

TOKEN TAXES: MITIGATING AGI’S ECONOMIC RISKS

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ABSTRACT

The development of AGI threatens to erode government tax bases, lower living standards, and disempower citizens—risks that make the 40-year stagnation of wages during the first industrial revolution look mild in comparison. While AI safety research has focused primarily on capability risks, comparatively little work has studied how to mitigate the economic risks of AGI. In this paper, we argue that the economic risks posed by a post-AGI world can be effectively mitigated by token taxes: usage-based surcharges on model inference applied at the point of sale. We situate token taxes within previous proposals for robot taxes and identify two key advantages: they are enforceable through existing compute governance infrastructure, and they capture value where AI is used rather than where models are hosted. We then present a research roadmap. For enforcement, we outline a staged audit pipeline — black-box token verification, norm-based tax rates, and white-box audits. For impact, we highlight the need for agent-based modeling of token taxes’ economic effects. Finally, we discuss alternative approaches including FLOP taxes, and how to prevent AI superpowers vetoing such measures.

1 INTRODUCTION

Throughout history, technological revolutions have displaced workers and ushered in periods of socioeconomic destabilization Allen (2009a). During the first industrial revolution, a period now referred to as *Engel’s Pause* (1790-1830) saw real wages stagnate as GDP per worker increased rapidly Allen (2009b); Frey (2019; 2025). Engel’s Pause triggered a downturn in living standards, as children worked in factories to support their parents and diseases such as cholera saw a substantial uptick in cases Allen (2009b).

Early evidence of AI’s impact on the labour market is already emerging in studies which show unemployment in AI-exposed early career roles rising by 16% Brynjolfsson et al. (2025). The development of AGI threatens a much greater level of destabilization Korinek & Stiglitz (2018). Whereas previous technologies automated narrow tasks, AGI will develop capabilities across all economically relevant tasks, allowing AI to replace humans in every industry Kulveit et al. (2025). This relative decline in the importance of human labour could ultimately engender a gradual disempowerment of human beings, as governments no longer rely on humans for tax revenues Kulveit et al. (2025).

This paper argues in favour of prioritizing the *token tax* as a first-line of defence against the economic risks of a post-AGI world. Token taxes are unique in two key respects. First, they are (1) enforceable: existing compute governance infrastructure enables token taxes to be audited and enforced using black-box or white-box methods Sastry et al. (2024). Second, they are (2) usage-based rather than firm-based. This enables taxation at the point of sale, mitigating inequality by capturing value where the model is used rather than where it is hosted.

We discuss the economic risks of a post-AGI world and situate the token tax within previous proposals for robot taxes. We then present a research roadmap recommending a staged auditing pipeline - black-box token verification, norm-based tax rates and white-box auditing. For impact, we also recommend agent-based modeling of token tax policies. Finally, we discuss alternative viewpoints including the FLOP (floating-point-operation) tax and how to prevent AI superpowers from vetoing token taxes.

2 BACKGROUND: THE ECONOMIC RISKS OF A POST-AGI WORLD

Thus far AI safety has focused heavily on the risks of increasing AI capabilities Bostrom (2014). Risks pertaining to the future of work have been less prominent but may be even more urgent than capability risks Acemoglu & Restrepo (2019). We discuss three key economic risks of a post-AGI world, motivating our argument in favour of promoting token taxes as an AGI governance instrument.

Government fiscal crises The first economic risk posed by a post-AGI world is fiscal crisis resulting from lost government revenues. Early studies have shown that high exposure to AI in companies increases unemployment risks while reducing labour demand Hampole et al. (2025); Frank et al. (2025). Recent work has also found that as automation increases, government fiscal revenues decrease, because taxes on labour are the highest source of government revenue Casas & Torres (2024). High unemployment in a post-AGI world would therefore reduce revenue and increase states’ fiscal costs, eroding the two main sources of public finance Korinek & Lockwood (2025).

Gradual disempowerment of citizens If AGI displaces humans by out-competing them in nearly all economic and societal functions, humans will gradually lose control over the levers of state power. Currently, governments and economic systems respond to human action since humans are the main drivers of economic and social progress Giddens (1984). If AI begins to displace humans as the main driver of economic growth, governments will become far less responsive to citizens’ needs, since they will no longer need their support Kulveit et al. (2025).

A similar phenomenon occurs in so-called *rentier states* Neal (2019). States such as Venezuela, Saudi Arabia and Oman have abundant resources and earn most of their income from oil rents as opposed to their citizens’ labour. Yet most of their citizens live in abject poverty. This is referred to as the *resource curse* Acemoglu et al. (2002; 2005). A core reason for the resource curse is the fact that states no longer have an incentive to look after their own people. In the context of AGI-driven unemployment, states will make most of their revenue from taxing AI capital and not workers and hence lose their incentive to tend to citizens’ needs.

Global inequality A post-AGI world will exacerbate global inequality. Absent new forms of taxation, AGI’s economic benefits are likely to be unevenly distributed, worsening existing divides between rich and poor countries. This is obvious when analysing the global distribution of compute. The advanced compute chips necessary to train and run frontier AI models are highly concentrated in a handful of economically developed nations. These nations, referred to as the *Compute North* in recent work, have priority access to and jurisdictional authority over advanced GPUs, while countries in the *Compute South* must rely on renting these chips from the Compute North Lehdonvirta et al. (2024). These countries will therefore benefit greatly from their ability to determine tax rates on these resources, while countries outside of the Compute North will be left behind. The global distribution of frontier model companies follows a similar pattern, with the vast majority located in the US and China.

3 ROBOT TAXES

Previous work has argued in favor of *robot taxes* as a solution to the increasing economic inefficiencies created by technology and automation Abbott & Bogenschneider (2018); Mazur (2018); Cabrales et al. (2020). Proponents argue that labour tax rates are much higher than capital tax rates because governments seek to incentivize innovation. Yet this system ceases to be efficient when capital **becomes** the labour. In the case of AI, foundation models are treated as capital even though they carry out tasks usually performed by human labour Acemoglu et al. (2020). A robot tax seeks to solve this problem by restoring *tax neutrality* to the system, removing incentives for companies to choose AI workers over human workers Abbott & Bogenschneider (2018); Mirrlees et al. (2011).

Prior suggestions for robot taxes include raising the corporate tax rate, levying an “automation tax”, and disallowing corporate tax deductions whenever a company invests in automation Abbott & Bogenschneider (2018). As AGI displaces human labour, the question of optimal robot taxation becomes central. We make the argument for why token taxes are the most promising AGI governance instrument for mitigating the economic risks of a post-AGI world.

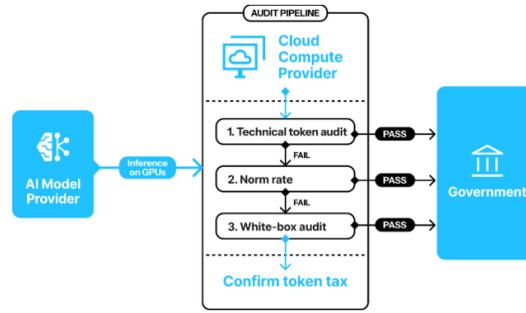


Figure 1: **Token tax collection procedure.** The cloud compute provider acts as an intermediary between the AI model provider and the government. It runs inference on behalf of the AI model provider, collects billed tokens, and determines tax liability through a staged audit pipeline. The provider first attempts (1) a black-box technical token audit. If this fails, it applies (2) a norm-based tax rate derived from average token usage. If the norm-based tax approach fails, it escalates to (3) a white-box audit. Once an audit pathway succeeds, the compute provider confirms the total token tax and reports it to the government.

4 THE TOKEN TAX

Definition: A token tax is a usage-based surcharge applied to model tokens at the point of sale.

Implementation: Drawing on existing usage-based billing procedures for language models, the token tax would be levied as a percentage markup on the provider’s billed token cost paid directly to the government. For instance, if the token tax is 10% and the cost-per-token for a given model is \$1, the AI company will pay \$0.10 in token tax. Similar to arguments made in favour of robot taxes, token taxes would enable governments to align the token tax rate with income tax, restoring tax neutrality and efficiency to the post-AGI economic system Mazur (2018); Abbott & Bogenschneider (2018).

4.1 ADVANTAGES:

We argue that there are two main advantages of using a token tax over other forms of robot tax. First, token taxes can be enforced by governments using: (1) black-box token audits (2) norm-based taxes and (3) white-box audits. To facilitate effective enforcement, we argue that compute providers should act as intermediaries between the government and AI model providers to apply a staged audit pipeline of (1), (2), and (3) (see Figure 1). Second, we claim that token taxes mitigate against the worsening of global inequality. By collecting the token tax where tokens are utilized rather than where the model is hosted, token taxes ensure that AI consumer countries (such as those in the Compute South) also benefit from AI tax receipts.

4.2 AUDITING PIPELINE

Currently, only model providers have full access to the generative process undertaken when a user calls an API, while auditors only have access to the outputs. This enables providers to inflate their profits by *strategically under-reporting* the number of tokens used Velasco et al. (2025); Wang et al. (2025); Sun et al. (2025). To properly enforce token taxes, we argue that compute providers should be legally required to implement the following 3-stage audit pipeline:

Stage 1: Black-box token audits Compute providers serve as *record keepers*, *verifiers*, and *enforcers* and are therefore able to ensure AI companies comply with obligations imposed on them by governments Heim et al. (2024). Delegating the task of auditing token count to compute providers can therefore solve the misreporting problem since providers can act as independent verifiers of compliance. Cloud hyperscalers already collect metadata on the compute consumed and the type of workloads (inference or pre-training) submitted to their servers Heim et al. (2024). By mandating that providers also collect token-level usage data, cloud providers can carry out oversight functions on behalf of the government. This would enable tax authorities to cross-check reported token usage

162 against independent logs, barring AI labs from misreporting their usage to maximize profits.
 163 *Stage 2: Norm-based taxes* The second audit stage draws on a common solution to tax evasion in
 164 rentier states such as Norway: *norm taxes* Morgan & Robinson (1976); Takle (2021). The idea of a
 165 norm tax is simple: if a company decides to inflate its profits by *strategically under-reporting* token
 166 counts, auditors can refer to norm-based tax rates for each model category and charge the company
 167 according to the norm rate. This would require a methodology for continuously updating the aver-
 168 age number of tokens used in every inference run for each type of model. It would limit the amount
 169 that companies can inflate their profits by ensuring that a company must pay a flat rate based on
 170 empirical evidence. It also has the added benefit of only requiring black-box access to models.
 171 *Stage 3: White-box audits* The third audit stage would require companies to share information about
 172 the generative process with third-party auditors. If companies are required to share this by law, they
 173 will no longer be able to game the number of tokens they report since auditors will have access to
 174 the information necessary to confirm the token count.

175 5 ALTERNATIVE VIEWS

176 While we argue that token taxes are a promising AGI governance instrument, we consider several
 177 alternative views to ensure balance.

178 **A1: Token taxes will disincentivize innovation** A salient counterargument to token taxes is that
 179 they will discourage innovation and incentivize leading AI firms to relocate to more favorable tax
 180 jurisdictions. Therefore, we recommend that technical governance researchers should partner with
 181 economists and *leverage Agent-Based Modeling (ABM)* to predict the impact of token taxes on mar-
 182 kets. ABMs have already been successfully applied to make predictions in economics Axtell &
 183 Farmer (2025); Vu et al. (2022), with LLM-based ABMs simulating large scale social interactions
 184 in digital environments such as Twitter and Reddit Yang et al. (2024). We recommend that the tech-
 185 nical governance research community partner with economists to estimate the economic impacts of
 186 token taxes on real-world markets under different hypothetical AGI growth scenarios. The tech-
 187 nical governance community can provide unique skills in this domain by understanding the policy
 188 significance of the work while also contributing technical expertise.

189 **A2: A FLOP tax is preferable to a token tax** A FLOP-based tax directly levied on compute rep-
 190 represents one of the most serious alternatives to a token tax. Compute-based thresholds are already
 191 applied as regulatory proxies for model capability and risk in AI legislation such as the EU AI Act
 192 and Biden’s executive order 14110 Act (2024); Joseph R. Biden, Jr. (2023). We claim that token
 193 taxes and FLOP taxes are not mutually exclusive. An optimal policy framework might combine
 194 token taxes with FLOP taxes in a hybrid approach which leverages the enforcement advantages of
 195 FLOP taxes while preserving the consumption-based and equity-promoting features of token taxes.
 196 Indeed, we welcome proposals to combine FLOP taxes with token taxes.

197 **A3: AI superpowers can veto token taxes** A further objection to token taxes is geopolitical; Since
 198 AI superpowers (the USA and China) contain the vast majority of the AI supply chain, they are able
 199 to leverage their economic and diplomatic power to veto or undermine token tax regimes proposed
 200 by smaller nations seeking to implement AI-related legislation Emery-Xu et al. (2025). Contentious
 201 negotiations surrounding Digital Services Taxes (DSTs) applied in many European countries serve
 202 as a precedent. To mitigate these trade tensions and retaliation from AI superpowers, we argue that
 203 regional agreements by coalitions of the willing could be established. The EU’s market size and
 204 regulatory capacity enabled successful implementation of the GDPR and the AI Act despite U.S.
 205 opposition Act (2024). A coordinated token tax by a coalition of the willing would be much harder
 206 to veto than unilateral national measures.

207 6 CONCLUSION

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 210 In this paper, we claim that token taxes can mitigate the economic risks of a post-AGI world. In con-
 211 trast to existing robot taxes, we argue that token taxes offer 2 unique advantages: they are enforce-
 212 able via existing compute governance infrastructure and they reduces global inequality by capturing
 213 value where AI tokens are used not where models are hosted. We encourage further research into
 214 the open technical governance research agendas we have identified.
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