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Adverse Selection and Loss Coverage in insurance markets

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1 Introduction

- 2) Why do people buy insurance?
- 3 What drives demand for insurance?
- 4 How much of population losses is compensated by insurance?
- 5 Which regime is most beneficial to society?

6) Conclusions

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Background

Background

Adverse selection:

If insurers cannot charge **risk-differentiated** premiums, then:

- higher risks buy more insurance, lower risks buy less insurance,
- raising the **pooled** price of insurance,
- lowering the demand for insurance,

usually portrayed as a bad outcome, both for **insurers** and for **society**.

In practice:

Policymakers often see merit in restricting insurance risk classification

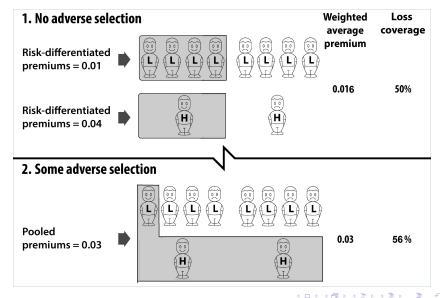
- EU ban on using gender in insurance underwriting.
- Moratoria on the use of genetic test results in underwriting.

Question:

How can we reconcile theory with practice?

Background

Motivating example



Agenda

Agenda

We ask:

- Why do people buy insurance?
- What drives demand for insurance?
- How much of population losses is compensated by insurance (with and without risk classification)?
- Which regime is most beneficial to society?

We find:

Social welfare is maximised by maximising loss coverage.

Definition (Loss coverage)

Expected population losses compensated by insurance.

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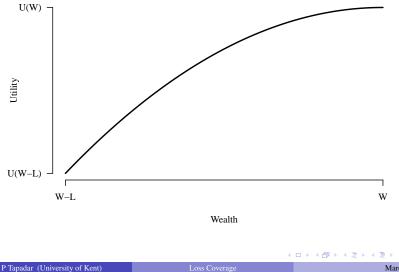
Why do people buy insurance?

Assumptions

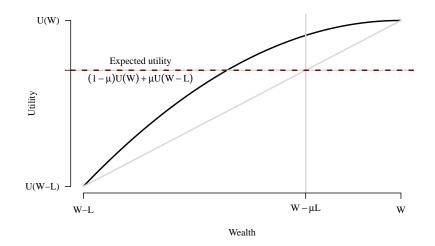
Consider an individual with

- an initial wealth W,
- exposed to the risk of loss *L*,
- with probability μ ,
- utility of wealth U(w), with U'(w) > 0 and U''(w) < 0,
- an opportunity to insure at premium rate π .

Utility of wealth

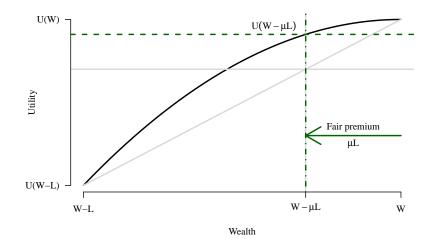


Expected utility: Without insurance



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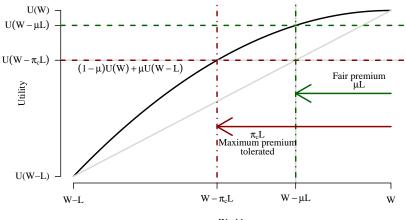
Expected utility: Insured at fair actuarial premium



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Maximum premium tolerated: π_c



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Modelling demand for insurance

Simplest model:

If everybody has exactly the same W, L, μ and $U(\cdot)$, then:

- All will buy insurance if $\pi < \pi_c$.
- None will buy insurance if $\pi > \pi_c$.

Reality: Not all will buy insurance even at fair premium. Why?

Heterogeneity:

- Individuals are **homogeneous** in terms of underlying risk.
- But they can be heterogeneous in terms of risk-aversion.

Source of Randomness:

An individual's utility function: $U_{\gamma}(w)$, where parameter γ is drawn from random variable Γ with distribution function $F_{\Gamma}(\gamma)$.

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Demand is a survival function

Standardisation

As certainty equivalent is invariant to positive affine transformations, we assume $U_{\gamma}(W) = 1$ and $U_{\gamma}(W - L) = 0$ for all γ .

Condition for buying insurance:

Given a premium π , an individual will buy insurance if:

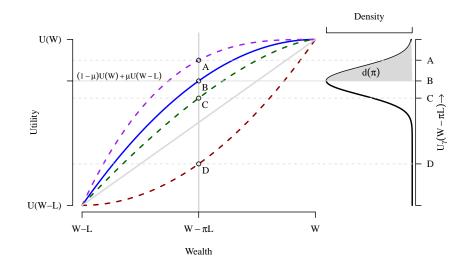
$$\underbrace{U_{\gamma}(W - \pi L)}_{\text{With insurance}} > \underbrace{(1 - \mu) U_{\gamma}(W) + \mu U_{\gamma}(W - L) = (1 - \mu)}_{\text{Without insurance}}.$$

Demand as a survival function:

Given a premium π , insurance demand, $d(\pi)$, is the survival function:

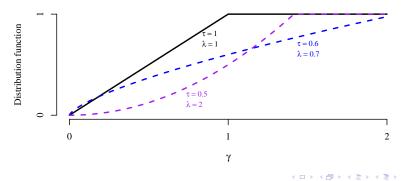
$$d(\pi) = \mathbf{P}\left[U_{\Gamma}\left(W - \pi L\right) > 1 - \mu\right].$$

Demand is a survival function



Illustrative example: W = L = 1 and $U_{\gamma}(w) = w^{\gamma}$:

$$F_{\Gamma}(\gamma) = \mathbf{P}\left[\Gamma \leq \gamma\right] = \begin{cases} 0 & \text{if } \gamma < 0\\ \tau \gamma^{\lambda} & \text{if } 0 \leq \gamma \leq (1/\tau)^{1/\lambda}\\ 1 & \text{if } \gamma > (1/\tau)^{1/\lambda}, \end{cases}$$

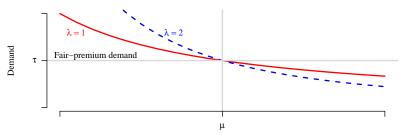


Illustrative example: W = L = 1 and $U_{\gamma}(w) = w^{\gamma}$:

$$d(\pi) = \mathbf{P}\left[U_{\Gamma}\left(W - \pi L\right) > 1 - \mu\right] \approx \tau \left(\frac{\mu}{\pi}\right)^{\lambda}$$

$$\Rightarrow \epsilon(\pi) = \left| \frac{\frac{\partial u(\pi)}{\partial (\pi)}}{\frac{\partial \pi}{\pi}} \right| = \lambda \quad \text{(constant elasticity} \Rightarrow \text{Iso-elastic demand).}$$

Iso-elastic demand for insurance



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Risk classification

Assume all have same W = L = 1 and constant demand elasticity λ .

Risk-groups

Suppose the population can be divided into 2 risk-groups, with:

- risk of losses: $\mu_1 < \mu_2$;
- population proportions: *p*₁ and *p*₂;
- fair premium demand: $d_1(\mu_1) = \tau_1$ and $d_2(\mu_2) = \tau_2$, i.e.

$$d_i(\pi) = \tau_i \left(\frac{\pi}{\mu_i}\right)^{-\lambda}, \quad i = 1, 2;$$

• premiums offered: π_1 and π_2 (note that $\pi_1 = \pi_2$ is allowed).

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Equilibrium

For a randomly chosen individual:

Define random variables:

Q = I [Individual is insured];

- X = I [Individual incurs a loss];
- $\Pi = \text{Premium offered to the individual.}$

Equilibrium

Expected Premium:

Expected Claim:

Equilibrium:

$$E[Q\Pi] = \sum_{i} p_i d_i(\pi_i) \pi_i.$$
$$E[QX] = \sum_{i} p_i d_i(\pi_i) \mu_i.$$
$$E[Q\Pi] = E[QX].$$

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Case 1: Risk-differentiated premium

Observations:

If risk-differentiated premiums are allowed,

- One possible equilibrium is achieved when $\pi_i = \mu_i$.
- No losses for insurers.
- No (actuarial/economic) adverse selection.

Loss coverage (Population losses compensated by insurance): Loss coverage: $E[QX] = \sum_i p_i d_i(\mu_i) \mu_i = \sum_i p_i \tau_i \mu_i$.

Case 2: Pooled premium

Equilibrium:

If only a pooled premium $\pi_1 = \pi_2 = \pi_0$ is allowed,

$$E[Q\Pi] = \sum_{i} p_i d_i(\pi_0) \pi_0;$$

$$E[QX] = \sum_{i} p_i d_i(\pi_0) \mu_i;$$

$$E[Q\Pi] = E[QX] \Rightarrow \pi_0 = \frac{\alpha_1 \mu_1^{\lambda+1} + \alpha_2 \mu_2^{\lambda+1}}{\alpha_1 \mu_1^{\lambda} + \alpha_2 \mu_2^{\lambda}}, \text{ where } \alpha_i = \frac{\tau_i p_i}{\tau_1 p_1 + \tau_2 p_2}.$$

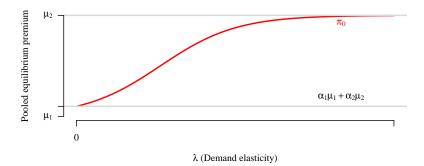
Observation:

No losses for insurers! \Rightarrow No (actuarial) adverse selection.

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Case 2: Pooled premium

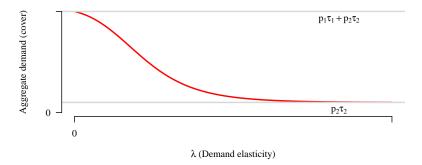


Observation:

Pooled equilibrium is greater than average premium charged under full risk classification: $\pi_0 > \alpha_1 \mu_1 + \alpha_2 \mu_2 \Rightarrow$ (Economic) adverse selection.

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Case 2: Pooled premium



Observation:

Aggregate demand (cover) is lower than under full risk classification \Rightarrow (Economic) adverse selection.

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Loss coverage ratio

Loss coverage under pooled premium:

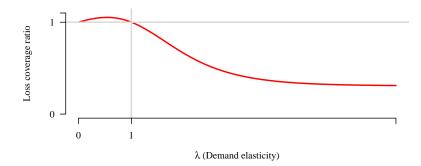
Loss coverage:
$$E[QX] = \sum_i p_i d_i(\pi_0) \mu_i$$
.

Loss coverage ratio:

$$C = \frac{\text{Loss coverage for pooled premium}}{\text{Loss coverage for risk-differentiated premium}},$$
$$= \frac{\sum_{i} p_{i} d_{i}(\pi_{0}) \mu_{i}}{\sum_{i} p_{i} d_{i}(\mu_{i}) \mu_{i}},$$
$$= \frac{1}{\pi_{0}^{\lambda}} \frac{\alpha_{1} \mu_{1}^{\lambda+1} + \alpha_{2} \mu_{2}^{\lambda+1}}{\alpha_{1} \mu_{1} + \alpha_{2} \mu_{2}}.$$

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Loss coverage ratio



- $\lambda < 1 \Rightarrow$ Pooled premium \succ Full risk classification.
- $\lambda > 1 \Rightarrow$ Pooled premium \prec Full risk classification.
- Empirical evidence suggests $\lambda < 1$ in many insurance markets.

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Social welfare

S

Definition (Social welfare)

Social welfare, S, is the sum of all individuals' expected (standardised) utilities:

$$\mathbf{E} = \mathbf{E} \left[Q U_{\Gamma}(W - \Pi L) \right] + (1 - Q) \left[(1 - X) U(W) + X U(W - L) \right],$$

$$= \sum_{i} \left[\underbrace{d_{i}(\pi_{i})U_{i}^{*}(W - \pi_{i}L)}_{\text{Insured population}} + \underbrace{(1 - d_{i}(\pi_{i}))\left\{ (1 - \mu_{i})U(W) + \mu_{i}U(W - L) \right\}}_{\text{Uninsured population}} \right] p_{i},$$

where $U_i^*(W - \pi_i L)$ is the expected utility of *i*-th risk-group's insured population.

Linking social welfare to loss coverage

Assuming $L\pi_i \approx 0$, so that $U(W) = U_i^*(W - \pi_i L)$, gives:

$$S = \text{Positive multiplier} \times \underbrace{\sum_{i} p_i d_i(\pi_i) \mu_i}_{\text{Loss Coverage}} + \text{Constant.}$$

Loss coverage provides a good proxy (which depends only on observable data) for social welfare (which depends on unobservable utilities).

Result: Maximising loss coverage maximises social welfare.

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Conclusions

Adverse selection need not always be adverse.

Restricting risk classification increases loss coverage if $\lambda < 1$.

Loss coverage is an observable proxy for social welfare.

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