Technology Neutrality vs. Policy Discrimination: Optimizing Support for Competing Green Technologies

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# Introduction

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### Context:

#### Competing Green Technologies: Hydrogen Supply and Demand

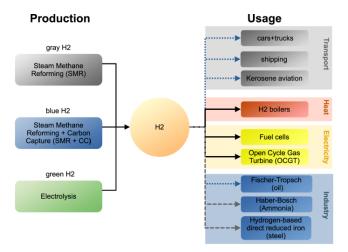


Figure: Hydrogen Production and Demand (Zeyen 2022)

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#### Motivation: Should Climate Policy be Technology-Neutral?

- **Context:** Many low-carbon technologies remain costly, immature, and face high uncertainty.
  - Particularly true in hard-to-abate sectors (e.g. industry, transport).
  - Multiple competing options may serve similar decarbonization goals.
- Focus on Carbon Contract for Differences: Forward carbon price to support clean investments, with different goals:
  - Correcting for a too-low expected EU ETS price
  - Derisking revenue streams exposed to carbon price volatility.
  - Supporting investment in less mature technologies whose risk profile translates into high capital costs.
- Core question: Should support remain technology-neutral or be technology-specific?
  - Cost structures and financing risks vary across technologies.
  - Uniform support may misallocate resources toward less risky options.
  - But targeting risks misjudging future competitiveness.

#### Our approach:

- Compare policy instruments under asymmetric information and cost heterogeneity.
- Apply to green vs. blue hydrogen competition under CCfDs.
- Study trade-offs between instruments (quantity vs. price) and targeting (neutral vs. specific).

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#### ■ Prices vs. Quantities under Uncertainty: Weitzman (1974) and extensions

- Trade-offs under asymmetric information and imperfect substitutability (Williams III 2002; Meunier 2011; Weitzman 2020)
- Our contribution: two competing abatement technologies + technology-specific cost wedges
- Technology-neutral vs. technology-specific policies:
  - Fabra and Montero (2023): discrimination justified by the cost of public funds
  - Our approach: support justified by financing barriers or externalities

#### Risk aversion and CCfDs:

- CCfDs lower investment risks and financing costs (Richstein 2017; Richstein and Neuhoff 2022; Jeddi et al. 2021)
- Role of revenue certainty vs. investment barriers (Chiappinelli and Neuhoff 2020; Chaton and Metta-Versmessen 2023)

#### Financing constraints for green technologies:

- Clean tech access to capital and maturity gaps (Hall and Lerner 2010; Polzin et al. 2021; Ang et al. 2017; Brunnschweiler 2010)
- Few papers model technology-specific support explicitly

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## Objectives and preview of the results

#### Main research questions

- Normative question: Should climate policy be technology-neutral or targeted?
- Application: What is the optimal design of CCfDs for green vs. blue hydrogen in Europe?
- Methodology: A partial equilibrium model with competing technologies under asymmetric information and risk premia

### Main findings

- Technology-neutral policies distort allocation when technologies face heterogeneous financing risks.
- Targeted subsidies (e.g. differentiated CCfDs) improve welfare by correcting these distortions.
- The gains from targeting are higher when technologies are close substitutes (strong competition).
- Combining a neutral quota with technology-specific subsidies outperforms specific and neutral quotas.
- In a calibrated model for green vs. blue H<sub>2</sub>, targeted support nearly doubles the share of blue hydrogen.

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# Main Model

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## Model Set-up: Two Competing Abatement Technologies

- **Technologies:** Two abatement options  $i \in \{1, 2\}$  with long-term capacities  $q_i$ , total abatement  $q = q_1 + q_2$
- Regulator's perspective (social planner):
  - Social cost function:

$$C(q_1, q_2) = (c_1 + \theta_1)q_1 + (c_2 + \theta_2)q_2 + \frac{\beta_1}{2}q_1^2 + \frac{\beta_2}{2}q_2^2 + \gamma q_1 q_2$$

 $\theta_i$ : cost shocks (mean zero);  $\gamma$ : substitutability between technologies

- Public benefit of abatement:

$$B(q) = \left(a - \frac{b}{2}q\right)q$$

Social welfare:

$$W(q_1, q_2, \theta_1, \theta_2) = B(q_1 + q_2) - C(q_1, q_2)$$

- Firm's perspective:
  - Perceived private cost:

$$\tilde{C}(q_1, q_2) = (c_1 + \rho_1 + \theta_1)q_1 + (c_2 + \rho_2 + \theta_2)q_2 + \frac{\beta_1}{2}q_1^2 + \frac{\beta_2}{2}q_2^2 + \gamma q_1 q_2$$

 $\rho_i$ : cost premiums reflecting risk aversion

– Profit maximization under prices (p<sub>1</sub>, p<sub>2</sub>):

$$\Pi = p_1 q_1 + p_2 q_2 - \tilde{C}(q_1, q_2)$$

## Model Set-up: Instruments and Timing

- Objective: compare four realistic policy instruments:
  - Price-based: one price p (neutral), or two prices p1, p2 (specific)
  - Quantity-based: one quota Q (neutral), or two quotas Q1, Q2 (specific)
- Timing:
  - Stage 1: regulator sets instrument level before knowing cost shocks  $\theta_i$
  - Stage 2: firms observe  $\theta_i$ , choose  $q_i$  accordingly
- Welfare notation:

$$W_P^N, W_P^S, W_Q^N, W_Q^S$$

where P = price, Q = quota; N = neutral, S = specific

The expected values of quantities are denoted  $\bar{q}_i$ .

The optimal allocation  $q_1^*$ ,  $q_2^*$  is the allocation that maximizes the social Welfare.

### Quotas: Technology-neutral vs. Technology-specific

#### ■ Technology-specific quotas Q<sub>1</sub>, Q<sub>2</sub>:

- Quantities fixed ex ante; unaffected by  $\theta_i$
- Optimal quotas:  $Q_i^* = \bar{q}_i^*$  (based on socially optimal allocation)
- Cost premiums  $\rho_i$  do not influence optimal allocation

#### **Technology-neutral quota** $Q = Q_1^* + Q_2^*$ :

- Market-clearing price  $\bar{p}$  ensures  $q_1 + q_2 = Q$
- Allocation distorted by risk premiums and shocks:

$$q_1 = Q_1^* - rac{
ho_1 - 
ho_2}{
ho_1 + 
ho_2 - 2\gamma} - rac{ heta_1 - heta_2}{
ho_1 + 
ho_2 - 2\gamma}$$

#### Proposition 1 – Welfare Difference

$$W_Q^N - W_Q^S = \frac{1}{2} \cdot \frac{\mathbb{E}[(\theta_1 - \theta_2)^2] - (\rho_1 - \rho_2)^2}{\beta_1 + \beta_2 - 2\gamma}$$

- A single quota allows quantities to adjust to cost shocks  $\Rightarrow$  gains from flexibility.
- But it fails to correct cost premiums ⇒ distorted allocation across technologies.
- These two effects work in opposite directions in the welfare comparison.

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## Technology-Neutral Quota with Two Subsidies

 Adding technology-specific subsidies to a single quota restores efficient allocation while preserving the adaptation to random cost shocks.

#### **Proposition 2**

A technology-neutral quota together with two subsidies  $\rho_1$  and  $\rho_2$  for quantities produced with technologies 1 and 2 outperforms two technology-specific quotas by:

$$\frac{1}{2} \frac{\mathbb{E}[(\theta_1 - \theta_2)^2]}{\beta_1 + \beta_2 - 2\gamma}$$

The gain from adding subsidies increases with  $\gamma$ , and equals:

$$\frac{1}{2} \frac{(\rho_1 - \rho_2)^2}{\beta_1 + \beta_2 - 2\gamma}$$

- Subsidies offset cost premium distortions without constraining total quantity.
- Higher substitutability (γ) magnifies both misallocation and the benefit of correcting it.

### Prices : Technology-specific vs. Technology-neutral

- With price-based instruments, the regulator sets prices ex ante based on expected firm behavior.
- Technology-specific prices internalize risk premiums:

$$p_i^* = B'(\bar{q}^*) + \rho_i$$

Uniform price averages across technologies, compounding distortions:

$$p^{*} = \frac{(\beta_{2} - \gamma)p_{1}^{*} + (\beta_{1} - \gamma)p_{2}^{*}}{\beta_{1} + \beta_{2} - 2\gamma}$$

#### Proposition 3 – Welfare Gain from Price Discrimination

$$W_{P}^{S} - W_{P}^{N} = \frac{1}{2} \cdot \frac{(\rho_{2} - \rho_{1})^{2}}{\beta_{1} + \beta_{2} - 2\gamma}$$

The gain increases with the substitutability parameter  $\gamma$ .

- Discrimination corrects misallocation due to heterogeneous risk.
- The gain from discriminating is the same as the gain obtained by introducing two subsidies into a single, technology-neutral auction

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## Neutral Price vs Quantity Instruments

- The distortion introduced by risk premiums plays no role here.
- Three key factors shape the price versus quantity comparison:
  - **Cost shocks**  $(\theta_i)$ : increases the value of flexibility.
  - Slope of marginal benefit and cost
  - Substitutability  $(\gamma)$ : determines the degree of competition between technologies.

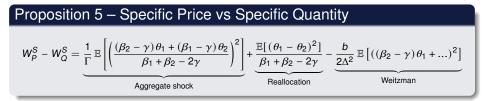
#### Proposition 4 – Neutral Price vs Neutral Quantity

$$W_P^N - W_Q^N = \frac{1}{2\Gamma^2} \cdot (\Gamma - b) \cdot \mathbb{E}\left[\left(\frac{(\beta_2 - \gamma)\theta_1 + (\beta_1 - \gamma)\theta_2}{\beta_1 + \beta_2 - 2\gamma}\right)^2\right]$$

in which  $\Gamma = \frac{\beta_1 \beta_2 - \gamma^2}{\beta_1 + \beta_2 - 2\gamma}$ 

- Γ: slope of marginal cost. b: slope of marginal benefit.
- This reflects the classic Weitzman (1974) trade-off:
  - If  $\Gamma > b$ : price-based instruments preferred.
  - If  $\Gamma < b$ : quantity-based instruments preferred.
- Welfare difference is proportional to the variance of the cost shock.

## Specific Price vs Quantity Instruments



- The comparison can be decomposed into three interpretable effects:
  - Weitzman trade-off (see proposition 4)
  - Agregate shock effect: captures the effect of the aggregate shock on the total quantity.
  - Reallocation effect: captures the gain for price instrument from reallocation between technologies, and depends on the dispersion of the individual shocks.
- Role of  $\gamma$ :
  - Higher  $\gamma$  leads to stronger reallocation gains for prices (technologies compete more).
  - But also increases aggregate volatility → ambiguous effect.

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# Application to CCfDs

#### Technology-neutral or specific support?

- Purpose of CCfDs: reduce investment risk by guaranteeing a CO<sub>2</sub> price ⇒ lower capital costs for clean tech developers (Richstein and Neuhoff 2022)
- Hybrid nature of CCfDs:
  - Theoretically price-based (guaranteed CO<sub>2</sub> price)
  - Practically quantity-driven (allocation via auctions, fixed decarbonization targets)
- Firms are risk-averse: risk-adjusted cost includes a premium :  $\rho_i = \lambda(\sigma^2 + \sigma_i^2)$  where  $\sigma^2 = carbon price risk, \sigma_i^2 = tech-specific cost risk, <math>\lambda$  risk aversion.
- **Naive CCfD effects:** Removes  $\sigma^2$ , but leaves  $\sigma_i^2$  untreated
- Policy implication:
  - Uniform CCfDs induce misallocation when  $\rho_1 \neq \rho_2$
  - Optimal CCfD should adjusts for ρ<sub>i</sub>
- Welfare loss of naive CCfD:

$$N_{P}^{N} - \hat{W} = \frac{\lambda^{2}}{2\Gamma + b} \left[ \frac{(\beta_{1} - \gamma) \sigma_{2}^{2} + (\beta_{2} - \gamma) \sigma_{1}^{2}}{\beta_{1} + \beta_{2} - 2\gamma} \right]^{2}$$

 $\Rightarrow$  Loss grows with premium differences and competition intensity ( $\gamma$ )

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# Numerical application

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# **Microfounded Model**

Analytical exploration, without asymmetric information

- A continuum of polluting sites faces a choice:
  - Adopt technology 1 or 2 (e.g., green or blue H<sub>2</sub>)
  - Or remain inactive (if both net gains are negative)
- Sites differ by **cost pairs**  $(c_1, c_2)$  drawn from joint distribution  $f(c_1, c_2)$
- With technology-specific prices p<sub>1</sub>, p<sub>2</sub> and cost premiums ρ<sub>1</sub>, ρ<sub>2</sub>:
  - Tech 1 chosen if  $p_1 (c_1 + \rho_1)$  dominates
  - Tech 2 chosen if  $p_2 (c_2 + \rho_2)$  dominates
- Resulting total welfare (without asymmetric information):

$$W = B(q_1 + q_2) - \int_{\mathcal{D}_1} c_1 f(c_1, c_2) dc_1 dc_2 - \int_{\mathcal{D}_2} c_2 f(c_1, c_2) dc_1 dc_2.$$
(1)

- The intensity of competition depends on how many firms are nearly indifferent i.e., when  $c_1 \approx c_2$ .
- The more firms are close to this frontier, the more responsive the technology mix is to small price changes (i.e  $\frac{\partial q_1}{\partial p_2}$ ). This responsiveness plays the same role as  $\gamma$  in the quadratic model.  $\Rightarrow$  A high  $\gamma$  corresponds to a dense indifference frontier in the microfounded model.

# Illustration: Misallocation from Ignoring Cost Premiums

Optimal vs. biased allocation across sites

- **Left:** Optimal allocation with decentralized prices  $p_i = B'(q^*) + \rho_i$
- Right: Biased equilibrium under uniform price without correcting for cost premiums
- Two distortions:
  - Sites that should adopt remain inactive (gray area)
  - Sites adopt the wrong technology (hatched area)

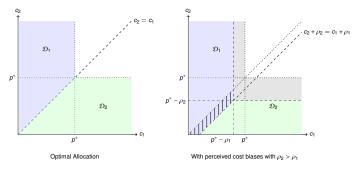


Figure: Distribution of abatement across sites at the optimum and biased equilibrium

# Calibration

From grey to low-carbon hydrogen in Europe

- Reference: Grey hydrogen via SMR without CO<sub>2</sub> capture.
- Clean alternatives:
  - Technology 1 Green H<sub>2</sub>: electrolysis with renewable electricity
  - Technology 2 Blue H<sub>2</sub>: SMR + CCS (high-capture variant, 95% )
- Data: 9.2 Mt/year of grey hydrogen production in Europe
- Decarbonization Target: 55% reduction target for industrial emissions by 2030

Parameter	Meaning	Value	Source
b <sub>1</sub> , b <sub>2</sub>	Energy use (green / blue H <sub>2</sub> )	0.05 / 0.04 MWh/kg	IEA
C1, C2	Fixed costs (€/kg)	1.75 / 1.9	EU H <sub>2</sub> Observatory
$\theta_1, \theta_2$	Cost shocks (€/kg)	$N(0, 1.1^2) / N(0, 1.3^2)$	Internal calibration
λ	Risk aversion	5	Epstein et al.
$\sigma_{K,1} / \sigma_{K,2}$	CAPEX uncertainty (€/kg)	0.14 / 0.38	OECD
$\sigma_{pCO_2}$	Carbon price volatility (€/kg)	0.29	EEA
$\sigma_{pCO_2} Q^*$	Decarbonization target	5.1 Mt H <sub>2</sub> /year	EU H <sub>2</sub> Observatory
(a, b)	Abatement benefit parameters	(8.6, 10 <sup>-6</sup> )	Internal

# Distribution of Energy Input Costs

Heatmap of hydrogen production by site-level energy prices

- Sites are positioned by their energy input costs:
  - x-axis: Estimated Levelized Cost of Electricity (LCOE) in 2030, impacting the cost of green H<sub>2</sub>
  - y-axis: Estimated natural gas price, impacting the cost of blue H<sub>2</sub>
- Each cell represents a grey H<sub>2</sub> production unit in 2030
- This cost landscape shapes spatial heterogeneity in our model

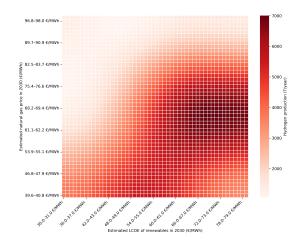


Figure: Grey  $H_2$  production density by projected energy costs (2030)

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# Main Results

#### Overview of policy outcomes

Scenario	Welfare (bn)	Price (/tCO <sub>2</sub> )	Green H <sub>2</sub> (%)	Blue H <sub>2</sub> (%)
0. CCfD naive	8.13	580	21.4	9.2
1. Price-based, techno-neutral	12.12	690	45.3	14.3
2. Price-based, techno-specific	13.24	(640, 770)	30.2	33.4
3. Quota-based, techno-neutral	12.04	680	40.2	16.1
4. Quota-based, techno-specific	13.04	(600, 770)	21.2	34.8
5. Quota-based, neutral, subsidies	13.19	(630, 750)	29.1	29.0

Table: Expected outcomes under alternative policy designs

- Naive CCfD underperforms: too little abatement, overly green mix.
- **Targeting matters:** tech-specific instruments significantly improve welfare and balance the mix.
- Best outcome: tech-specific prices (Scenario 2).
- Neutral quota + subsidies (Scenario 5) nearly as good.
- Targeting gains greater than instrument choice gains.

# Sensitivity Analysis: Spatial Distribution of Sites

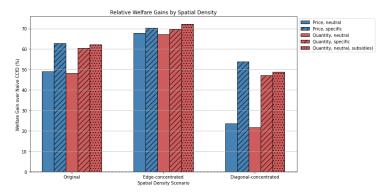


Figure: Welfare gains vs. naïve CCfD across spatial scenarios

- Higher site concentration along the diagonal  $\Rightarrow$  higher substitutability (high  $\gamma$ )  $\Rightarrow$  larger benefit from technology-specific policies
- Edge-concentrated configuration  $\Rightarrow$  lower  $\gamma \Rightarrow$  technology-neutral instruments perform relatively better

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# Sensitivity Analysis: Decarbonization Target

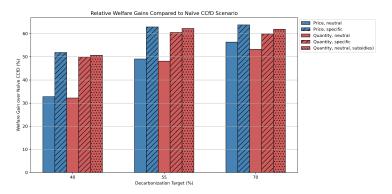


Figure: Welfare gains vs. naïve CCfD at 40%, 55%, and 70% decarbonization targets

- Higher targets ⇒ more costly to under-decarbonize ⇒ quota-based instruments gain relative importance
- At low ambition levels, technology-specific instruments bring clearer benefits, especially for quantity-based policies

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## Sensitivity Analysis: Risk Aversion

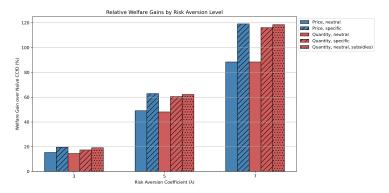


Figure: Welfare gains vs. naïve CCfD for different values of risk aversion  $\lambda$ 

- Higher  $\lambda \Rightarrow$  stronger cost premium distortions  $\Rightarrow$  greater gains from technology-specific support
- At high risk aversion, quantity-specific instruments close the gap with price instruments

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# Sensitivity Analysis: Cost of Missing Target

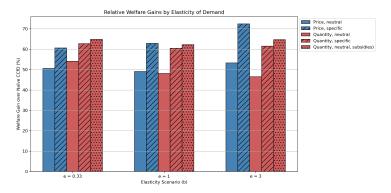


Figure: Welfare gains vs. naïve CCfD for different values of penalty parameter e

- Low e (high penalty b) ⇒ quantity-based instruments preferred: they guarantee the target is met
- High *e* (low penalty) ⇒ price-based instruments regain advantage; targeting remains robust in both cases

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# Conclusion

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## Conclusion

#### Main insights

 Policy design under uncertainty: Our model shows how asymmetric information and heterogeneous risk premiums distort technology allocation.

#### Key theoretical results:

- With quantities, neutral quotas + targeted subsidies outperform all other designs.
- With prices, technology-specific instruments always dominate uniform ones.
- The choice between price vs. quantity depends on sensitivity to cost shocks.
- Numerical illustration: Applied to green vs. blue hydrogen competition in Europe. Confirms: *Targeting is most valuable when technologies are close substitutes.*

#### Limitations and future directions

- Risk premiums assumed observable learning or monitoring could be modeled.
- Full substitutability in emissions is a strong assumption (esp. for blue H<sub>2</sub>).
- Uncertainty on true abatement potential (e.g., methane leakage) should be internalized.

# Appendix

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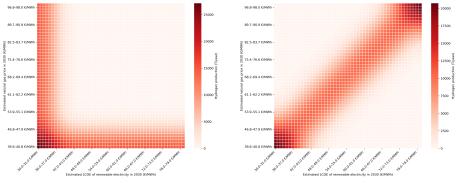
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# Appendix



(a) Edge-concentrated scenario

#### (b) Diagonal-concentrated scenario

Figure: Smoothed hydrogen production by projected energy costs in 2030 under two alternative spatial distributions of sites. Production values are diffused using a Gaussian filter.