ML research papers and Drawbacks

Paper	Technical Drawbacks	Non-technical Drawbacks
A Comparative Study of Statistical and Machine Learning Models on Near- Real-Time Daily CO2 Emissions Prediction	 Models may not generalize well across different regions or emission sources. Limited accuracy in real-time prediction. Data quality and availability can affect model performance. 	 Lack of policy implications or real-world applicability. Potential resistance from industries due to economic concerns. May not address public awareness or behavior change aspects.
A Machine Learning Algorithm to Explore the Drivers of Carbon Emissions in Chinese Cities	 Potential bias due to regional differences in data collection methods. Overfitting risk in machine learning models due to small or non-diverse datasets. The model may not account for all influencing factors of emissions (e.g., social, cultural). 	 Government regulations may limit access to certain datasets. Findings may not be easily interpretable for policymakers and stakeholders. Ethical concerns regarding data privacy and surveillance.
Nonparametric Approaches for Analyzing Carbon Emissions: From Statistical and Machine	Nonparametric methods can be computationally expensive for large datasets. These approaches may not always provide clear, interpretable results. The lack of assumptions in nonparametric methods may limit their applicability in some contexts.	Lack of awareness or expertise in industries to adopt these methods. Resistance to new approaches in policymaking. Limited engagement with stakeholders outside academia.

Paper	Technical Drawbacks	Non-technical Drawbacks
Optimization of Carbon Footprint Management Model of Electric Power Enterprises Based on Al Learning Perspectives	Al models require large datasets, which may not be available in some sectors. Model's results may be too theoretical and lack realworld implementation details. Possible limitations in handling dynamic and evolving data over time.	Implementation may be costly for small enterprises. Lack of government incentives for adoption. Public perception of Albased decisions may lead to skepticism.
Predicting CO2 Emission Footprint Using AI Through Machine Learning	Al-based models may struggle with extreme outliers in emissions data. High computational cost for large datasets and real-time predictions. Difficulty in incorporating external factors (e.g., economic, political changes).	Ethical concerns over Al's decision-making transparency. Potential conflicts with industries resisting Al-driven regulations. Risk of Al-generated policies lacking human-centered considerations.