(Multistage) risk-averse electricity capacity expansion

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Application: Planning for a net-zero carbon economy

Recommendation 1

Emissions budget levels

We recommend the Government set and meet the emissions budgets as outlined in the table below. These emissions budgets are expressed using GWP_{100} values from the IPCC's *Fifth Assessment Report (AR5)* for consistency with international obligations relating to Inventory reporting.

	2019	Emissions budget 1 (2022 - 2025)	Emissions budget 2 (2026 - 2030)	Emissions budget 3 (2031 - 2035)
All gases, net (AR5)		290 MtCO ₂ e	312 MtCO ₂ e	253 MtCO ₂ e
Annual average	78.0 MtCO ₂ e	72.4 MtCO ₂ e/yr	62.4 MtCO ₂ e/yr	50.6 MtCO ₂ e/yr

New Zealand CO₂ emission budgets (NZCCC May 31, 2021).

Electricity investment in renewable energy

- Many countries want to grow renewable electricity capacity.
- NZ: Climate Change Commission (CCC) sets carbon budgets to reach a net zero status by 2050.
- This will translate to emission prices for electricity generators.
- Principal-agent model: government policy determines emission price and generators invest in (mainly renewable) technology (see e.g. Quiroga, Sauma, and Pozo, 2019).
- Sesearch question: what happens when there is uncertainty and investors are risk averse?

Principal agent model in perfect, complete markets [Ralph &Smeers 2015, Ehrenman et al 2011, P., Ferris & Wets, 2016]

Theorem

(Risked equilibrium) If markets are competitive, convex and complete, and agents optimize using similar coherent risk measures, then partial equilibrium of the electricity market investment game is the same as the solution to a risk averse stochastic optimization problem (social planning problem).

How to deal with an incomplete market for risk.

Summary



- 2 Risk-averse social planning problem
- A principal-agent model
- 4 Solving the principal agent problem





Summary

Introduction

2 Risk-averse social planning problem

A principal-agent model

4 Solving the principal agent problem

6 Results

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Two-stage social planning problem [Ferris & P., 2021]

- Social planner *s* minimizes a coherent risk measure ρ_s .
- Example: convex combination of expectation \mathbb{E} and worst case \mathbb{W} of loss distribution Z_s :

$$\rho_s(Z) = (1 - \sigma) \mathbb{E}[Z_s] + \sigma \mathbb{W}[Z_s]$$

• When distribution finite, express as optimal value function.

$$\rho_{s}(Z) = \min \quad \sigma \theta_{s} + (1 - \sigma) \sum_{\omega} \mathbb{P}(\omega) Z_{s}(\omega)$$

s.t. $\theta_{s} \ge Z_{s}(\omega)$

Social planning problem example [Kok, P., Zakeri, 2018]

SP: min
$$\sum_{a} \kappa_{a} x_{a} + \rho_{s} (\sum_{t \in T} Z_{s}(\omega, t))$$

s.t.
$$Z_s(\omega, t) = \sum_a (C_a + e_a \tau) y_a(\omega, t)$$

 $+ \sum_b v_b q_b(\omega, t) - r_b(d_b(\omega, t) - q_b(\omega, t))$

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$$\begin{array}{rcl} z_a & \leq & u_a + x_a, \\ y_a(\omega, t) & \leq & \mu_a(\omega, t) z_a, \end{array}$$
$$\begin{array}{rcl} q_b(\omega, t) & \leq & d_b(\omega, t), \\ \sum_b d_b(\omega, t) & \leq & \sum_a y_a(\omega, t) + \sum_b q_b(\omega, t), \\ x, z, y, q & \geq & 0. \end{array}$$

Summary

Introduction

- 2 Risk-averse social planning problem
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5 Results

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Generator agent (a)

Generators minimize capital cost plus risk-adjusted losses

 $\begin{array}{lll} \min & \kappa_a x_a + \rho_a (\sum_{t \in T} Z_a(\omega, t)), \\ s.t. & Z_a(\omega, t) &= (C_a + \tau e_a - \pi(\omega, t)) y_a(\omega, t) \\ & z_a &\leq u_a + x_a, \quad [v_a] \\ & y_a(\omega, t) &\leq \mu_a(\omega, t) z_a, \quad [\sigma_a(\omega, t)] \\ & x_a, z, y &\geq 0. \end{array}$

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Risk measure is combination of expectation and worst case

$$Z_{a}(\omega, t) = (C_{a} + \tau e_{a} - \pi(\omega, t))y_{a}(\omega, t)$$

min
$$\kappa_a x_a + \alpha \theta_a + (1 - \alpha) \sum_{\omega} \mathbb{P}(\omega) \sum_{t \in \mathcal{T}} Z_a(\omega, t))$$

s.t.
$$\theta_a \geq \sum_{t \in \mathcal{T}} Z_a(\omega, t), \quad [\alpha \lambda_a(\omega)]$$

$$z_a \le u_a + x_a, \qquad [v_a]$$

$$y_a(\omega, t) \le \mu_a(\omega, t) z_a, \qquad [\sigma_a(\omega, t)]$$

 $x, z, y \geq 0.$

KKT(a): Optimality conditions for generator a

1 = $\sum \lambda_a(\omega)$ θ_{a} $\lambda_a(\omega) \geq 0$ $\theta_a - Z_a(\omega)$ 0 < \bot $v_a > 0$ 0 < $u_a + x_a - z_a$ $\perp \sigma_a(\omega, t) \geq 0$ $\mu_{a}(\omega,t)z_{a}-y_{a}(\omega,t)$ 0 < 0 < $x_a \ge 0$ $\kappa_a - \nu_a$ $\nu_a - \sum_{\omega} \sum_{t \in \mathcal{T}} \mu_a(\omega, t) \sigma_a(\omega, t)$ $z_a \geq 0$ 0 < \bot $0 \leq \alpha \lambda_a(\omega) (C_a + \tau e_a - \pi(\omega, t)) + \sigma_a(\omega, t)$ $+(1-\alpha)\mathbb{P}(\omega)(C_a+\epsilon\eta_a-\pi(\omega,t))$ $\perp y_a(\omega,t) \ge 0$

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Buyer agent (b)

Electricity buyer (retailer) b minimizes risk-adjusted net cost of customer supply.

$$\begin{split} \min & \rho_b(\sum_{t \in \mathcal{T}} Z_b(\omega, t)), \\ s.t. & Z_b(\omega, t) &= (r_b + v_b - \pi(\omega, t)) \cdot q_b(\omega, t) \\ & + (\pi(\omega, t) - r_b) \cdot d_b(\omega, t), \\ q_b(\omega, t) &\geq 0. \end{split}$$

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Risk measure is combination of expectation and worst case

$$Z_b(\omega, t) = (r_b + v_b - \pi(\omega, t)) \cdot q_b(\omega, t)$$
$$+ (\pi(\omega, t) - r_b) \cdot d_b(\omega, t)$$

$$\begin{array}{ll} \min & \beta \theta_b + (1 - \beta) \sum_{\omega} \mathbb{P}(\omega) \sum_{t \in \mathcal{T}} Z_b(\omega, t) \\ s.t. & \theta_b \geq \sum_{t \in \mathcal{T}} Z_b(\omega, t), \qquad [b \lambda_b(\omega)] \\ & q_b(\omega, t) \leq d(\omega, t), \qquad [\psi_b(\omega, t)] \end{array}$$

 $q_b(\omega, t) \geq 0.$

KKT(b): Optimality conditions for buyer b

$$\begin{aligned} Z_{b}(\omega, t) &= (r_{b} + v_{b} - \pi(\omega, t)) \cdot q_{b}(\omega, t) \\ &+ (\pi(\omega, t) - r_{b}) \cdot d_{b}(\omega, t) \end{aligned}$$

$$\begin{aligned} 1 &= \sum \lambda_{b}(\omega) & \perp \theta_{b}, \\ 0 &\leq \theta_{b} - \sum_{t \in T} Z_{b}(\omega, t) & \perp \lambda_{b}(\omega) &\geq 0 \\ 0 &\leq \beta \lambda_{b}(\omega)(r_{b} + v_{b} - \pi(\omega, t)) + \psi_{b}(\omega, t) \\ &+ (1 - \beta) \mathbb{P}(\omega)(r_{b} + v_{b} - \pi(\omega, t)) & \perp q_{b}(\omega, t) &\geq 0 \\ 0 &\leq d_{b}(\omega, t) - q_{b}(\omega, t) & \perp \psi_{b}(\omega, t) &\geq 0 \end{aligned}$$

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MC: Market-clearing conditions

$0 \leq \sum_{\mathsf{a}} y_{\mathsf{a}}(\omega, t) + \sum_{\mathsf{b}} q_{\mathsf{b}}(\omega, t) - \sum_{\mathsf{b}} d_{\mathsf{b}}(\omega, t) \quad \bot \quad \pi(\omega, t) \geq 0.$

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Principal solves optimization problem

$$P: \min \sum_{a} \kappa_{a} x_{a} + \sigma \theta_{s}$$

$$+(1-\sigma) \sum_{\omega} \mathbb{P}(\omega) (\sum_{a} Z_{a}(\omega, t) + \sum_{b} Z_{b}(\omega, t))$$

$$s.t. \quad \theta_{s} \ge \sum_{t \in T} \sum_{a} Z_{a}(\omega) + \sum_{t \in T} \sum_{b} Z_{b}(\omega)$$

$$KKT(a), \quad a \in \mathcal{A},$$

$$KKT(b), \quad b \in \mathcal{B},$$

$$MC.$$

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Introduction

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- 4 Solving the principal agent problem

6 Results

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Reformulate KKT conditions using binary variables [Fortuny-Amat & McCarl, 1981]

Replace

$$0 \leq F(x, y) \perp x \geq 0$$

with
$$0 \leq x \leq Mz$$

$$0 \leq F(x, y) \leq M(1 - z)$$

$$z \in \{0, 1\}$$

Principal-agent problem is MIQP

- Because of bilinear terms from choosing worst-case probability distribution using multpliers, we get a mixed integer quadratic program.
- Solve MIQP using Gurobi.
- Advantage: Global optimality of MIQP enables search for multiple equilibria (Gerard et al 2018).
- Disadvantage: Big M constraints do not scale well.

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Case study

Example

	Wind	Coal	Gas
Initial capacity (GW)	0	0	0
Capital costs (\$/MW/h)	23.75	75.00	22.50
SRMC (\$/MWh)	0.01	42.00	70.00
CO2Emission (t/MWh)	0.00	0.99	0.40

Example

Load (GW)	NI SI		Total	Hours/day	
peak	4.1	2.2	6.3	6	
offpeak	2.4	1.7	4.1	12	
shoulder	3.1	2	5.1	6	

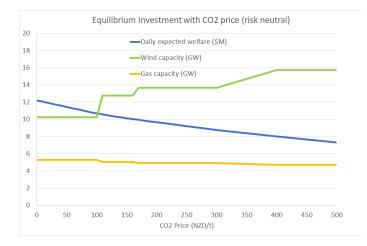
Wind load factors have three scenarios

Example

Wind load factor	Low	Medium	High
peak	0.1	0.3	0.4
offpeak	0.3	0.4	0.4
shoulder	0.2	0.3	0.4

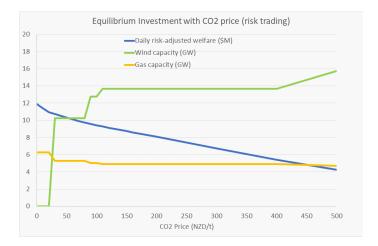
We set retail price of electricity = \$200/MWh and VOLL=\$1000/MWh for all consumers

Risk neutral social plan (sigma=0)



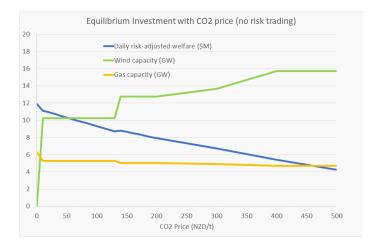
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Risk averse social plan (alpha = beta = sigma = 0.5)



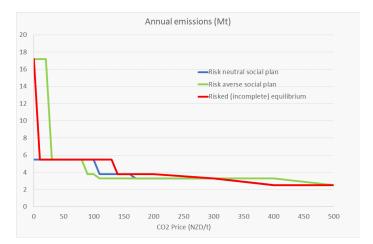
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Risk averse equilibrium (sigma = 0.5)



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Average CO2 emissions with CO2 price



In this example, CO2 emissions decrease with carbon tax but risk trading can increase emissions.

Equilibrium solutions when CO2 tax = 10/t

Solution for CO2 price = \$10/t	Wind	Coal	Gas	MaxPrice	Purchaser	System
Capacity (GJ) (incomplete)	10.250	0.000	5.275	\$614.05		
Capacity (GJ) (complete)	0.000	0.000	6.300	\$212.70		
RA daily welfare (M) (incomplete)	\$ -	\$	\$ -		\$ 1.64	\$ 11.13
RA daily welfare (M) (complete)	\$ -	\$ -	\$ -		\$ 11.42	\$ 11.42

- Incomplete case: in low-wind peak load scenario, high prices pay for wind capacity. Lack of risk trading between generators and purchaser leads to low purchaser welfare.
- Complete case: Welfare is enhanced by risk trading between gas generator and purchaser but more gas generation increases emissions from 5.5Mt p.a. to 17.2 Mt p.a.

Conclusion

- Emission budgets stimulate renewable investment through carbon prices.
- Investment might be socially suboptimal if risk averse agents cannot trade risk.
- Competitive equilibrium in incomplete risk market might achieve CO2 budget at a lower carbon price.
- Next talk: assume there is enough risk trading to deliver social optimum and plan accordingly.

Conclusion

THE END

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