## Artificial intelligence and algorithmic exclusion: A proposal to prioritize addressing data deserts and exclusion

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## The challenge

Al systems are only as good as the data that they learn from. Policymakers focused on Al fairness have proposed policies that would prevent apparent discrimination related to the quality of the inferences algorithms make about people from training data. But what if the training data are not only biased, but also incomplete?

The absence of data—whether the data are fragmented, outdated, low quality, or missing—can be just as harmful as the inclusion of overtly biased data or the use of models trained on biased training data. All systems can fail not only because they make biased predictions, but also because they make no meaningful predictions at all for certain individuals or populations. That is, unfairness may arise not only because an algorithm sees individuals incorrectly, but also because it does not see them at all when it tries to make real-time predictions.

Algorithmic exclusion formally describes failure when an Al-driven system lacks enough data on an individual to return an output about them. This phenomenon is not a side effect of poor design; rather, it is a structural feature of digital inequality. Often, the same economic and social forces that marginalize individuals offline—including lack of internet access, fewer interactions with institutions that collect data, and lower participation in activities that generate traceable data—also reduce their digital visibility.

These gaps in data, or "data deserts," are zones where AI cannot function effectively. They lead to systematic under-recognition of the very populations that an equity-focused policy would be designed to protect.

## The proposal

This proposal suggests a concrete, policy-relevant addition to regulations and proposals on Al fairness: incorporate algorithmic exclusion as a class of algorithmic harm equal in importance to bias and discrimination. When a regulation aims to prevent "bias, discrimination, and other harms," it should also specifically address the ways in which certain people are excluded from, or unable to derive benefits from, having information relating to them included in an algorithm's training data or assumptions.

Addressing algorithmic exclusion benefits not only lower-income Americans, but also anyone operating outside dominant data-generating regimes: new immigrants, survivors of domestic violence using alias names, the digitally disconnected, and others with complex data footprints. By explicitly incorporating the problem of data nonavailability and missing predictions into algorithmic auditing and regulatory frameworks, this proposal ensures that the absence of data is treated as a substantive fairness issue, rather than as a mere technical inconvenience.

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TABLE 1
Addressing algorithmic exclusion in US and EU AI laws

Aspect	US Eliminating Bias in Algorithmic Systems Act (S.3478; H.R.10092)	Tucker proposal for US	EU AI Act	Tucker proposal for EU
Definition of bias	Refers to "bias, discrimination, and other harms" from covered algorithms.	Add "algorithmic exclusion," meaning harms to those for whom there are no input data.	Focuses on unjustified detrimental treatment or exploitation of vulnerabilities.	Expand to include the idea of unjustified exclusion from algorithmic decisionmaking due to missing data and missing output.
Risk assessment tools	Requires risk assessments on bias and discrimination.	Add representation audits to identify who may not have the requisite data for the algorithm to make predictions.	Risk assessment mandatory for high-risk Al.	Include evaluation of the risks of missing data and missing output.
Audit standards	Focuses on documenting mitigation of known bias.	Add obligation to identify risks of missing data and missing output.	Mandates accuracy, robustness, and human oversight.	Integrate missing data and missing output checks to human oversight.

