Green Innovation and Financing Constraints

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Motivation

- Access to finance is one of the major barriers to firms' innovative activity
- This is even more relevant in the context of green innovations
 - Green innovations suffer from a double externality problem
 - Access to debt is more difficult for new and immature technologies
 - Financing costs are higher in the absence of a lending relationship with a bank

- A directed technological change model with clean and dirty inputs
- The cost of financing negatively depends on the market share of each technology
- Policies that redirect investment towards clean technology incentivises innovation in this sector and ease credit constraints

Previous Literature

- Environment and directed technological change (Gerlagh, 2008, Acemoglu et al., 2012, 2016, Greaker et al., 2018, Lennox and Witajewski-Baltvilks, 2017, Fried, 2018, Hart, 2019, Smulders et al., 2020, Lemoine, 2020, Wiskich, 2021)
- We introduce financing costs (King and Levine, 1993, D'Orazio and Valente, 2019, Haas and Kempa, 2021, Pan et al., 2021)
- Empirical literature on green innovation, financing constraints, and costs of capital (Olmos et al., 2012, Ghisetti et al., 2017, Howell, 2017, Egli et al., 2018, Jensen et al., 2019, Cecere et al., 2020, Kempa et al., 2021, Polzin et al., 2021, Noailly and Smeets, 2021)

The Model

The Economy

- The economy features three sectors:
 - 1. A manufacturing sector producing a homogeneous good using a clean intermediate input and a dirty intermediate input
 - 2. Two intermediate sectors, producing the two intermediate inputs using labour and a continuum of machines
 - 3. Two research sectors producing machines using foregone consumption good and innovating by employing scientists
- Footlose workers and scientists
- A representative infinitely-lived household inhabited by a unit mass of research firms in each sector, *L* workers, and *H* scientists with a CRRA utility function

- The unique final good, *Y*_t, is produced competitively by a representative firm according to

$$Y_t = \left(Y_{ct}^{(\epsilon-1)/\epsilon} + Y_{dt}^{(\epsilon-1)/\epsilon}\right)^{(\epsilon-1)/\epsilon} \tag{1}$$

- The intermediate goods are gross substitute

The Intermediate Sectors

- Each intermediate input is produced using labour and a unit mass of sector-specific machines,

$$Y_{jt} = L_{jt}^{1-\alpha} \int_0^1 A_{jit}^{1-\alpha} x_{jit}^{\alpha} di, \quad \forall j = \{c, d\}, \qquad (2)$$

- L_{jt} is labour demand in sector *j* at time *t*
- A_{jit} is the quality of machine $i \in [0, 1]$ in sector j at time t
- x_{jit} is the quantity demanded of this machine

The Intermediate Sectors and Carbon Emissions

- Carbon emissions are $E_t = \kappa Y_{dt}$
- Cumulative emissions at time *t* are given by

$$S_t = \sum_{\tau=0}^t \kappa Y_{d\tau}.$$
 (3)

The Research Sectors: Machine Production

- Each machine is supplied by monopolistically competitive firms
- One unit of each machine can be produced at cost $\psi \equiv \alpha^2$ units of the final good

The Research Sectors: Innovation

- A machine producer can hire H_{jit} scientists to increase the quality of its machine to

$$A_{jit} = \begin{cases} A_{jt-1} \left(1 + \gamma H_{jit}^{\eta} \left(\frac{A_{t-1}}{A_{jt-1}} \right)^{\phi} \right), & \text{with probability } \lambda_{jt} \\ A_{jt-1}, & \text{with probability } 1 - \lambda_{jt}, \end{cases}$$
(4)

where 0 $\leq \eta <$ 1, $\gamma >$ 0, and 0 $\leq \phi \leq$ 1

- A_{jt} is the average quality of the machines in sector j at the end of period t
- $A_t \equiv A_{ct} + A_{dt}$ is aggregate technology

The Research Sectors: Spillovers

- There are two types of spillovers
- Within a sector after one period, when discoveries are observed by other machine producers in the same sector and can be incorporated into their own innovation processes
- Across sectors, as a relatively backward sector *j* has a productivity advantage equal to the catch-up ratio

The Financing Costs I

- We assume a flow mismatch between the payments of the factor of production and the realised revenues (Mendoza, 2010, Jermann and Quadrini, 2012)
- With perfect credit markets, there is no interest on the intra-period loan
- We introduce heterogenous financing costs, which are assumed to be proportional to the volume of credit (Bernanke et al., 1999)
- These could represent monitoring costs (Townsend, 1979, Gale and Hellwig, 1985), screening costs (King and Levine, 1993), heterogeneous risk assessment, due diligence processes, and the lack of standardised investment structures (Egli et al., 2018)

- The maximisation problem of the producer of machine *i* in sector *j* is

$$\max_{p_{jit}, x_{jit}, H_{jit}} \left(p_{jit} - \frac{\psi}{\nu_{jt}} \right) x_{jit} - \frac{w_{jt}^s}{\lambda_{jt} \nu_{jt}^s} H_{jit}$$

s.t. demand for machine

innovation possibility frontier

$$\frac{H_{dt}}{H_{ct}} = \left[\left(\frac{A_{dt-1}}{A_{dt-1}} \right)^{1-\phi} \left(\frac{p_{dt}}{p_{ct}} \right)^{1/(1-\alpha)} \left(\frac{L_{dt}}{L_{ct}} \right) \left(\frac{\nu_{dt}}{\nu_{ct}} \right)^{1/(1-\alpha)} \left(\frac{\lambda_{dt}}{\lambda_{dt}} \right)^2 \right]^{\frac{1}{1-\eta}}$$
(5)

$$\frac{H_{dt}}{H_{ct}} = \left[\left(\frac{A_{dt-1}}{A_{dt-1}} \right)^{1-\phi} \left(\frac{p_{dt}}{p_{ct}} \right)^{1/(1-\alpha)} \left(\frac{L_{dt}}{L_{ct}} \right) \left(\frac{\nu_{dt}}{\nu_{ct}} \right)^{1/(1-\alpha)} \left(\frac{\lambda_{dt}}{\lambda_{dt}} \right)^2 \right]^{\frac{1}{1-\eta}}$$
(5)

- A productivity effect which directs innovation to the relatively more advanced sector

$$\frac{H_{dt}}{H_{ct}} = \left[\left(\frac{A_{dt-1}}{A_{dt-1}} \right)^{1-\phi} \left(\frac{p_{dt}}{p_{ct}} \right)^{1/(1-\alpha)} \left(\frac{L_{dt}}{L_{ct}} \right) \left(\frac{\nu_{dt}}{\nu_{ct}^s} \right) \left(\frac{\nu_{dt}}{\nu_{ct}} \right)^{1/(1-\alpha)} \left(\frac{\lambda_{dt}}{\lambda_{dt}} \right)^2 \right]^{\frac{1}{1-\eta}}$$
(5)

- A productivity effect which directs innovation to the relatively more advanced sector
- A price effect directing innovation towards the more backward sector

$$\frac{H_{dt}}{H_{ct}} = \left[\left(\frac{A_{dt-1}}{A_{dt-1}} \right)^{1-\phi} \left(\frac{p_{dt}}{p_{ct}} \right)^{1/(1-\alpha)} \left(\frac{L_{dt}}{L_{ct}} \right) \left(\frac{\nu_{dt}}{\nu_{ct}^s} \right) \left(\frac{\nu_{dt}}{\nu_{ct}} \right)^{1/(1-\alpha)} \left(\frac{\lambda_{dt}}{\lambda_{dt}} \right)^2 \right]^{\frac{1}{1-\eta}}$$
(5)

- A productivity effect which directs innovation to the relatively more advanced sector
- A price effect directing innovation towards the more backward sector
- A market size effect incentivising innovation in the largest market for machines

$$\frac{H_{dt}}{H_{ct}} = \left[\left(\frac{A_{dt-1}}{A_{dt-1}} \right)^{1-\phi} \left(\frac{p_{dt}}{p_{ct}} \right)^{1/(1-\alpha)} \left(\frac{L_{dt}}{L_{ct}} \right) \left(\frac{\nu_{dt}}{\nu_{ct}^s} \right) \left(\frac{\nu_{dt}}{\nu_{ct}} \right)^{1/(1-\alpha)} \left(\frac{\lambda_{dt}}{\lambda_{dt}} \right)^2 \right]^{\frac{1}{1-\eta}}$$
(5)

- A productivity effect which directs innovation to the relatively more advanced sector
- A price effect directing innovation towards the more backward sector
- A market size effect incentivising innovation in the largest market for machines
- A financial market effect, directing innovation towards the sector with the lower cost of capital

Inefficiencies in the Decentralised Equilibrium

- Under-utilisation of machines due to monopoly pricing that can be corrected with a subsidy to machines use
- Environmental externality to the production of the dirty intermediate input that can be corrected by introducing a carbon tax on the use of this input in the production of the final good
- Knowledge externality that can be corrected by a research subsidy
- Financing costs distort choices by research and machine-producing firms; if asymmetric, they also chance the direction of research

Simulations

Calibration and Policy Experiments

- Calibration based on previous literature
- Three simulations, where the economy starts on the same balanced growth path and then it is shocked by policies (carbon tax and research subsidy)
 - 1. The baseline scenario excludes financing costs
 - 2. Constant and exogenous financing costs,

$$\nu_{dt} = \lambda_{dt} \nu_{dt}^s = 100\% \qquad \nu_{ct} = \lambda_{ct} \nu_{ct}^s = 90\%$$

3. Endogenous financing costs, inversely related to the market share of each technology,

$$\nu_{jt} = \lambda_{jt} \nu_{jt}^{s} = \left(\frac{Y_{jt}}{Y_{dt} + Y_{ct}}\right)^{\omega}$$
(6)

Optimal Paths to 2°C



Optimal Paths to 2°*C*



Conclusions

Conclusions

- Access to finance is more difficult for green technologies than for incumbent and widely-known technologies
- Heterogeneous financing costs can be a threat to the transition to a decarbonised economy as they stifle innovation in the green sector
- If they endogenously depend on the market share of the technology, climate policies becomes relatively more effective in incentivising the green transition
- In reality, certain frictions may not be solved by carbon tax and research subsidies, but may need policies directed to ease the green financial gap, e.g. state investment banks and credit guarantees

Thank you for your attention!





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Calibration

Description	Parameter	Value
Annual discount rate	ρ	4%
Relative risk aversion	σ	1.5
Elasticity of substitution	ϵ	3
Machines share	α	1/3
Number of workers	L	1
Initial global GDP	Y ₀	US\$100 trillion
Initial clean energy share	$Y_{c0}/(Y_{d0}+Y_{c0})$	20%
Number of scientists	Н	1
Scientist efficiency	γ	0.02
Returns in research	η	0.7
Cross-sector spillovers	$\dot{\phi}$	0.933
2020 carbon emissions (GtCO2)	Y _{d0}	35
Emission Intensity	κ	1
Cumulative emissions limit (GtCO2)		1150
Exogenous credit constraints	ν _{c0} , ν _{d0}	90%, 100%
Endogenous credit constraints	ω	0.0655

Optimal Paths to $2^{\circ}C$



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Optimal Paths to 2°C



Optimal Paths to 2°*C*



Fixed Carbon Tax

- Second-best scenario where a carbon tax is introduced to respect a carbon budget of 1150GtCO2 and kept fixed (as a share of output)
- Under the baseline scenario, the required starting tax in the first period (2025) is \$278/*tCO*2, which then grows with output by assumption.
- Constant clean finance constraints imply a higher tax of \$431 is required to meet the climate target, an increase of \$69.
- If financing constraints are endogenous, with clean costs reducing and dirty costs increasing under the clean transition, the tax required to meet the degree target is only \$13 higher than the baseline