

## Optimal Taxation and R&D Policies

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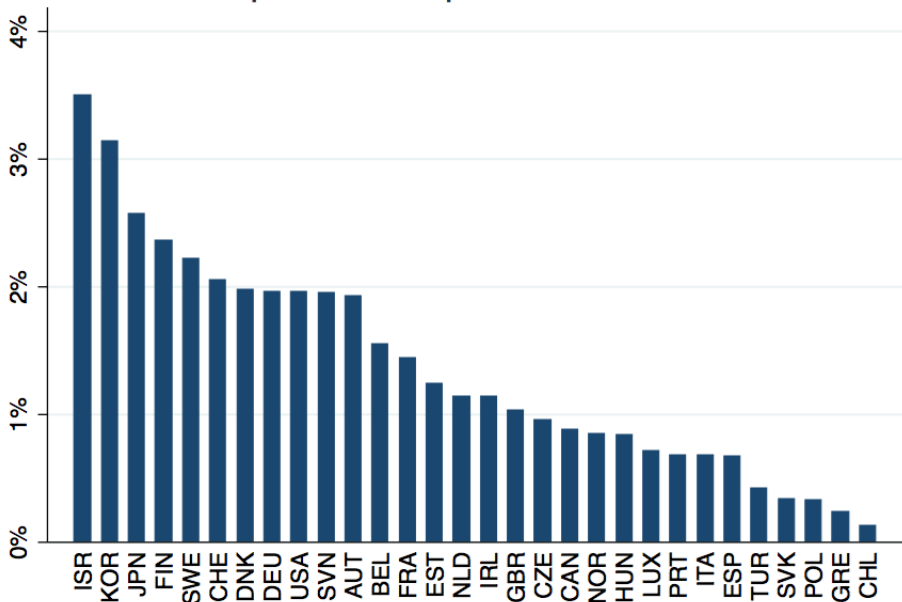
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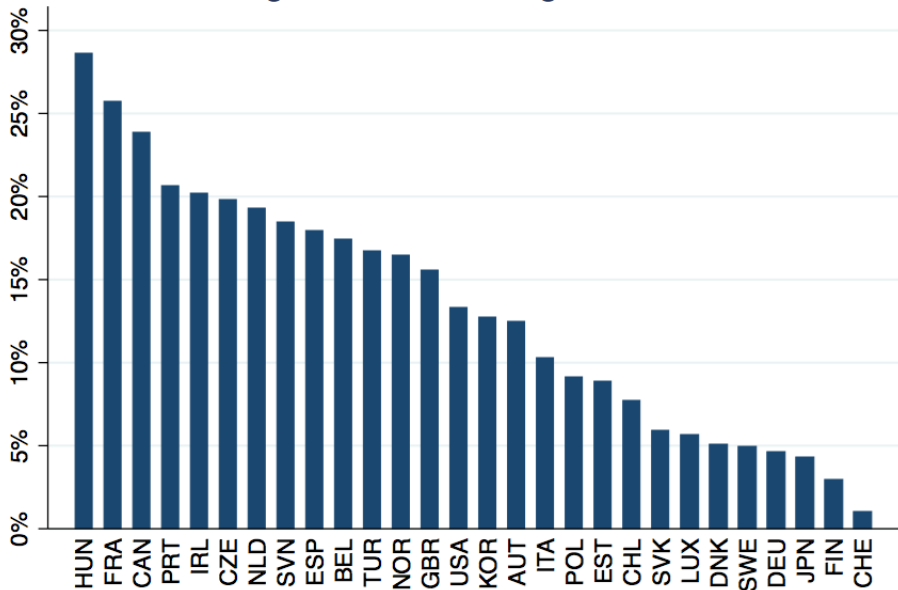
## Motivation: Widespread R&D and Industrial Policies

- Industrial policies are widespread, costly and not fully understood.
- EU: 9.6% of EU GDP on industrial policies in 2010.
- Lots of policies target **innovation and R&D**.

## Business enterprise R&D expenditures in 2012 as % of GDP



## Share of government funding in Business R&D



# Goal: Optimal Design of R&D and Corporate Taxation

- Little agreement on how effective these policies are and how to design them.
- In this paper: Optimal R&D and Corporate Taxation Policies.
  - ▶ Consider heterogeneous firms → Firm quality is **private information**.
  - ▶ Firm faces uncertainty about its quality over time.
  - ▶ R&D returns are stochastic.
  - ▶ **Spillovers** between firms.
  - ▶ Do not restrict policies ex ante: let optimal policy tools arise endogenously.
  - ▶ First: optimal mechanism design, then consider “simpler” policies.

## What we do:

- Model of R&D and endogenous growth with heterogeneous firms.
- Mechanism design setup: Firms are agents, government is the Principal.
- Derive analytically fully unrestricted optimal dynamic mechanism → characterize allocations using “wedges” or implicit taxes and subsidies.
- Estimate important parameters of the model using patent data matched to Compustat (not yet today)
  - ▶ Numerically illustrate life cycle path of optimal policies for firms (by age, quality..).
- How well do simpler policies perform (linear size-independent age-dependent, linear size and age-independent policies, etc..).
  - ▶ Formulas in terms of “sufficient statistics.”

## Related Literature

**R&D Policies and Growth:** Acemoglu, Akcigit, Bloom and Kerr (2013), Atkeson and Burstein (2014), Lentz and Mortensen (2015).

**Empirics of R&D Policies:** Bloom, Griffith, Van Reenen (2000), Criscuolo *et al.* (2013), Hall (1992), Goolsbee (1998), Romer (2001), Wilson (2009)

**Optimal Policy Design:** Hopenhayn and Nicolini (1997), Golosov, Tsyvinski and Werning (2006), Shourideh (2012), Farhi and Werning (2013, 2014), Stantcheva (2014), Pavan, Segal and Toikka (2013).

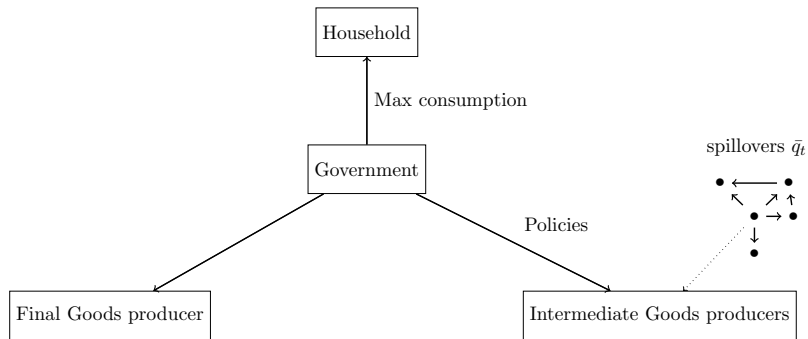
# Outline

- 1 Model
- 2 Optimal Unrestricted Mechanism
- 3 Simpler Policies and Sufficient Statistics
- 4 Conclusion

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# Model: Summary



$$Y(\bar{q}_t) = \int_{\Theta^t} k(\theta^t)^{1-\beta} q(\theta^t)^\beta p(\theta^t) d\theta^t$$

- Manager: effort  $l_t$
- R&D :  $q(\theta^t) = q(\theta^{t-1}) + \lambda_t(r(\theta^{t-1}), l(\theta^t), \theta_t)$
- Produce  $k(\theta^t)$
- $\pi(\theta^t, \bar{q}_t) = \max_k \{p(\theta^t)k - C(k, \bar{q}_t)\}$

# Model: Intermediate Producers R&D Investments and Efforts

- Each producer can improve his quality  $q_t$  through R&D:  
$$q_t = (1 - \delta)q_{t-1} + \lambda_t$$
  - ▶  $\lambda_t$ : step size,  $\delta$ : depreciation factor.
- The step size  $\lambda_t(r_{t-1}, l_t, \theta_t)$  depends on:
  - ▶ R&D resources  $r_t$  cost  $M_t(r_t)$ .
  - ▶ Type  $\theta_t$  (managerial/firm quality).
  - ▶ Managerial/firm effort:  $l_t$  at cost  $\phi_t(l_t)$ .
  - ▶  $\frac{\partial \lambda}{\partial \theta} > 0$ ,  $\frac{\partial \lambda}{\partial r} > 0$ ,  $\frac{\partial \lambda}{\partial l} > 0$ ,  $\frac{\partial^2 \lambda}{\partial \theta \partial l} > 0$  (screening).
  - ▶ Returns to R&D are stochastic, depend on stochastic type.

## Model: Intermediate Producers' Production

- Firm born with  $\theta_1$ , evolves with Markov  $f^t(\theta_t|\theta_{t-1})$ .
- History:  $\theta^t = \{\theta_1, \dots, \theta_t\}$  with probability  $P(\theta^t) = f^t(\theta_t|\theta_{t-1})\dots f^1(\theta_1)$ .
- Firm lifecycle is  $T$  periods (max time after which exogenous death).
- $k(\theta^t)$ : quantity and  $q(\theta^t)$ : quality.
- Linear production cost:  $C(k, \bar{q}_t) = \frac{k}{\bar{q}_t}$
- Aggregate quality (Spillovers):  $\bar{q}_t = \int_{\Theta^t} q(\theta^t) P(\theta^t) d\theta^t$
- Hence profits are:  
$$\pi(\theta^t, \bar{q}_t) \equiv \max_k \{p(\theta^t)k - C(k, \bar{q}_t)\} = q_t(\theta^t)(1-\beta)^{\frac{1-\beta}{\beta}} \cdot \beta \cdot \bar{q}_t^{\frac{1-\beta}{\beta}}$$
- Final good:  $Y_t = \int_{\Theta^t} q(\theta^t)^\beta k(\theta^t)^{1-\beta} P(\theta^t) d(\theta^t)$

## Government Policies Studied

- Fully unrestricted mechanism.
- Linear (size-independent) age-dependent corporate tax and R&D subsidies.
- Linear (size- and age-independent) corporate tax and R&D subsidies.
- Nonlinear (i.e., size dependent) corporate tax and R&D subsidies.

What do these policies look like at optimum?

What is the welfare gain from each?

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- 3 Simpler Policies and Sufficient Statistics
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## Direct Revelation Mechanism with Spillovers

- In this section:  $\theta_t = \rho\theta_{t-1} + \psi_t$ , with  $\psi_t$  distributed  $N(0, \sigma_\psi^2)$ .
- History  $\theta^t$  and effort  $l_t$  are private info.
- Government sees: step size  $\lambda$ , realized quality  $q$ , R&D spending  $r$ , and production  $k$ .
- Direct revelation: Firm reports  $\theta'_t(\theta^t)$ .
- History of reports:  $\theta'^t = \{\theta'_1(\theta_1), \dots, \theta'_t(\theta^t)\}$ .
- Allocations as function of history of reports:  $\{\lambda(\theta'^t), r(\theta'^t), T_t(\theta'^t)\}$ .
- Maximize household consumption (wlog, zero weight on manager):  
$$\mathbb{E} \left\{ \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} \left\{ \pi(\theta^t, \bar{q}_t) - M_t(r(\theta^t)) - T_t(\theta^t) \right\} \right\}$$
- First: let's assume full depreciation  $\delta = 1$ .

# Incentive compatibility and a First-order Approach

- Lifetime expected utility from assigned allocations under truthful reporting:  $V_1(\theta_1) \equiv V_1(\{\lambda(\theta^s), r(\theta^s), T_s(\theta^s)\}_{s=1}^T, \theta_1) = \sum_{t=1}^T (\frac{1}{R})^{t-1} \cdot \left\{ \int_{\Theta^t} \{ T_t(\theta^t) - \phi_t(\lambda(\theta^t), r(\theta^{t-1}), \theta_t) \} P(\theta^t | \theta_1) d\theta_t \right\}$
- Utility for a given sequence of realizations  $\theta^T$ :  
 $\tilde{U}(\theta^T) = \sum_{t=1}^T (\frac{1}{R})^{t-1} \{ T_t(\theta^t) - \phi_t(\lambda(\theta^t), r(\theta^{t-1}), \theta_t) \}$
- Envelope condition (Pavan, Segal and Toikka, 2014):  
 $\frac{\partial V_t(\theta^t)}{\partial \theta_t} = \mathbb{E} \left\{ \sum_{s=t}^T l_{t,s} \frac{\partial \tilde{U}(\theta^T)}{\partial \theta_s} \right\}$ 
  - ▶  $l_{t,s}$ : impulse response of shock  $\theta_t$  on time  $s$  shock  $\theta_s$ . For AR(1) is  $\rho^{s-t}$ .
- Informational rent:  $V_1(\theta_1) = \mathbb{E} \left\{ \frac{1 - F^1(\theta_1)}{f^1(\theta_1)} \sum_{t=1}^T \left( l_{1,t} \frac{\partial \tilde{U}(\{\lambda(\theta^s), r(\theta^s), T_s(\theta^s)\}_{s=1}^T, \theta^T)}{\partial \theta_t} \right) \right\} + V_1(\underline{\theta}_1)$ .

## Program: Maximize Virtual Surplus with Spillovers

Notation:  $\lambda(\theta^t) = \lambda(r(\theta^{t-1}), I(\theta^t), \theta_t)$ .

$$\max \sum_{t=1}^T \left(\frac{1}{R}\right)^{t+1} \left\{ \int_{\Theta^t} \{ \pi_t(\theta^t, \bar{q}_t) - M_t(r(\theta^t)) - \phi(I(\theta^t)) - \frac{1 - F^1(\theta_1)}{f^1(\theta_1)} p^t [\phi'(I(\theta^t)) \frac{\partial \lambda(\theta^t) / \partial \theta_t}{\partial \lambda(\theta^t) / \partial I_t}] - V_1(\underline{\theta}_1) \} P(\theta^t) d\theta^t \right\}$$

s.t.:

$$\int_{\Theta^t} \lambda(\theta^t) P(\theta^t) d\theta^t = \bar{q}_t$$

- Solve using this first-order approach, then verify global (IC) (“sufficiency”) *ex post* (Farhi and Werning, 2013).
- Cohort-by-cohort problem (pooling across cohorts can be a second step).

## Wedges: Measures of Implicit taxes and Subsidies

- Implicit corporate tax:

$$(1 - \tau(\theta^t)) \frac{\partial \pi_t(\theta^t)}{\partial \lambda_t} \frac{\partial \lambda(\theta^t)}{\partial l_t} = \phi_{lt}(\lambda(\theta^t), r(\theta^{t-1}), \theta_t)$$

- Implicit R&D subsidy:

$$(1 - \tilde{s}(\theta^t)) M'_t(r(\theta^t)) = \frac{1}{R} \mathbb{E} \left\{ \frac{\partial \pi_{t+1}(\theta^{t+1})}{\partial \lambda_{t+1}} \cdot \frac{\partial \lambda(\theta^{t+1})}{\partial r_t} (1 - \tau(\theta^{t+1})) \right\}$$

- Net R&D subsidy:

$$s(\theta^t) = \tilde{s}(\theta^t) - \frac{1}{M'_t(r(\theta^t))} \frac{1}{R} \mathbb{E} \left\{ \frac{\partial \pi_{t+1}(\theta^{t+1})}{\partial \lambda_{t+1}} \cdot \frac{\partial \lambda(\theta^{t+1})}{\partial r_t} \tau(\theta^{t+1}) \right\}$$

## Sufficient Statistics: Elasticities and Coefficient of Complementarity

- Elasticity of  $x_t$  w.r.t  $y_t$ :  $\varepsilon_{xy,t} \equiv \frac{\partial x_t}{\partial y_t} \frac{y_t}{x_t}$
- Hicksian coefficient of complementarity between  $x$  and  $y$  in  $\lambda$ : For  $\{x, y\} \in \{\theta_t, r_{t-1}, l_t\}$

$$\rho_{xy} = \frac{\frac{\partial^2 \lambda}{\partial x \partial y} \lambda}{\frac{\partial \lambda}{\partial x} \frac{\partial \lambda}{\partial y}}$$

- $\lambda_t(r, l, \theta) = r l \theta \rightarrow \rho_{\theta l} = \rho_{\theta r} = \rho_{l r} = 1.$
- $\lambda_t(r, l, \theta) = r + l + \theta \rightarrow \rho_{\theta l} = \rho_{\theta r} = \rho_{l r} = 0.$

# Optimal R&D Subsidy and Corporate Tax

- Optimal Corporate Tax:

$$\left( \tau(\theta^t) + \underbrace{\frac{\int \frac{\partial \pi_t(\theta^t, \bar{q}_t)}{\partial \bar{q}_t}}{\frac{\partial \pi_t(\theta^t, \bar{q}_t)}{\partial \lambda_t}}}_{\text{Externality}} \right) \frac{1}{1 - \tau(\theta^t)} = \underbrace{\frac{1 - F^1(\theta_1)}{f^1(\theta_1)} \rho^{t-1}}_{\text{Type distribution and persistence}} \underbrace{\frac{\varepsilon_{\lambda\theta,t}}{\theta_t} \left( \frac{1 + \rho_{\theta I,t} \varepsilon_{\lambda(1-\tau),t}}{\varepsilon_{\lambda(1-\tau),t}} \right)}_{\text{Efficiency Cost}}$$

- Optimal R&D Subsidy: tension between externality and asymmetric info.

$$s(\theta^t) = \frac{1}{M'_t} \frac{1}{R} \mathbb{E} \left( \underbrace{\frac{\partial \lambda(\theta^{t+1})}{\partial r_t} \int \frac{\partial \pi_{t+1}(\theta^{t+1}, \bar{q}_{t+1})}{\partial \bar{q}_{t+1}}}_{\text{Externality}} \right) + \underbrace{\frac{1 - F^1(\theta_1)}{f^1(\theta_1)} \rho^t (1 - \tau(\theta^{t+1}))}_{\text{Type distribution and persistence}} \frac{\partial \pi_{t+1}(\theta^t, \bar{q}_{t+1})}{\partial \lambda_{t+1}} \frac{\partial \lambda_{t+1}}{\partial \theta_{t+1}} \frac{\partial \lambda_{t+1}}{\partial r_t} \frac{1}{\lambda_{t+1}} \underbrace{(\rho_{I r} - \rho_{\theta r})}_{\text{Complementarity coefficients}}$$

## Special Cases for the Optimal R&D Subsidy when only Externality Matters

- Multiplicatively separable step size:  $\lambda_t(r, l, \theta) = h^1(r)h^2(\theta)h^3(l)$  has  $\rho_{lr} = \rho_{\theta l} = \rho_{\theta r} = 1$ .

$$s(\theta^t) = \frac{1}{M'_t(r(\theta^t))} \frac{1}{R} \mathbb{E} \left( \frac{\partial h^1(r(\theta^t))}{\partial r_t} h^2(\theta_{t+1}) h^3(l(\theta^{t+1})) \int \frac{\partial \pi_{t+1}(\theta^{t+1}, \bar{q}_{t+1})}{\partial \bar{q}_{t+1}} \right)$$

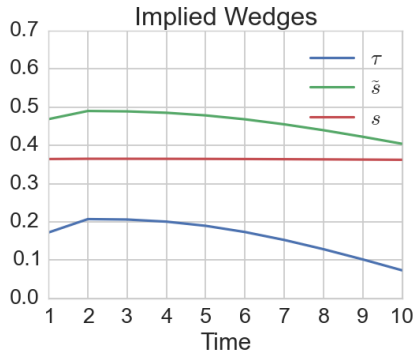
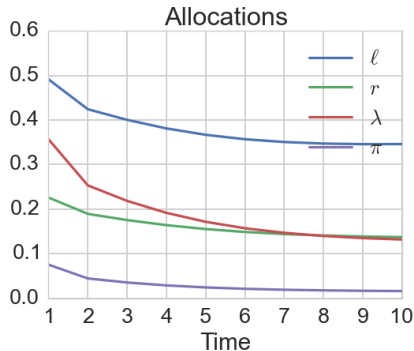
- CES step size:  $\lambda(l, r, \theta) = (\alpha_1 l^{(1-\rho)} + \alpha_2 r^{(1-\rho)} + \alpha_3 \theta^{(1-\rho)})^{\frac{1}{1-\rho}}$  has  $\rho_{lr} = \rho_{\theta l} = \rho_{\theta r} = \rho$ .

$$s(\theta^t) = \frac{1}{M'_t(r(\theta^t))} \frac{1}{R} \mathbb{E} \left( \alpha_2 \left( \frac{\lambda(\theta^t)}{r(\theta^t)} \right)^\rho \int \frac{\partial \pi_{t+1}(\theta^{t+1}, \bar{q}_{t+1})}{\partial \bar{q}_{t+1}} \right)$$

## A Simple Numerical Illustration

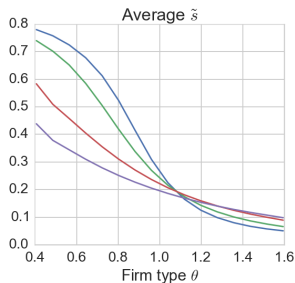
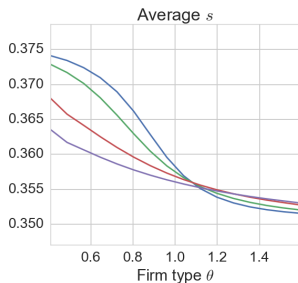
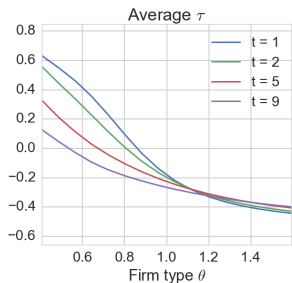
- Next step: Estimation from Patent data matched to Compustat.
- Functional forms:  $\lambda(r, l, \theta) = w(\theta, r) = (\alpha r^{(1-\rho)} + (1-\alpha)\theta^{(1-\rho)})^{\frac{1}{1-\rho}} l$ .
  - ▶ Two cases:  $\rho = 0.8$  and  $\rho = 1.2$  (keep average  $\lambda$  constant across two cases).
  - ▶  $\alpha = 0.2$ .
- Constant elasticity disutility:  $\phi(l) = \frac{l^{1+\gamma}}{1+\gamma}$ .
  - ▶  $\gamma = 2$  generates Frisch elasticity of 0.5.
- Constant elasticity R&D cost:  $M_t(r) = \frac{r^{1+\kappa}}{1+\kappa}$ .
  - ▶  $\kappa = 1$  (typically quadratic R&D cost estimates).
- AR(1) process for type:  $\theta_t = \rho\theta_{t-1} + \psi_t$ 
  - ▶  $\rho = 0.9$ ,  $\sigma_\psi^2 = 0.01$ .

## Policies for Old and New Firms, $\rho = 0.8$



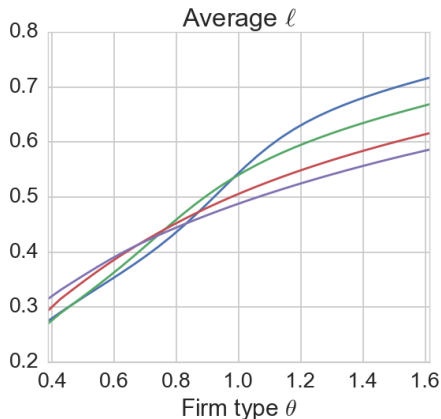
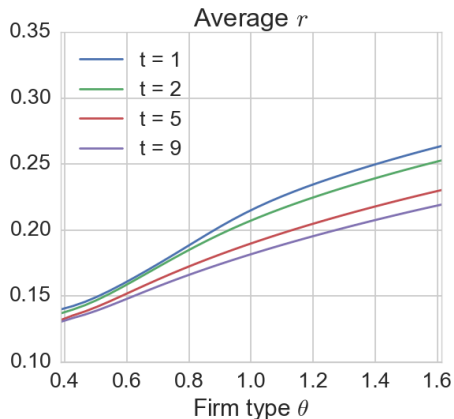
- Can incentivize less and less effort and R&D over time.
- Optimal to make young firms invest most.
- An almost constant net subsidy is optimal (but as  $T \rightarrow \infty$ ,  $s_t \rightarrow 0$ .)

## Wedges for Higher and Lower Quality Firms, $\rho = 0.8$



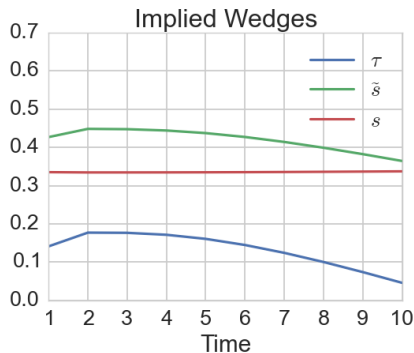
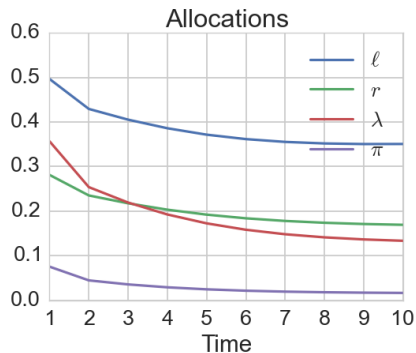
- Gross subsidy tracks corporate tax (else too high disincentive on R&D).
- Higher quality firms are subsidized to produce more (big externality).
- Optimal subsidy highly nonlinear.

## Policies for Higher and Lower Quality Firms, $\rho = 0.8$



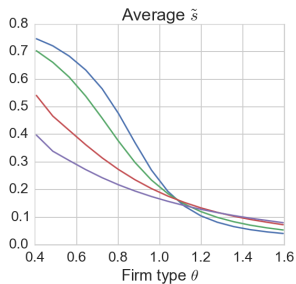
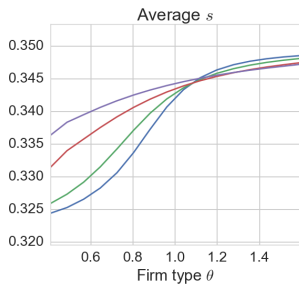
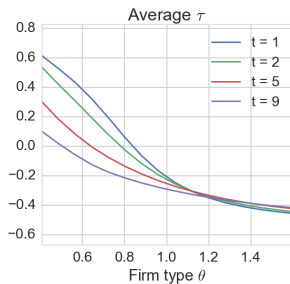
- Monotonic allocations: high quality firms have more inputs.
- Allocations get flatter over time: less able to screen.

# Policies for Old and New Firms, $\rho = 1.2$



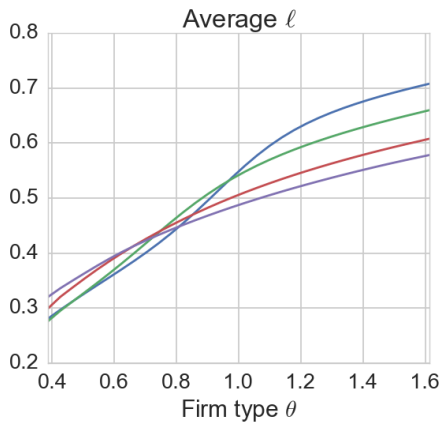
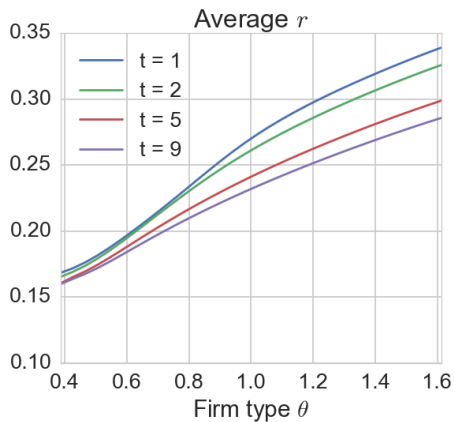
- As  $\rho$  is higher,  $s$  is lower.

# Wedges for Higher and Lower Quality Firms, $\rho = 1.2$



- As  $\rho$  is higher,  $s$  is lower.

# Policies for Higher and Lower Quality Firms, $\rho = 1.2$



## Case with no Depreciation: Externality Persists Forever

$$\left( \tau(\theta^t) + \frac{\sum_{s=t}^T \int \frac{\partial \pi_s(\theta^s, \bar{q}_s)}{\partial \bar{q}_s}}{E(\sum_{s=t}^T (\frac{1}{R})^{s-t} \frac{\partial \pi_s(\theta^s, \bar{q}_s)}{\partial \lambda_t} \frac{\partial \lambda_t}{\partial l_t})} \right) \frac{1}{1 - \tau(\theta^t)} =$$

$$\frac{1 - F^1(\theta_1)}{f^1(\theta_1)} \rho^{t-1} \left( \frac{\varepsilon_{\lambda \theta, t}}{\theta_t} \right) \left( \frac{1 + \rho \theta_{l, t} \varepsilon_{\lambda(1-\tau), t}}{\varepsilon_{\lambda(1-\tau), t}} \right)$$

$$s(\theta^t) = \frac{1}{M'_t(r(\theta^t))} \frac{1}{R} \mathbb{E} \left( \frac{\partial \lambda_{t+1}}{\partial r_t} \sum_{s=t+1}^T \int \frac{\partial \pi_s(\theta^s, \bar{q}_s)}{\partial \bar{q}_s} + \right.$$

$$\left. \frac{1 - F^1(\theta_1)}{f^1(\theta_1)} \rho^t \frac{\partial \lambda_{t+1}}{\partial r_t} \frac{\partial \lambda_{t+1}}{\partial \theta_{t+1}} \frac{\phi'(l_{t+1})}{\lambda_{t+1}} \frac{1}{\frac{\partial \lambda_{t+1}}{\partial l_{t+1}}} (\rho l_r - \rho \theta_r) \right)$$

- Same special cases apply when subsidy only set based on (now persistent) externality.

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# Optimal Linear (Size-independent) Age-Dependent Policies

- Linear age-dependent corporate tax  $\tau_t$  and R&D subsidy  $s_t$ .
- Manager's optimization problem:

$$\max_{\{s_{ti}, r_{ti}\}_{t=1}^T} \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} ((1 - \tau_t)\pi_{ti} - (1 - s_t)M_t(r_{ti}) - \phi_t(l_{ti}))$$

- Hence  $\pi_{ti}((1 - \tau_t), s_{t-1}, \bar{q}_t((1 - \tau_t), s_{t-1}))$  and  $r_{t-1i}((1 - \tau_t), s_{t-1}, \bar{q}_t((1 - \tau_t), s_{t-1}))$  are solutions as functions of policies.
- Social welfare is now just total revenue:

$$\max_{\{\tau, s\}_{t=1}^T} \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} (\tau_t \mathbb{E}(\pi_{ti}) - s_t \mathbb{E}(M_t(r_{ti})))$$

- Aggregate (or cross-sectional averages):  $r_t = \mathbb{E}(r_{ti})$  and  $\pi_t = \mathbb{E}(\pi_{ti})$ .

## Optimal Formulas as Function of Sufficient Statistics

- Formulas valid even if other instrument not optimally set.
- Useful to evaluate “reforms.”
- Could use structural or reduced-form (sufficient stats) approach.
- Fiscal externality from one tax on other tax' base.

$$\tau_t^* = \frac{1 + s_{t-1} \varepsilon_{r,t-1}^{1-\tau} \frac{r_{t-1}}{\pi_t}}{1 + \varepsilon_{\pi,t}^{1-\tau}} \quad s_t^* = \frac{1 + \tau_t \varepsilon_{\pi,t}^{s-1} \frac{\pi_t}{r_t}}{1 + \varepsilon_{r,t}^{s-1}}$$

with

$$\varepsilon_{r,t-1}^{1-\tau} = \frac{d\mathbb{E}(M(r_{t-1i}))}{d(1-\tau)} \frac{1-\tau}{r_{t-1}} \quad \varepsilon_{\pi,t}^{1-\tau} = \frac{d\pi_t}{d(1-\tau)} \frac{1-\tau}{\pi_t}$$

$$\varepsilon_{\pi,t}^{s-1} = \frac{d\pi_t}{d(s-1)} \frac{s-1}{\pi_t} \quad \varepsilon_{r,t}^{s-1} = \frac{d\mathbb{E}(M(r_{ti}))}{d(s-1)} \frac{s-1}{r_t}$$

## Optimal Size and Age-Independent Policies

- Same Formula and same sufficient stats – different estimation required in data.

$$\tau^* = \frac{1 + s\varepsilon_r^{1-\tau} \frac{r}{\pi}}{1 + \varepsilon_\pi^{1-\tau}} \quad s^* = \frac{1 + \tau\varepsilon_\pi^{s-1} \frac{\pi}{r}}{1 + \varepsilon_r^{s-1}}$$

$$\pi = \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} \pi_t \quad r = \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} r_{t-1}$$

$$\varepsilon_r^{1-\tau} = \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} \varepsilon_{r,t-1}^{1-\tau} \frac{r_{t-1}}{r} \quad \varepsilon_\pi^{1-\tau} = \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} \varepsilon_{\pi,t}^{1-\tau} \frac{\pi_t}{\pi}$$

$$\varepsilon_\pi^{s-1} = \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} \varepsilon_{\pi,t}^{s-1} \frac{\pi_t}{\pi} \quad \varepsilon_r^{s-1} = \sum_{t=1}^T \left(\frac{1}{R}\right)^{t-1} \varepsilon_{r,t}^{s-1} \frac{r_t}{r}$$

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## Conclusion

- Model of R&D investments and endogenous growth with heterogeneous firms, private information, and spillovers.
  - ▶ Use mechanism design to solve for constrained efficient allocations.
  - ▶ Characterize using “wedges.”
- Optimal to subsidize R&D investments because of externality.
- Asymmetric info: could tend to reduce R&D subsidy if very complementary to firm quality.
  - ▶ But optimal R&D subsidy is nonlinear.
- Simple sufficient stats formula for R&D subsidy and corporate tax in linear, age-dependent or age-independent cases.
- Next step: estimate model using patent + compustat data.