How to trust AI-generated maps

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Abstract:

The increasing potential role of artificial intelligence (AI) in generating maps leads to the need for robust criteria and plausibility checks to ensure their trustworthiness. This paper develops a framework of key criteria and indicators for assessing the plausibility of maps produced by AI, as well as systematic methodologies to judge their trustworthiness. The discussion focuses on technical, contextual, and ethical dimensions, aiming to establish a comprehensive framework for evaluating AI-generated cartographic outputs, following respective rationales (Ceballos et al. 2023, Gartner 2022, Prestby 2023).

1. Technical Criteria

Technical factors serve as the foundational layer for evaluating map plausibility. The first step involves assessing the quality and provenance of the training data used to develop the AI models. High-quality training datasets that are diverse, accurate, and up-to-date enhance the model's ability to produce reliable outputs. The inclusion of metadata, such as data sources, timestamps, and geographic coverage, is critical for understanding the limitations and applicability of the generated maps.

Next, the resolution and granularity of the map outputs are evaluated. Plausible maps should offer appropriate detail for their intended purpose without oversimplifying or overloading users with unnecessary information. Additionally, the consistency of map elements, such as geographic boundaries, topographic features, and thematic overlays (e.g., population density or climate data), should align with established standards and existing authoritative maps.

The performance of the AI model is another key factor. Metrics such as accuracy and precision provide quantitative measures of the model's effectiveness. Robust validation against known benchmarks or ground truth datasets is essential to confirm that the AI system is generating plausible outputs.

2. Contextual Plausibility Checks

Beyond technical parameters, contextual plausibility indicators play a significant role in judging the reliability of AI-generated maps. One important check involves evaluating the coherence and alignment of the map with existing geographic knowledge and real-world conditions. For instance, a map that depicts unexpected anomalies, such as rivers flowing uphill or cities located in uninhabitable regions, raises red flags about its trustworthiness.

Cross-referencing with authoritative sources, such as government agencies, scientific institutions, and established cartographic databases, further enhances the reliability of contextual assessments.

Temporal consistency is another vital factor. Maps that incorporate dynamic features, such as weather patterns or traffic data, must reflect real-time updates accurately. Plausible maps should exhibit temporal alignment with the most current data available, avoiding outdated or obsolete representations. Additionally, spatial coherence—ensuring that neighboring regions on a map share logical relationships—helps maintain plausibility.

3. Ethical and Transparency Indicators

Ethical considerations and transparency are critical for fostering trust in AI-generated maps. Transparency begins with disclosing the methodologies and algorithms used to create the maps, including the AI model's architecture, data preprocessing steps, training procedures and the cartographic decisions. This information allows users to understand the capabilities and limitations of the system.

Bias detection is another crucial aspect. AI models may inadvertently perpetuate or amplify biases present in their training data, leading to maps that misrepresent certain regions or populations. Employing fairness metrics and conducting audits can mitigate these risks.

User interpretability also factors into trustworthiness. AI-generated maps should present information in a manner that is accessible and comprehensible to diverse users. Simplifying technical jargon, employing intuitive symbology, and providing clear legends enhance usability and trust. Additionally, interactive features that allow users to query the data or adjust parameters improve engagement and confidence in the map's outputs.

4. Independent Verification and Peer Review

A crucial step in judging the trustworthiness of AI-generated maps is subjecting them to independent verification and peer review. Third-party evaluations, including reviews by experts in cartography, geospatial analysis, and domain-specific fields, can uncover errors or biases that might have been overlooked during the development process. Publicly available datasets and open-source algorithms further facilitate collaborative validation and foster accountability.

Simulation-based testing can also serve as a verification mechanism. For example, stress-testing the AI model with edge cases, such as rare geographic features or extreme environmental conditions, helps identify its limitations and areas for improvement. Robustness under diverse scenarios is a quality criteria of a reliable mapping system.

5. Human Oversight and Hybrid Approaches

While AI offers benefits in processing large datasets and identifying patterns, human oversight remains indispensable in ensuring the trustworthiness of generated maps. Hybrid approaches that combine AI-generated outputs with human expertise might be useful in many cases. Expert review ensures that subtle contextual nuances, ethical considerations, and domain-specific insights are incorporated into the final map.

Training end-users to critically evaluate AI-generated maps is equally important. Equipping users with tools and knowledge to perform basic plausibility checks, such as identifying anomalies, cross-referencing sources, and interpreting metadata, empowers them to make informed decisions.

Conclusion

The trustworthiness of AI-generated maps is dependent on a multidimensional evaluation framework encompassing technical, contextual, and ethical criteria. Ensuring high-quality data, robust performance metrics, and alignment with real-world conditions forms the technical foundation for plausibility. Contextual checks, such as temporal consistency and cross-referencing with authoritative sources, further validate the outputs. Ethical considerations, including transparency, bias detection, and user interpretability, foster trust and inclusivity. Independent verification, peer review, and human oversight can help to overcome errors and biases, while hybrid approaches leverage the strengths of both AI and human expertise.

By establishing rigorous plausibility indicators and checks, stakeholders can enhance the reliability and credibility of AI-generated maps, paving the way for their safe and effective use in a wide range of applications. The ongoing evolution of AI technologies demands continuous refinement of these criteria to keep pace with emerging challenges and opportunities in geospatial mapping

References

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