

Simulating the End of Work and Money: A Microsimulation of AGI-Driven Automation

Fabien Furfaro*

2025

Abstract

The development of Artificial General Intelligence (AGI) raises questions about the future of labor and monetary systems. This paper models firm-level automation decisions as a non-cooperative game, structurally analogous to the Prisoner’s Dilemma. Simulations show that automation levels depend critically on competitive pressure (γ) and implementation costs (k), with high γ/k ratios leading to near-complete labor substitution. These results suggest that policy tools—such as automation taxation or universal basic income funded by AI rents—could mitigate adverse economic outcomes. The study underscores that AGI’s impact on work and money is not technologically predetermined but shaped by institutional choices.

1 Introduction

1.1 Context: AGI and the Future of Labor

Lorem ipsum dolor sit amet, consectetuer adipiscing elit. Ut purus elit, vestibulum ut, placerat ac, adipiscing vitae, felis. Curabitur dictum gravida mauris. Nam arcu libero, nonummy eget, consectetuer id, vulputate a, magna. Donec vehicula augue eu neque. Pellentesque habitant morbi tristique senectus et netus et malesuada fames ac turpis egestas. Mauris ut leo. Cras viverra metus rhoncus sem. Nulla et lectus vestibulum urna fringilla ultrices. Phasellus eu tellus sit amet tortor gravida placerat. Integer sapien est, iaculis in, pretium quis, viverra ac, nunc. Praesent eget sem vel leo ultrices bibendum. Aenean faucibus. Morbi dolor nulla, malesuada eu, pulvinar at, mollis ac, nulla. Curabitur auctor semper nulla. Donec varius orci eget risus. Duis nibh mi, congue eu, accumsan eleifend, sagittis quis, diam. Duis eget orci sit amet orci dignissim rutrum.

1.2 Research Problem and Contribution

Our study addresses the following question: *Under what conditions do competitive dynamics between firms lead to excessive automation, and how do these outcomes challenge traditional economic equilibria?* We contribute to the literature by:

- Formalizing firm-level automation decisions as a **non-cooperative game** with a Prisoner’s Dilemma structure.
- Simulating the impact of **competitive pressure (γ)** and **automation costs (k)** on long-term equilibrium outcomes.
- Discussing policy implications, including **automation taxation** and **universal basic income (UBI)** funded by AI rents.

1.3 Outline

Section ?? presents the theoretical framework. Section ?? describes the simulation methodology and results. Section ?? discusses the implications for labor markets and policy design. Section ?? concludes and outlines future research directions.

*fabien.furfaro@gmail.com

2 Theoretical Framework: Why a Non-Cooperative Game?

2.1 Limitations of Existing Approaches

Nam dui ligula, fringilla a, euismod sodales, sollicitudin vel, wisi. Morbi auctor lorem non justo. Nam lacus libero, pretium at, lobortis vitae, ultricies et, tellus. Donec aliquet, tortor sed accumsan bibendum, erat ligula aliquet magna, vitae ornare odio metus a mi. Morbi ac orci et nisl hendrerit mollis. Suspendisse ut massa. Cras nec ante. Pellentesque a nulla. Cum sociis natoque penatibus et magnis dis parturient montes, nascetur ridiculus mus. Aliquam tincidunt urna. Nulla ullamcorper vestibulum turpis. Pellentesque cursus luctus mauris.

Table 1: Comparison of Modeling Approaches for AGI-Driven Automation

Approach	Strengths	Limitations	Relevance
CES Production Functions	Simple, macro-level insights	No strategic firm behavior	Low
Network Models	Captures interdependencies	High complexity, data-intensive	Medium
Real Options	Handles uncertainty and irreversibility	Static competitive environment	Medium
Our Game-Theoretic Model	Firm-level strategies, clear policy levers	Homogeneous firms, static	High

2.2 Profit Function and Cournot-Nash Equilibrium

The profit function for firm i is:

$$\Pi_i(a_i, \bar{a}_{-i}) = \gamma a_i(1 - \bar{a}_{-i}) + \beta a_i - k a_i^2,$$

where γ is the competitive advantage, k the cost of automation, and β the baseline gain. The symmetric Cournot-Nash equilibrium is:

$$\bar{a}^* = \min\left(1, \frac{\gamma + \beta}{2k + \gamma}\right).$$

For $\gamma = 2$, $k = 0.5$, and $\beta = 1$, the equilibrium $\bar{a}^* = 0.8$ implies that firms automate 80% of tasks, even if social costs are high.

2.3 Assumptions and Scope

We assume homogeneous firms to isolate the effect of γ and k , but discuss extensions with heterogeneity in Section ??.

3 Model and Simulation Results

3.1 Methodology

Firms update their automation levels a_i via:

- Imitation:** Adopt the level of a more profitable firm with probability $p_{\text{adopt}} = \frac{1}{1+e^{-(\Pi_j - \Pi_i)}}$.
- Mutation:** Random perturbation with 5% probability.

Table 2: Simulation Parameters and Economic Interpretation

Parameter	Values	Interpretation
γ	0.5, 1.0, 2.0, 3.0	Competitive advantage (low to high)
k	0.2, 0.8, 1.4, 2.0	Automation cost (low to high)
β	1.0	Baseline automation gain

3.2 Results

Figure ?? shows final automation levels \bar{a}^* for varying γ and k . Figure ?? compares simulation and analytical convergence.

Table 3: Equilibrium Automation Levels \bar{a}^* for Varying γ and k

$\gamma \downarrow, k \rightarrow$	0.2	0.8	1.4
2.0			
0.5	0.53	0.21	0.15
0.12			
1.0	0.76	0.45	0.30
0.23			
2.0	0.96	0.83	0.63
0.50			
3.0	1.00	0.88	0.70
0.57			

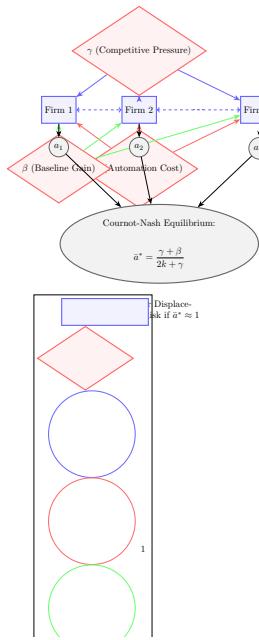


Figure 1: Conceptual diagram of the model: firms, competitive pressure (γ), automation costs (k), and equilibrium \bar{a}^* .

4 Implications: Toward a Post-Labor Economy?

4.1 Key Scenarios

- **Full automation** ($\gamma = 3, k = 0.2$): $\bar{a}^* \approx 1 \rightarrow$ Risk of demand collapse.
- **Partial automation** ($\gamma = 1, k = 1.4$): $\bar{a}^* \approx 0.3 \rightarrow$ Human labor persists.
- **Unstable equilibrium** ($\gamma = 2, k = 0.8$): $\bar{a}^* \approx 0.8 \rightarrow$ Sensitive to shocks.

4.2 Policy Responses

4.3 Limitations and Extensions

- **Assumptions:** Homogeneous firms, fixed β .
- **Future work:** Heterogeneity, macro feedback, dynamic $k(t)$.

5 Conclusion

Our study shows that:

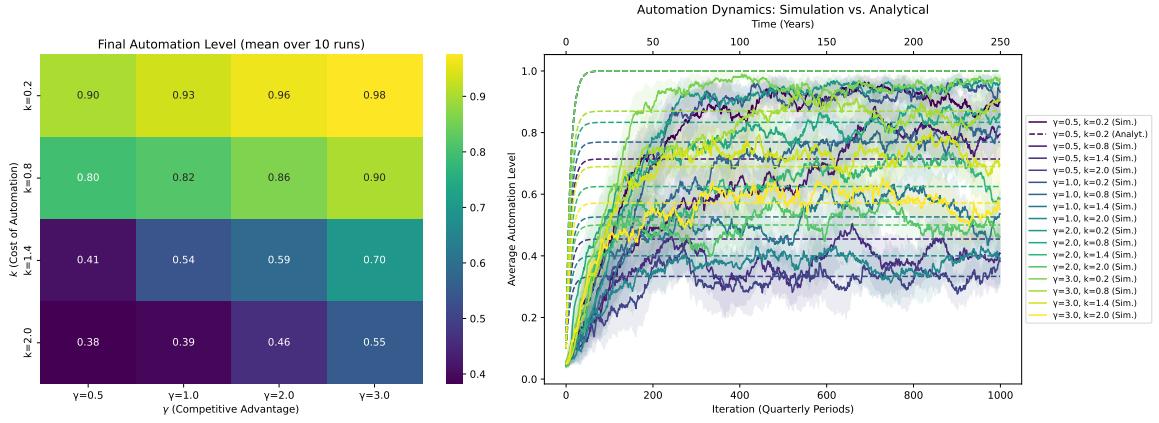


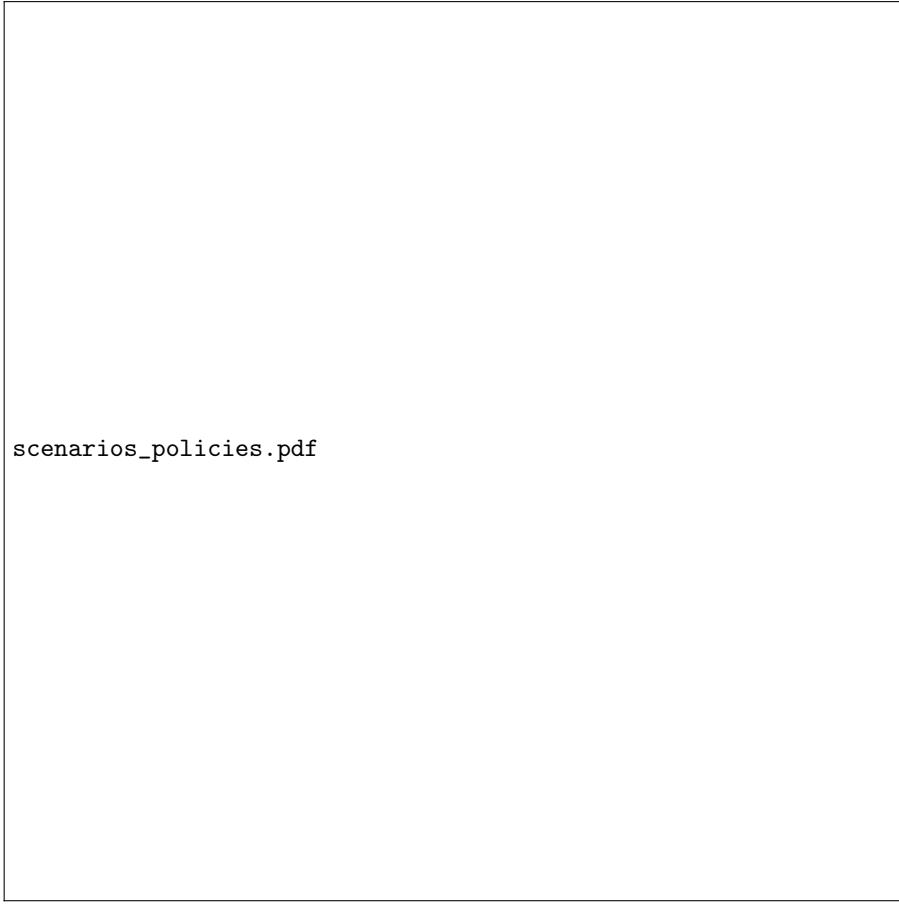
Figure 2: (Left) Heatmap of final automation levels \bar{a}^* . (Right) Dynamics of average automation (simulation vs. analytical).

- AGI-driven automation is **not inevitable** but depends on γ and k .
- Competitive dynamics can lead to **excessive automation**, even if socially suboptimal.
- Policies like **automation taxation** or **UBI** could mitigate risks.

Future research should:

1. Calibrate γ and k using sectoral data.
2. Integrate macroeconomic feedback loops.
3. Test policy interventions via simulation.

References



scenarios_policies.pdf

Figure 3: Three automation scenarios and associated policy responses.