

# The Explainability of Downsampling

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### Introduction

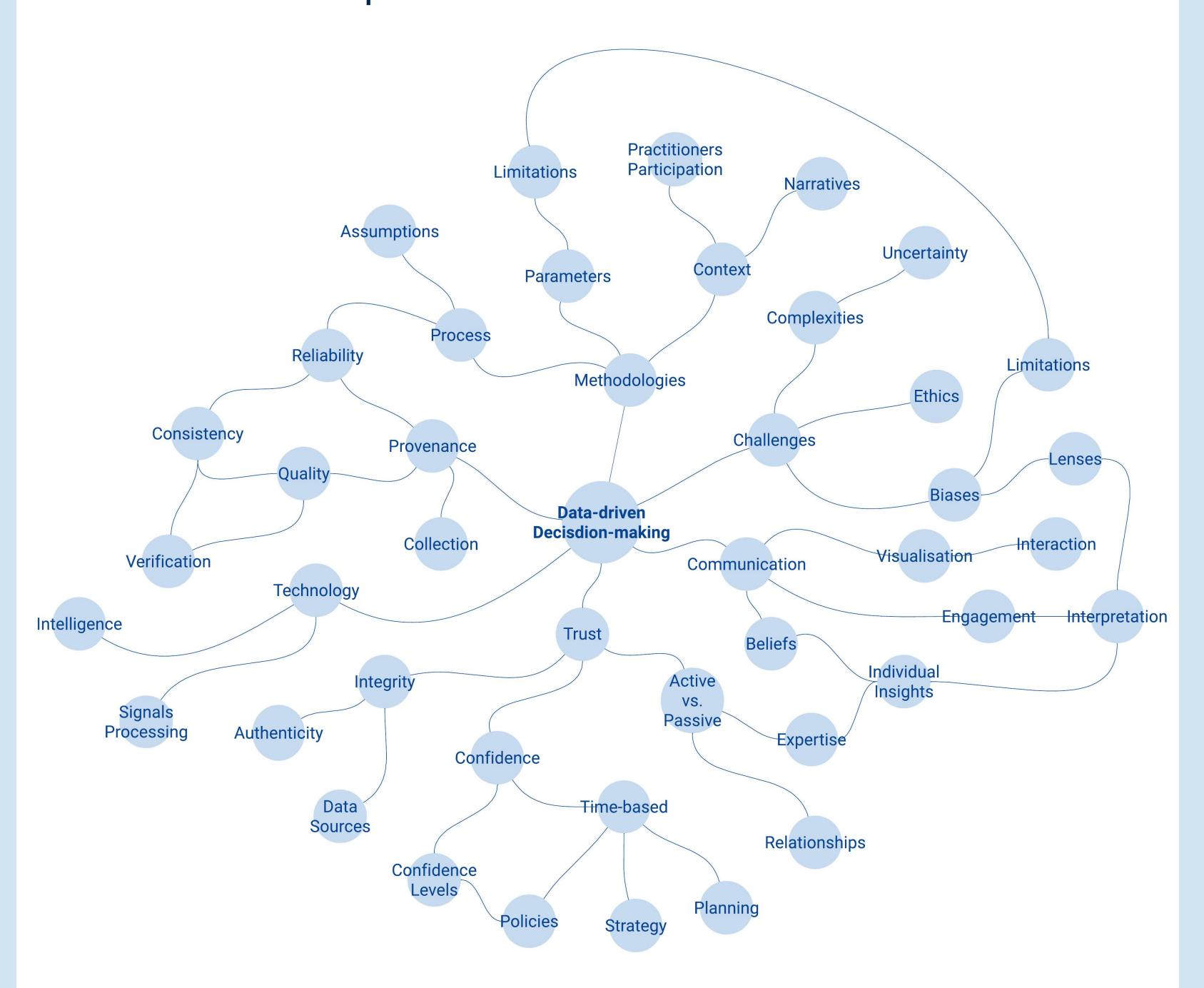
HM Government is committed to making data-driven decisions that engender public trust1. Achieving this requires transparency of how data is used in government decision-making<sup>2</sup>.

#### This research aims to:

- Highlight that government decision-makers need to trust how the data is used.
- 2 Share how transparency can be created in the data processing pipeline, focused on downsampling.

## User Research

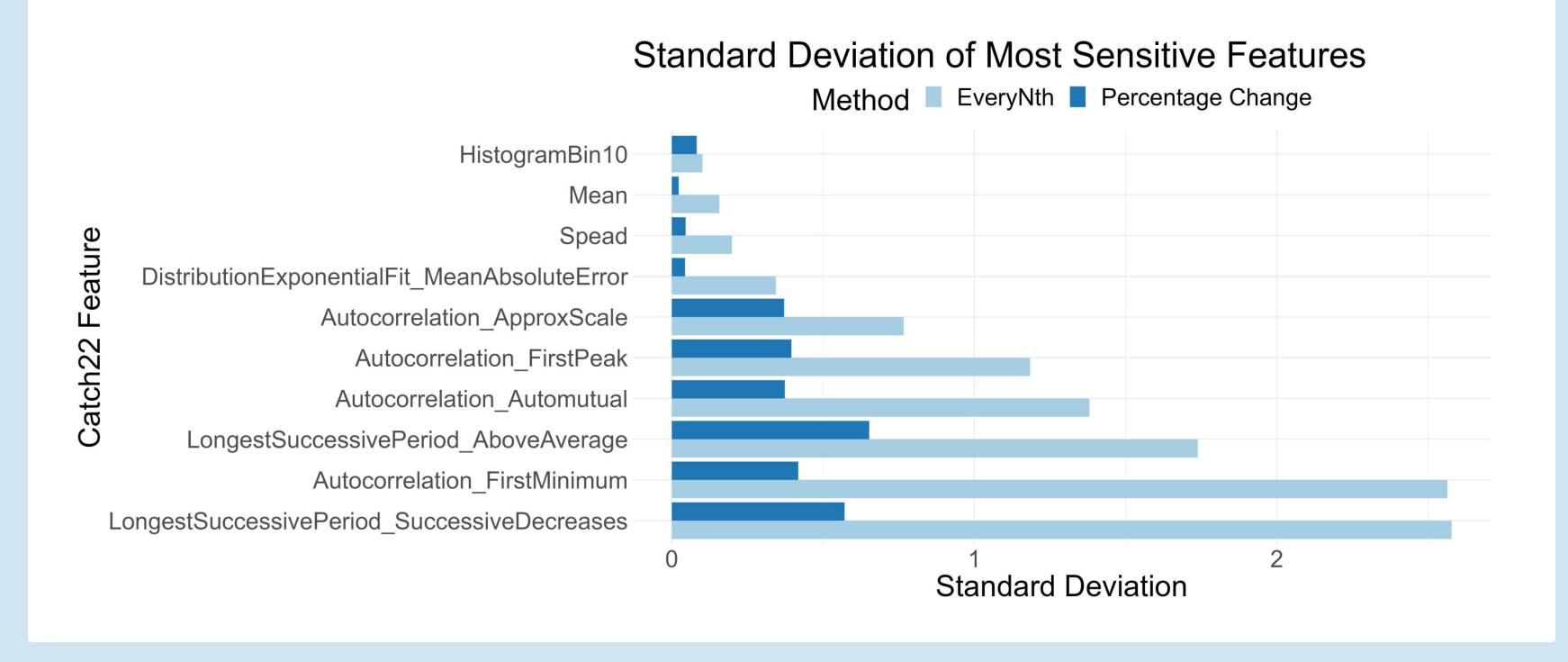
Sixteen officials were interviewed; nine decision-makers and seven data practitioners. Themes are shared below:



"[Decision-makers] often have an incorrect view of how much they can trust the data..."

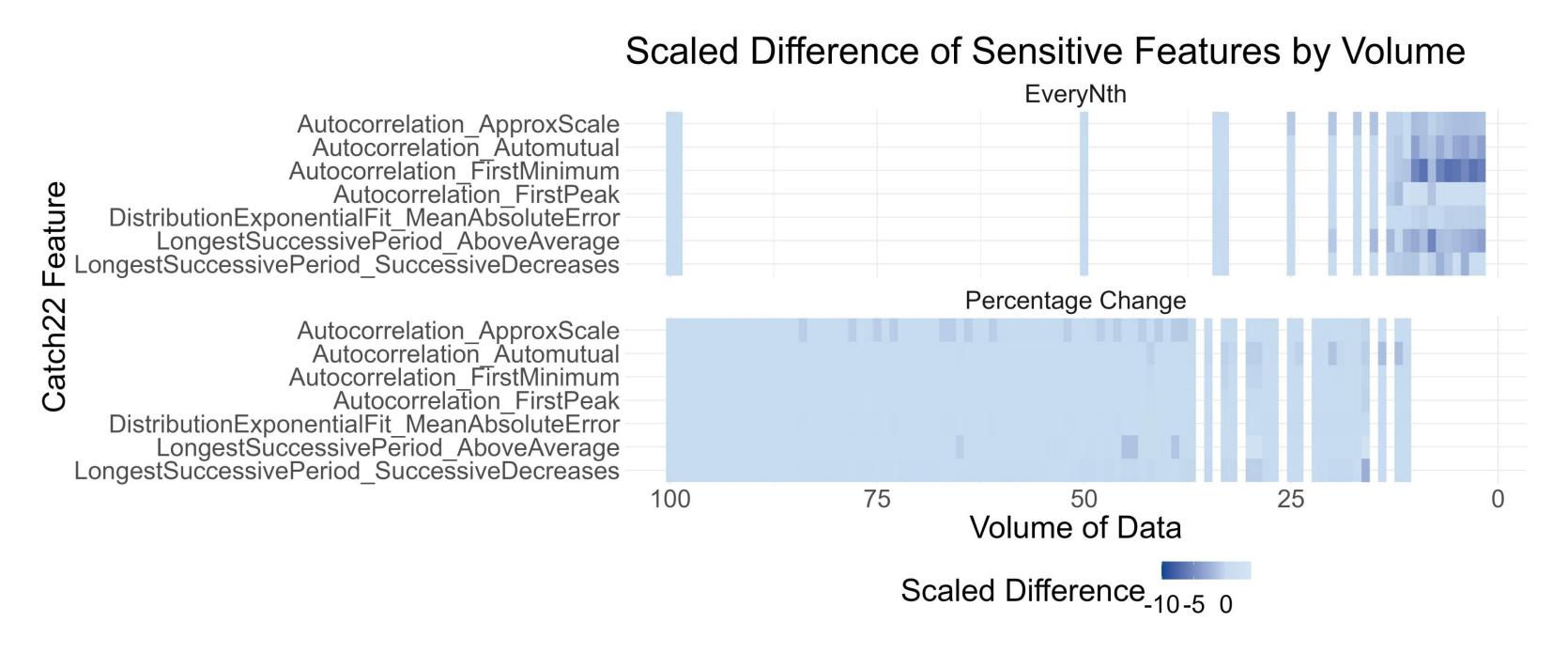
# Downsampling Sensitivity

The time series features most sensitive to downsampling were identified by calculating the standard deviation of difference (scaled) between the original and processed data.



# Downsampling Impact

The EveryNth and Percentage Change downsampling algorithms were applied to the most sensitive features.



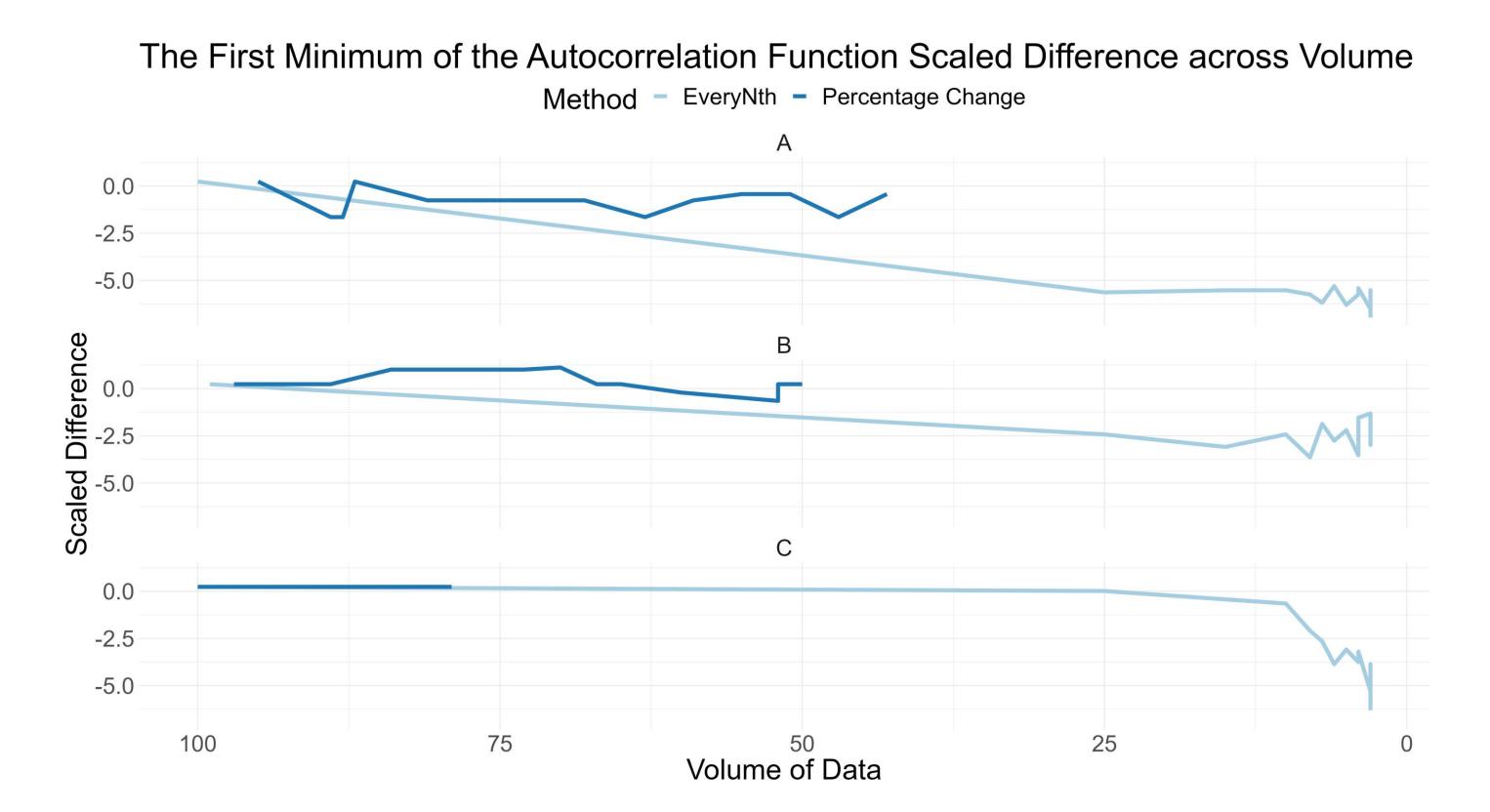
The impact of downsampling changes as more data is discarded; this is acute for EveryNth and subtle, but observably inconsistent for Percentage Change.

#### References

- Cabinet Office, "Government transformation strategy: Better use of data." HM Government, 2017;
- 2. Department for Digital, Culture, Media & Sport, "National data strategy." HM Government, 2020; 3. C. H. Lubba, B. Fulcher, T. Henderspn, B. Harris, O. r. TL, and O. Cliff, "catch22: CAnonical timeseries CHaracteristics." R Journal, 2022.

# Feature Variation

The impact of the downsampling algorithms on the most sensitive features varies across different time series.



This difference suggests that it is possible to select the features that best preserve each time series.

### Discussion

- User research suggests that transparency of how data is used, and its limitations, engenders trust for decision-makers and data practitioners.
- This research identifies seven time series features that appear to be especially sensitive to the impacts of downsampling.
- These features offer a new approach for identifying and communicating the downsampling algorithms and parameters that preserve representativeness with the lowest demand on computing resources.

Future work applying this research to alternative time series is needed to further test the approach of this research. Conducting usability research with decisionmakers and data practitioners will also help refine these insights.