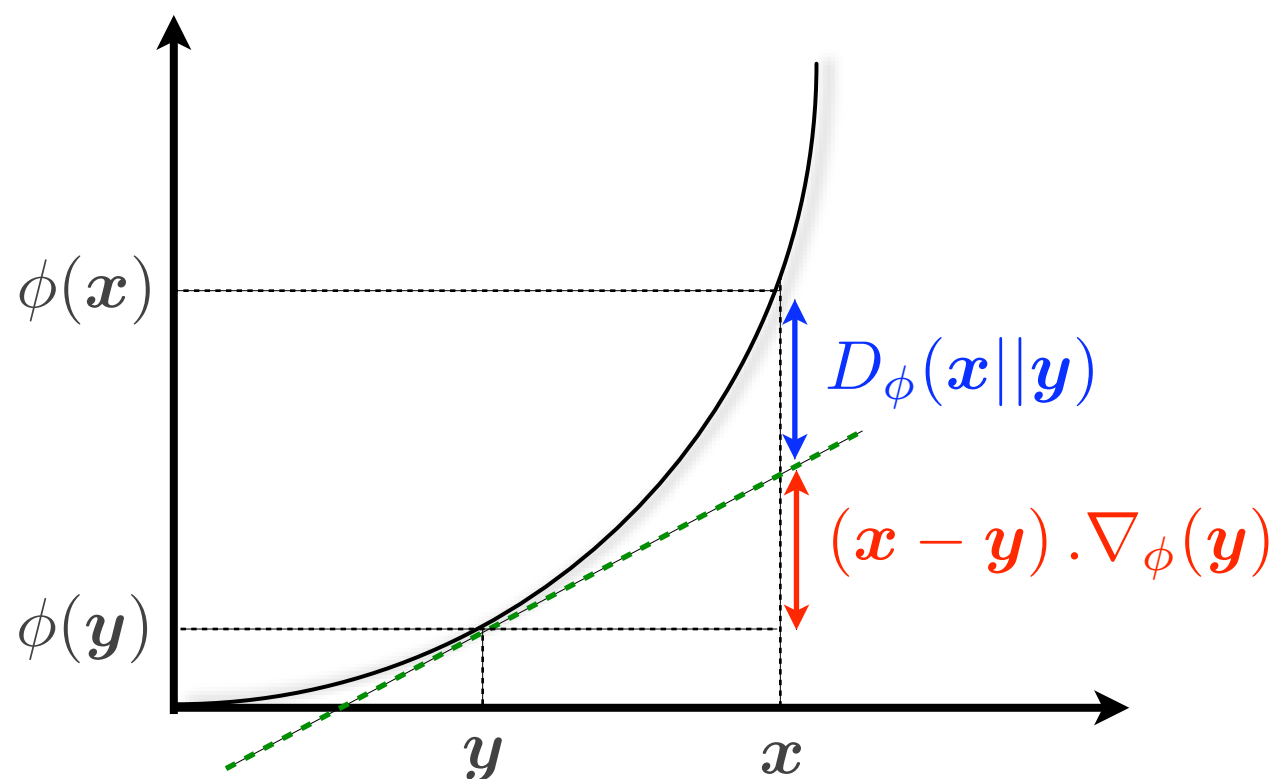


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The rest of a Taylor's series expansion

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$$D_\phi(\mathbf{x} \parallel \mathbf{y}) = \phi(\mathbf{x}) - \phi(\mathbf{y}) - (\mathbf{x} - \mathbf{y}) \cdot \nabla \phi(\mathbf{y}) .$$

**Proposition 8.** Consider a function  $\mathcal{V} \in \mathcal{D}$ , an initial situation  $\mathbf{x} \in \mathcal{X}^n$  and a risk  $\mathbf{z} \in \mathcal{X}^n$ . The utility premium can be rewritten as:

$$\mathcal{J}(\mathbf{z}; \mathbf{x}) = (\nabla \mathcal{V}(\mathbf{x}) \cdot \mathbf{1}) \left[ \mathbb{E}_{\pi(\mathbf{x})}(\mathbf{z}) - D_{\phi_{\mathbf{x}}}(\mathbf{x} + \mathbf{z} \parallel \mathbf{x}) \right] .$$

where  $\phi_{\mathbf{x}}(\mathbf{y}) = -\mathcal{V}(\mathbf{y}) / (\nabla \mathcal{V}(\mathbf{x}) \cdot \mathbf{1})$ . It is worthwhile observing that, for small risks  $\mathbf{z} \in \mathcal{X}^n$ ,  $D_{\phi_{\mathbf{x}}}(\mathbf{x} + \mathbf{z} \parallel \mathbf{x})$  corresponds to the risk premium.

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## Illustration: EU with Zero-Mean Risk

$$\max_{\mathbf{z} \in \mathcal{X}^n} \mathcal{J}_{\text{EU}}(\mathbf{z}; \mathbf{x}) \quad \Longleftrightarrow \quad \min_{\mathbf{z} \in \mathcal{X}^n} D_{\phi_{\mathbf{x}}}(\mathbf{x} + \mathbf{z} \| \mathbf{x})$$

## Illustration: EU with Zero-Mean Risk

$$\max_{\mathbf{z} \in \mathcal{X}^n} \mathcal{J}_{\text{EU}}(\mathbf{z}; \mathbf{x}) \iff \min_{\mathbf{z} \in \mathcal{X}^n} D_{\phi_{\mathbf{x}}}(\mathbf{x} + \mathbf{z} \| \mathbf{x})$$

Utility function	Related Divergence
Quadratic	Squared Euclidean distance
Logarithmic	Itakura-Saito divergence
Power (CRRA)	Bregman-Csiszar divergence
Exponential (CARA)	(see Magdalou & Nock, JET 2011)

## Illustration: EU with Zero-Mean Risk

$$\max_{z \in \mathcal{X}^n} \mathcal{J}_{\text{EU}}(z; \mathbf{x}) \iff \min_{z \in \mathcal{X}^n} D_{\phi_{\mathbf{x}}}(\mathbf{x} + z \| \mathbf{x})$$

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If riskless initial situation :

**Variance**

**2nd Theil measure**



# Applications

- Generalizing Markowitz:
  - Mean-Divergence / *non-Normal distributions* (Briys & al., 2010)
- Generalizing Tobin's asset allocation:
  - Optimization with *cumulant generating functions* (Martin, 2012)
- ... and applications in other economic disciplines:
  - Measure of *goodness-of-fit* (Cowell, Davidson & Flachaire, 2011),
  - Measure of *distance between fair and unfair income distributions* (Magdalou & Nock, 2011) ... etc.

Thanks for your attention ...

Many thanks to Mr Louis !



## More Results

**Proposition 1.** Consider a differentiable function  $u : \mathcal{X} \longrightarrow \mathbb{R}$ , an initial riskless situation  $\mathbf{c} \in \mathcal{X}^n$  and a small risk  $\mathbf{z} \in \mathcal{X}^n$ . We obtain the following approximation:

$$\mathcal{J}_{\text{EU}}(\mathbf{z}; \mathbf{c}) \approx u'(\mathbf{c}) [\mathbb{E}(\mathbf{z}) - \mathcal{R}_{\text{EU}}(\mathbf{z}; \mathbf{c})] .$$

**Proposition 2.** Consider a twice differentiable function  $u : \mathcal{X} \longrightarrow \mathbb{R}$ , an initial riskless situation  $\mathbf{c} \in \mathcal{X}^n$  and a small risk  $\mathbf{z} \in \mathcal{X}^n$ . We obtain the following approximation for the utility premium:

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where  $a(\mathbf{c}) = -u''(\mathbf{c})/u'(\mathbf{c})$  be the Arrow-Pratt coefficient of absolute risk aversion, at  $\mathbf{c} \in \mathcal{X}$ .

## More Results

**Proposition 3.** Consider a differentiable function  $u : \mathcal{X} \longrightarrow \mathbb{R}$ , an initial riskless situation  $\mathbf{c} \in \mathcal{X}^n$  and a risk  $\mathbf{z} \in \mathcal{X}^n$ . The utility premium can be rewritten as:

$$\mathcal{J}_{\text{EU}}(\mathbf{z}; \mathbf{c}) = u'(\mathbf{c}) [\mathbb{E}(\mathbf{z}) - D_{\phi_{\mathbf{c}}}(\mathbf{c} + \mathbf{z} \| \mathbf{c})] ,$$

where  $\phi_{\mathbf{c}}(\mathbf{y}) = -\mathcal{V}_{\text{EU}}(\mathbf{y})/u'(\mathbf{c})$ . It is worthwhile observing that, for small risks  $\mathbf{z} \in \mathcal{X}^n$ , the second term in (3) within bracket is the risk premium.

**Proposition 4.** Consider a function  $\mathcal{V} \in \mathcal{D}$ , an initial situation  $\mathbf{x} \in \mathcal{X}^n$  and a risk  $\mathbf{z} \in \mathcal{X}^n$ . The utility premium can be rewritten as:

$$\mathcal{J}(\mathbf{z}; \mathbf{x}) = \int_0^1 [\mathbf{z} \cdot \nabla_{\mathcal{V}}(\mathbf{x} + \lambda \mathbf{z})] d\lambda .$$

## More Results

**Proposition 5.** Consider a function  $\mathcal{V} \in \mathcal{D}$ , an initial situation  $\mathbf{x} \in \mathcal{X}^n$  and a risk  $\mathbf{z} \in \mathcal{X}^n$ . The utility premium can be rewritten as:

$$\mathcal{J}(\mathbf{z}; \mathbf{x}) = \int_0^1 \alpha(\mathbf{x} + \lambda \mathbf{z}) P(\mathbf{z}; \mathbf{x} + \lambda \mathbf{z}) d\lambda,$$

where  $\alpha(\mathbf{y}) = (\nabla_{\mathcal{V}}(\mathbf{y}) \cdot \mathbf{1}) \geq 0$  by virtue of monotonicity of  $\mathcal{V}$ .

**Proposition 6.** Consider a function  $\mathcal{V} \in \mathcal{D}$  twice differentiable, an initial situation  $\mathbf{x} \in \mathcal{X}^n$  and a small risk  $\mathbf{z} \in \mathcal{X}^n$ . We obtain the following approximation:

$$\mathcal{J}(\mathbf{z}; \mathbf{x}) \approx (\nabla_{\mathcal{V}}(\mathbf{x}) \cdot \mathbf{1}) \left[ \mathbb{E}_{\pi(\mathbf{x})}(\mathbf{z}) - \mathcal{R}(\mathbf{z}; \mathbf{x}) \right].$$



## More Results

**Proposition 7.** Consider a function  $\mathcal{V} \in \mathcal{D}$  twice differentiable, an initial situation  $\mathbf{x} \in \mathcal{X}^n$  and a small risk  $\mathbf{z} \in \mathcal{X}^n$ . We obtain the following approximation for the utility premium:

$$\mathcal{J}(\mathbf{z}; \mathbf{x}) \approx (\nabla_{\mathcal{V}}(\mathbf{x}) \cdot \mathbf{1}) \left[ \mathbb{E}_{\pi(\mathbf{x})}(\mathbf{z}) - \frac{1}{2} \mathbf{z} \cdot \Sigma_{\pi(\mathbf{x})} \cdot \mathbf{z} \right],$$

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**Proposition 8.** Consider a function  $\mathcal{V} \in \mathcal{D}$ , an initial situation  $\mathbf{x} \in \mathcal{X}^n$  and a risk  $\mathbf{z} \in \mathcal{X}^n$ . The utility premium can be rewritten as:

$$\mathcal{J}(\mathbf{z}; \mathbf{x}) = (\nabla_{\mathcal{V}}(\mathbf{x}) \cdot \mathbf{1}) \left[ \mathbb{E}_{\pi(\mathbf{x})}(\mathbf{z}) - D_{\phi_{\mathbf{x}}}(\mathbf{x} + \mathbf{z} \| \mathbf{x}) \right].$$

where  $\phi_{\mathbf{x}}(\mathbf{y}) = -\mathcal{V}(\mathbf{y}) / (\nabla_{\mathcal{V}}(\mathbf{x}) \cdot \mathbf{1})$ . It is worthwhile observing that, for small risks  $\mathbf{z} \in \mathcal{X}^n$ ,  $D_{\phi_{\mathbf{x}}}(\mathbf{x} + \mathbf{z} \| \mathbf{x})$  corresponds to the risk premium.

## More Results

**Proposition 9.** *Consider a function  $\mathcal{V} \in \mathcal{D}$  twice differentiable, an initial situation  $\mathbf{x} \in \mathcal{X}^n$ , a risk  $\mathbf{z} \in \mathcal{X}^n$  and a function  $\phi_{\mathbf{x}}$  as defined in Proposition 8. We have:*

$$D_{\phi_{\mathbf{x}}}(\mathbf{x} + \mathbf{z} \| \mathbf{x}) = \int_0^1 (1 - \lambda) [\mathbf{z} \cdot \Sigma_{\pi(\mathbf{x})}(\lambda) \cdot \mathbf{z}] d\lambda.$$

where  $\Sigma_{\pi(\mathbf{x})}(\lambda) = -\nabla_{\mathcal{V}}^2(\mathbf{x} + \lambda \mathbf{z}) / (\nabla_{\mathcal{V}}(\mathbf{x}) \cdot \mathbf{1})$ . For small risks  $\mathbf{z} \in \mathcal{X}^n$ , we have  $\Sigma_{\pi(\mathbf{x})}(\lambda) \approx \Sigma_{\pi(\mathbf{x})}$ , as defined in Proposition 7, and  $D_{\phi_{\mathbf{x}}}(\mathbf{x} + \mathbf{z} \| \mathbf{x}) \approx \frac{1}{2} \mathbf{z} \cdot \Sigma_{\pi(\mathbf{x})} \cdot \mathbf{z}$ .