



Aalto University  
School of Economics

# Efficiency analysis as a principle-agent game

An application to the regulation of  
electricity distribution networks

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HECER lunch seminar  
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# Plan for the presentation

- Regulation of electricity distribution in Finland: some recent developments
- Game theoretic interpretation of efficiency analysis as a principle-agent game

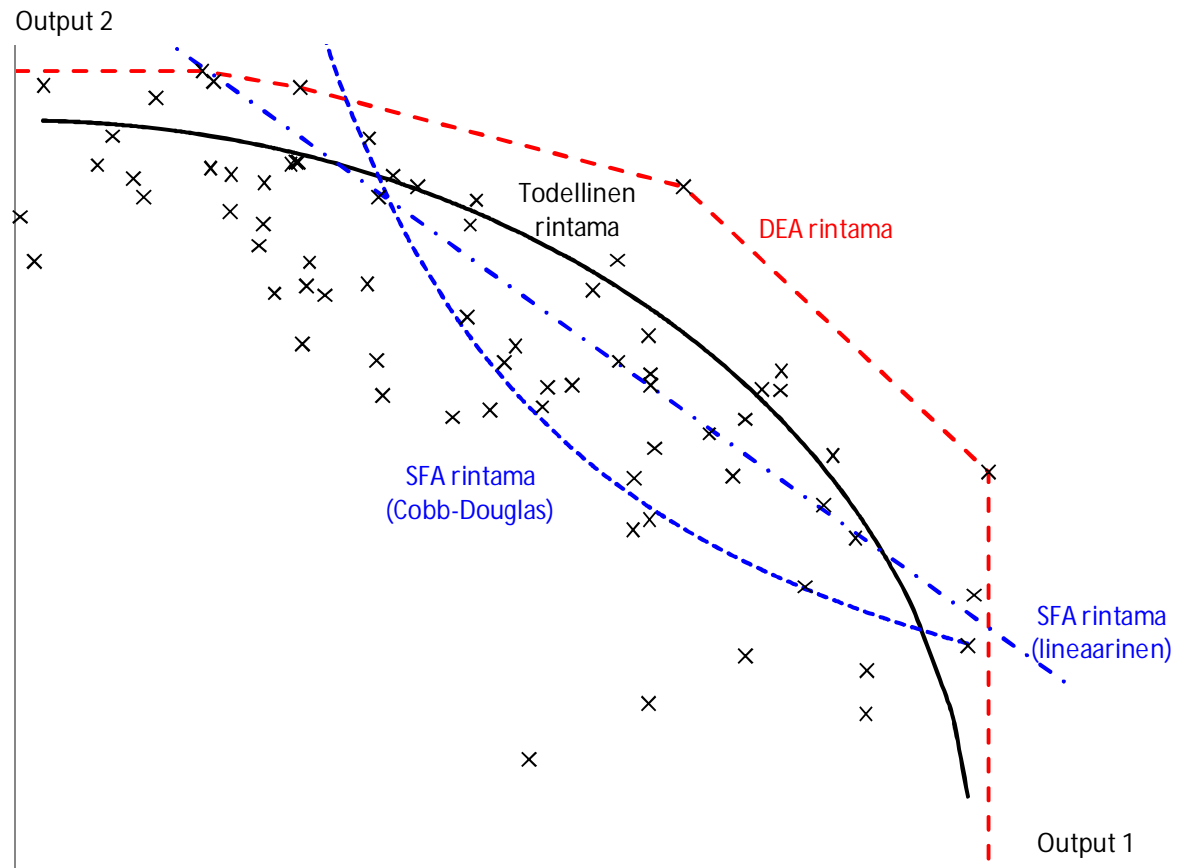
# Motivation & background

- In Finland, firms and households are free to purchase electricity from any producer or intermediary, but the distribution remains a local monopoly
- The Finnish Energy Market Authority (EMV) regulates the distribution to counter abuse of monopoly power
- The acceptable rate of return for the invested capital based on the capital asset pricing model (CAPM)
- The acceptable cost level based on the efficient cost frontier estimated by DEA and SFA methods

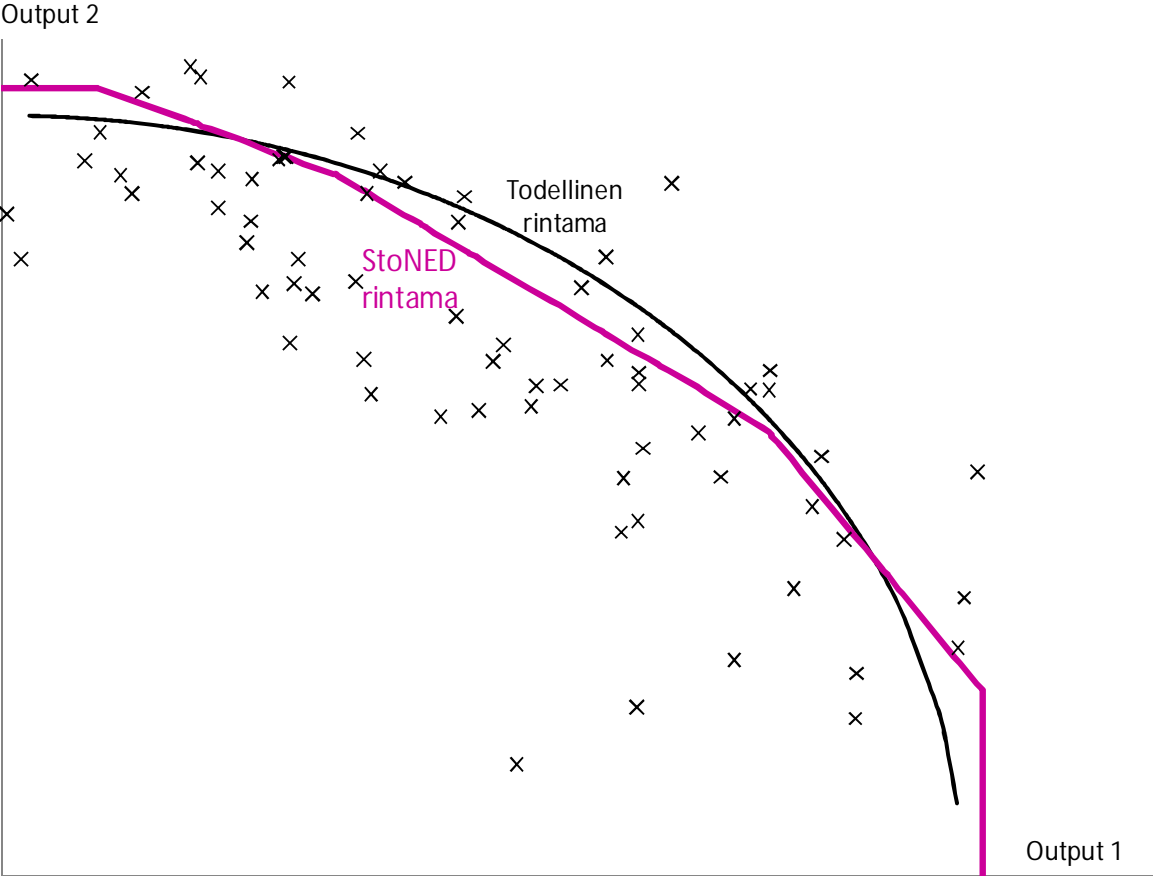
# Motivation & background

- Recent study by Kuosmanen, Kortelainen, Kultti, Pursiainen, Saastamoinen & Sipiläinen (2010) proposes several major reforms for the 3rd regulation period in 2012-2015:  
([www.energiamarkkinavirasto.fi](http://www.energiamarkkinavirasto.fi))
- Replace the average of DEA and SFA efficiency by the StoNED estimator that combines a nonparametric frontier with stochastic inefficiency and noise terms

# Simulated example with two outputs



# Simulated example with two outputs



# Cost frontier model

$$x = C(y_1, y_2, y_3) \cdot \exp(\delta z + u + v)$$

$C$  is a convex, monotonic increasing cost function that exhibits CRS

$y_1$  is the weighted amount of energy transmission (GWh of 0.4 kV equivalents)

$y_2$  is the total length of the network (km)

$y_3$  is the total number of customers connected to the network

$z$  is the proportion of underground cables

$u$  is a half-normally distributed inefficiency term

$v$  is a normally distributed noise term

# StoNED estimator

Step 1: solve the CNLS problem

$$\min_{\gamma, \beta, \delta, \varepsilon} \sum_{i=1}^n \varepsilon_i^2$$

s.t.

$$\ln x_i = \ln \gamma_i + \delta z_i + \varepsilon_i \quad \forall i$$

$$\gamma_i = \beta'_i y_i \quad \forall i$$

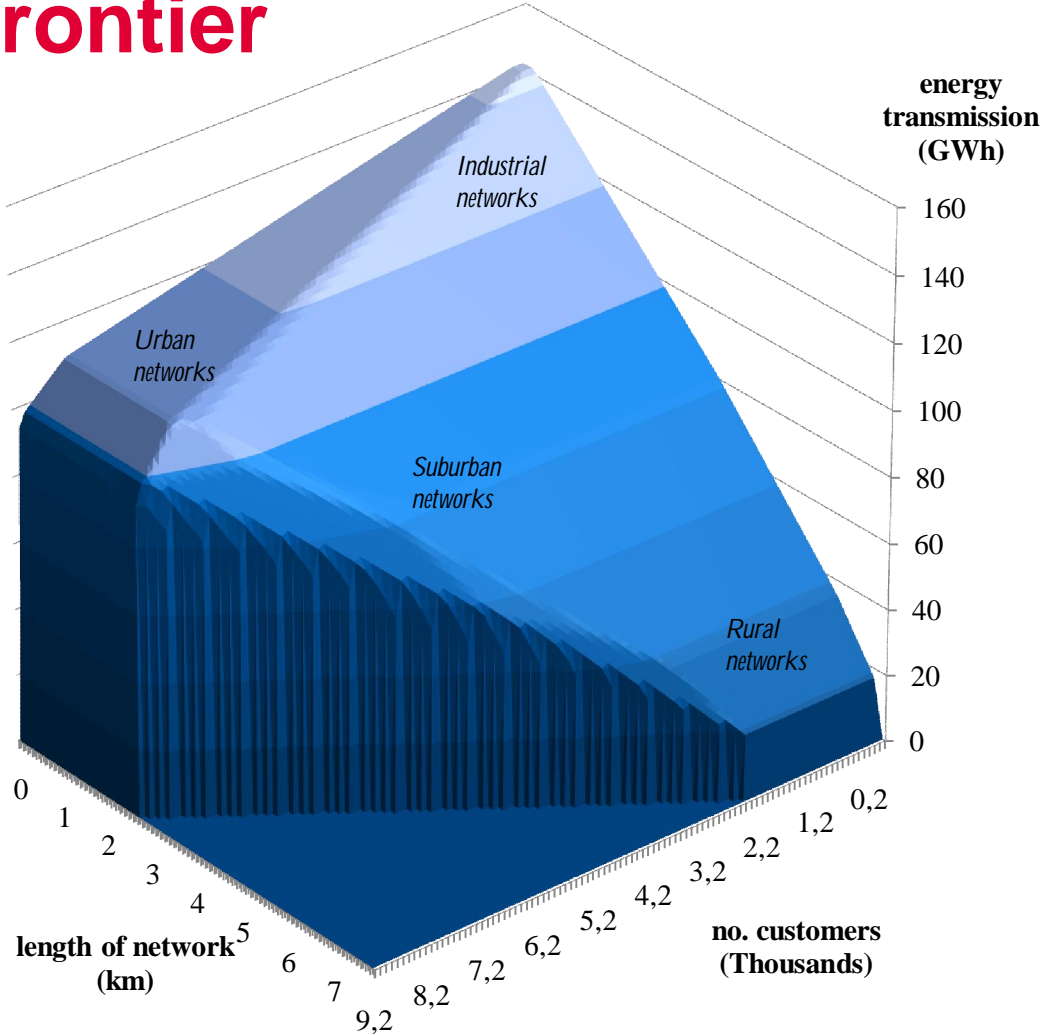
$$\gamma_i \geq \beta'_h y_i \quad \forall h, i$$

$$\beta_i \geq 0 \quad \forall i$$

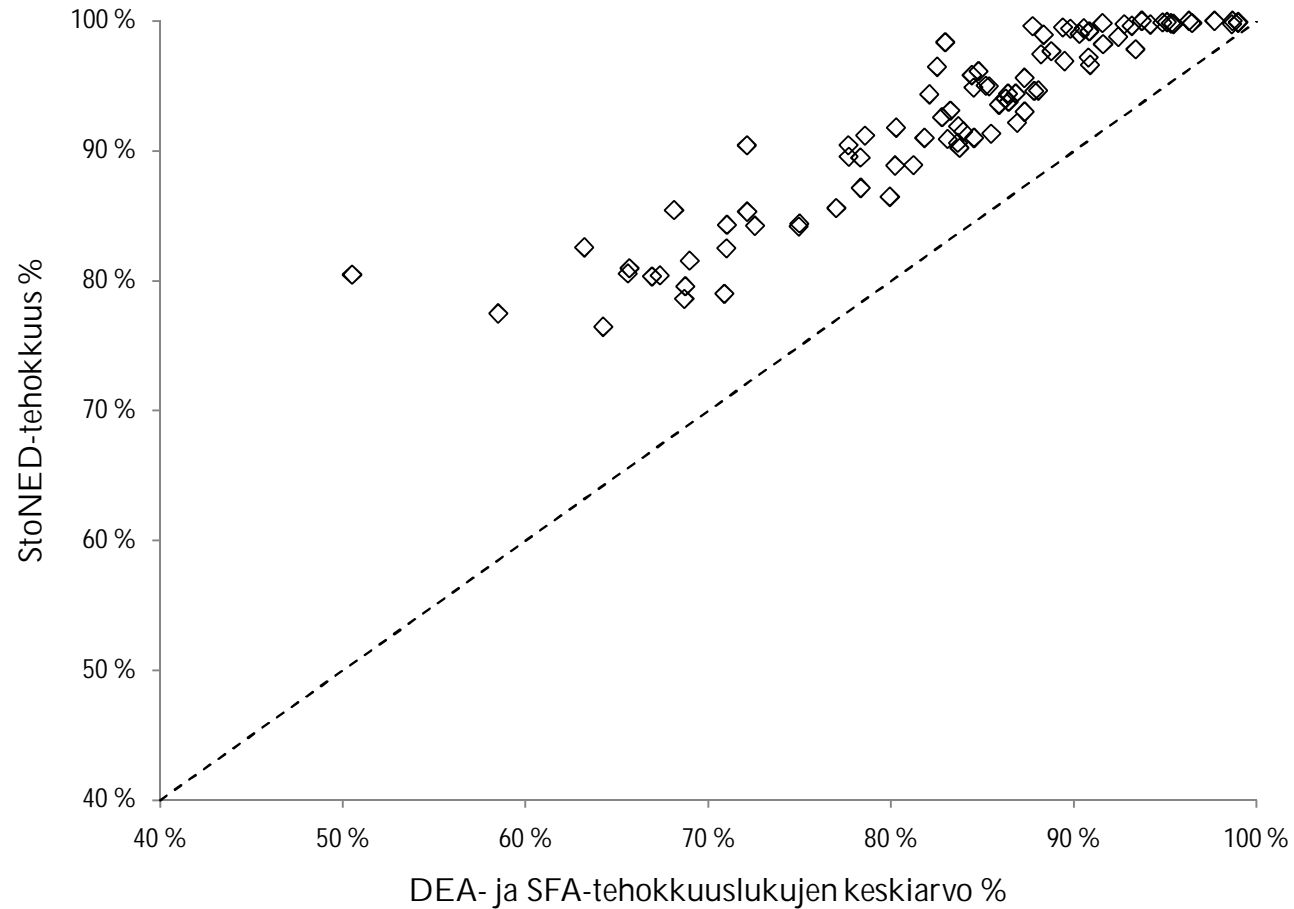
Step 2: Estimate the variance of  $u$  based on the skewness of residuals.

$$\hat{C}^{\text{StoNED}}(y) = \gamma_i + \hat{\sigma}_u \sqrt{2/\pi}$$

# StoNED frontier



# Comparison of efficiency estimates



## Efficiency improvement targets in Mill. € (prices of 2008)

	toimiala yht.	keskiarvo	keski- hajonta	minimi	maksimi
DEA	141,382	1,589	2,223	0,000	27,654
SFA	95,481	1,073	3,888	0,024	12,830
ka*	118,431	1,331	2,959	0,016	20,242
StoNED	47,508	0,534	1,326	0,000	11,113

# Game theoretic perspective

- DEA has a game theoretic interpretation as a Nash equilibrium of a principle-agent game (Banker 1980, *EJOR*)
- DEA is a restricted special case of StoNED (Kuosmanen & Johnson 2010, *Operations Research*)

# Game theoretic perspective

- DEA has a game theoretic interpretation as a Nash equilibrium of a principle-agent game (Banker 1980, *EJOR*)
- DEA is a restricted special case of StoNED (Kuosmanen & Johnson 2010, *Operations Research*)

⇒ StoNED must have a game theoretic interpretation as well

**Backward induction:** how to specify an elegant/intuitive game that has the StoNED method as its Nash equilibrium?

# Data Envelopment Analysis (DEA)

Outputs  $\mathbf{y}$ , Total cost  $x$

Envelopment problem

$$\min_{\lambda \geq 0} \theta$$

*s.t.*

$$\theta x_i \geq \sum_{h=1}^n \lambda_h x_h$$

$$\mathbf{y}_i \leq \sum_{h=1}^n \lambda_h \mathbf{y}_h$$

Multiplier problem

$$\max_{\mathbf{p}} \frac{\mathbf{p}' \mathbf{y}_i}{x_i}$$

*s.t.*

$$\frac{\mathbf{p}' \mathbf{y}_h}{x_h} \leq 1 \quad \forall h = 1, \dots, n$$

# Principle-agent game

- **Players:** principle (= regulator), agent (= regulated firm  $i$ )
- **Moves:** principle selects the benchmark firm  $h$  ( $h=1, \dots, n$ ), agent selects output  $k$  ( $k=1, 2, 3$ ). Simultaneous moves.
- **Strategies:** mixed strategies permitted
- **Payoffs:** based on relative efficiency (zero-sum game)  
Agent:  $(y_{ik} / x_i) / (y_{hk} / x_h)$   
Principle:  $1 - [(y_{ik} / x_i) / (y_{hk} / x_h)]$

# Data Envelopment Analysis (DEA)

*Outputs*  $\mathbf{y}$ ,      *Total cost*  $x$

Minimax formulation:

$$\min_{h \in \{1, \dots, n\}} \max_{\mathbf{p}} \left( \frac{\mathbf{p}' \mathbf{y}_i}{x_i} \right) / \left( \frac{\mathbf{p}' \mathbf{y}_h}{x_h} \right)$$

The optimal solution to the DEA problem provides the mixed strategy Nash equilibrium of the previous principle-agent game

# Interpretation

- In the DEA game, principle plays a separate game with each agent (lack of coordination)
- Principle chooses the benchmark to minimize efficiency (advocator of energy consumers)
- In the regulatory model of EMV, the role of EMV is to balance the interests of consumers and energy companies (and avoid interruptions)

# Variation 1

- **Sequential moves:** principle moves first, agent observes the benchmark and selects the output  
⇒ Non-convex FDH efficiency is the Nash equilibrium
- Agent better off if the benchmark is fixed first
- Principle better off in the simultaneous move game

## Variation 2

### Union of agents:

- What if agents form a union to represent them and play a single coordinated game with the principle?

- **Payoffs:** based on relative efficiency

Union:  $\sum_i (y_{ik} / x_i) / (y_{hk} / x_h)$

Principle:  $1 - [\sum_i (y_{ik} / x_i) / (y_{hk} / x_h)]$

- Nash equilibrium: 
$$\min_{h \in \{1, \dots, n\}} \max_{\mathbf{p}} \sum_{i=1}^n \left( \frac{\mathbf{p}' \mathbf{y}_i}{x_i} \right) / \left( \frac{\mathbf{p}' \mathbf{y}_h}{x_h} \right)$$

## Variation 2

### Union of agents:

- The parametric programming (PP) model by Aigner and Chu (1968, *AER*) gives the Nash equilibrium
  - The PP frontier envelops the DEA frontier
  - Principle is better off in the case of a union, none of the agents is better off
- ⇒ Agents do not have incentive to let a union choose the same strategy for all agents
- ⇒ If the union selects firm-specific output separately for each agent, we fall back to the DEA case

# Nash equilibrium of a bi-matrix game

- Mangasarian and Stone (1964) show that the Nash equilibrium of a two-person, nonzero-sum game with a finite number of pure strategies is equivalent to the optimal solution to a certain quadratic programming problem with linear constraints.
- CNLS estimator is obtained as the optimal solution to a quadratic programming problem.
- To develop the game theoretic interpretation, we can specify the payoff matrices to match the CNLS problem

## Variation 3

### Benevolent regulator:

- Suppose the objective of the principle is to minimize the sample variance of efficiency scores (show the industry in positive light)

- Payoff of the principle:

$$1 - \sum_{i=1}^n \left( Eff_{i,h} - \overline{Eff} \right)^2, \quad Eff_{i,h} = \left( \frac{\mathbf{p}'\mathbf{y}_i}{x_i} \right) / \left( \frac{\mathbf{p}'\mathbf{y}_h}{x_h} \right)$$

**Thank you!**

Ideas, comments, feedback & help  
are more than welcome



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