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    - **APA Citation:** Zhang, W., Yao, R., Du, X., Liu, Y., Wang, R., & Wang, L. (2023). *Traffic flow prediction under multiple adverse weather conditions based on self-attention mechanism and deep learning models*. Physica A: Statistical Mechanics and its Applications, 625, 128988. https://doi.org/10.1016/j.physa.2023.128988
12. **A Prediction Model of Traffic Congestion Using Weather Data (2015)**
    - **Abstract:** Weather factors such as temperature and rainfall in residential areas and tourist destinations affect traffic flow on the surrounding roads. In this study, we attempt to find new insights into the relationship between traffic congestion and weather by leveraging big data processing technology【110†L2-L4】【112†L1-L4】. *(Additional details:* The authors propose a traffic congestion prediction model for

specific roads and regions, and experimental results demonstrate its effectiveness in predicting congestion using day, time, and weather inputs. *)*

- **APA Citation:** Lee, J., Hong, B., Lee, K., & Jang, Y.-J. (2015). *A prediction model of traffic congestion using weather data*. In **2015 IEEE International Conference on Data Science and Data Intensive Systems (DSDIS)** (pp. 81–88). IEEE.

13. **The Impact of Weather Data on Traffic Flow Prediction Models (2022)**

    - **Abstract:** In this paper, we present a detailed investigation on the impact of weather data on different traffic flow prediction models. Multiple forecasting models were evaluated with and without incorporating weather features. The study found that including weather data can improve the performance of traffic flow prediction models, highlighting the importance of weather factors in traffic forecasting【118†L1-L4】.

    - **APA Citation:** Al-Selwi, H. F., Abd Aziz, A. B., Abas, F. S., Hamzah, N. A. A., & Mahmud, A. B. (2022). *The impact of weather data on traffic flow prediction models*. IAES International Journal of Artificial Intelligence, 11(4), 1223–1231. https://doi.org/10.11591/ijai.v11i4.336