

MISSING PIECE PREVENTS SEEING THE BIG PICTURE

Dealing with missing values in air pollution and weather
to identify environmental risk factors associated with diabetes

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1. WHY?

Do you know that Diabetes is linked with air pollution and weather?

- There is a direct link between the global diabetes epidemic and climate change[1, 2, 3]
- Diabetes attributable to PM2.5 air pollution is significant [4, 5, 6]
- Diabetes incidence rate in the USA and prevalence of glucose intolerance worldwide increase with higher outdoor temperature [7]

Missing value problem of sensor data! 

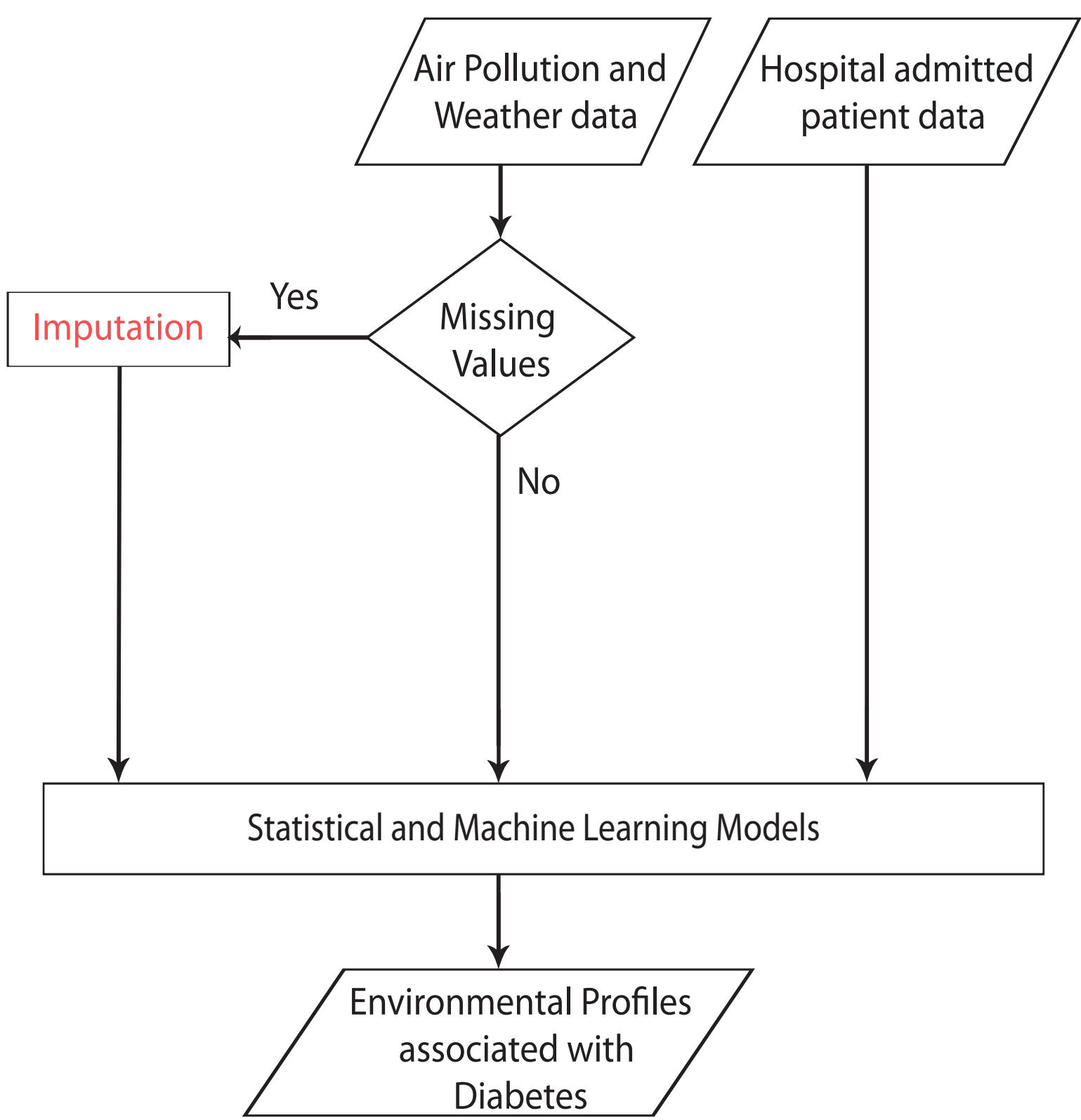
- Air pollution and weather data are collected through sensors
- It is very common to have missing values due to failures of sensors
- Large gaps with missing data is a problem where research is rarely carried out to deal with

2. WHAT?

Two-fold objectives of our research! 

- To develop a framework to deal with missing values in weather and pollution data, particularly for large gaps (Data Science Discipline)
- To identify vulnerability profiles of environmental conditions on diabetes (Application in Population Health)

3. HOW?



- Replacing a missing value by a plausible value is known as imputation

4. WHAT WE DID?

- We proposed an algorithm to impute large gaps using a bi-directional method based on regularized regression models [8,9]
- This method outperforms a set of baseline and well-established imputation methods (Figure 1)

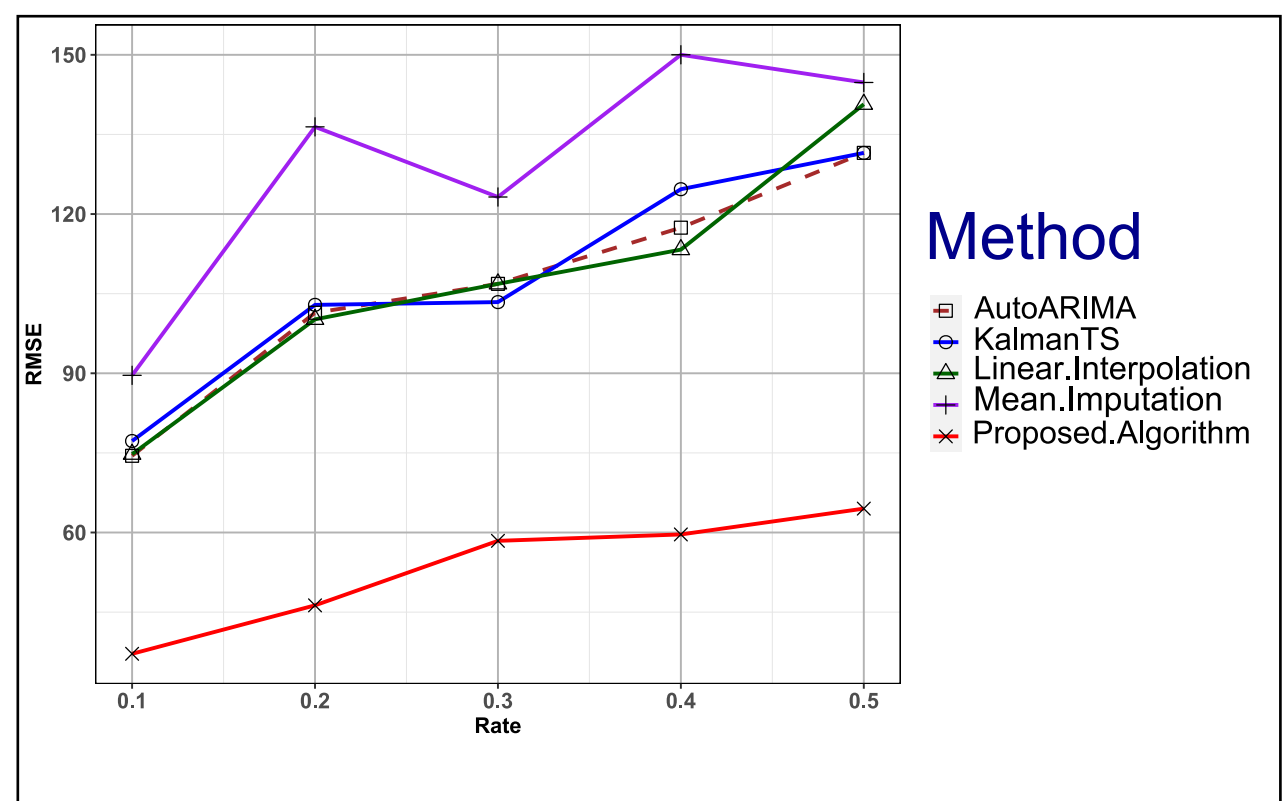
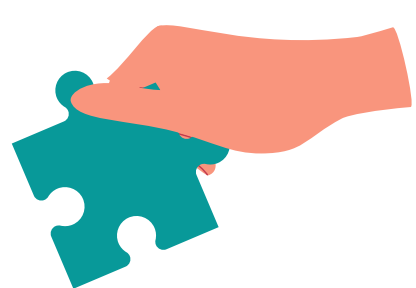


Figure 1: Root Mean Squared Error of the proposed method in comparison with a set of other methods for different proportions of missing values

- Developed a dashboard to visualize the patterns of missing values in NSW pollution and weather data (Figure 2)

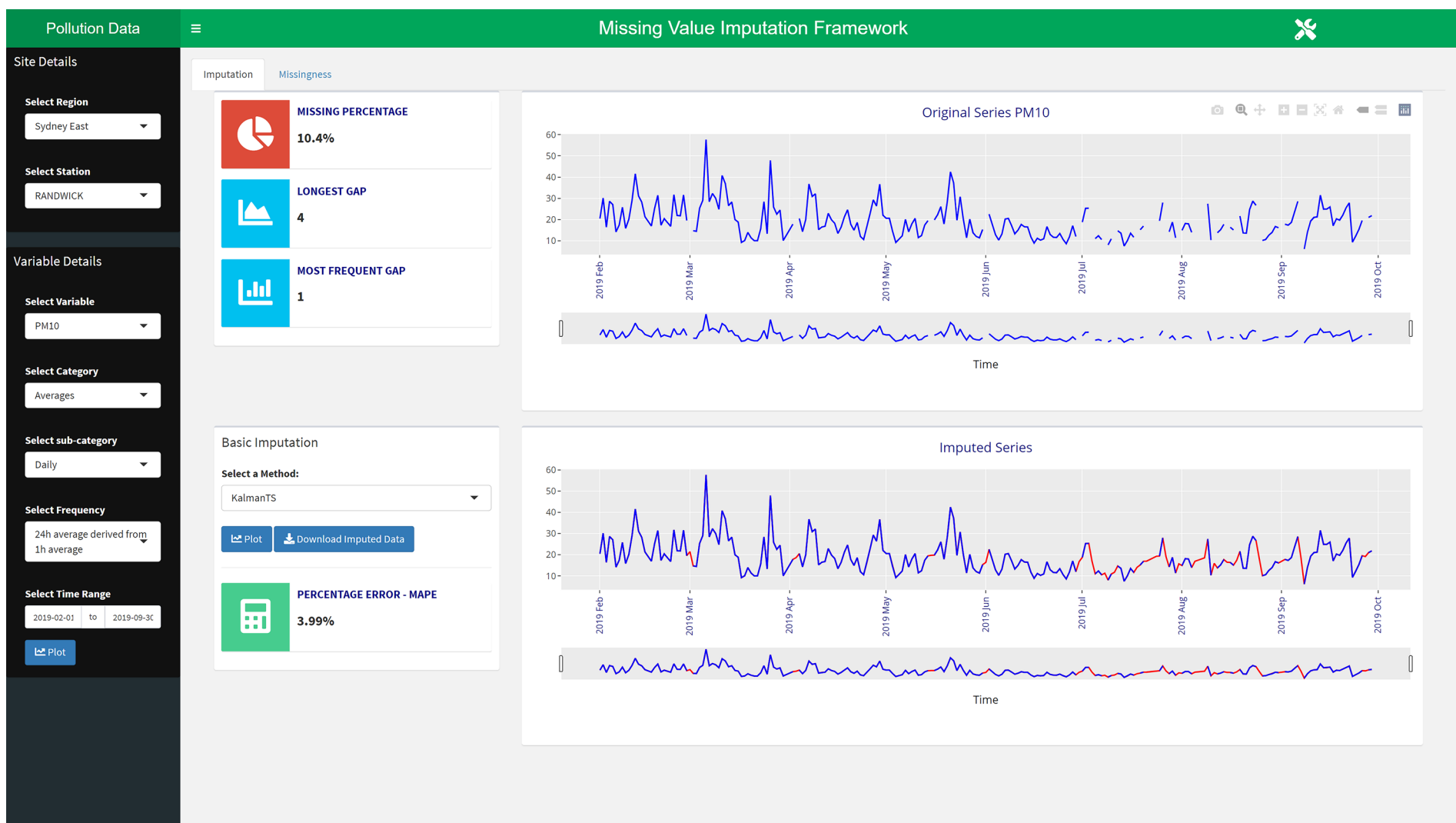


Figure 2: Dashboard to analyse missing values in NSW pollution data


- This provides facility to impute missing values using several methods and to download imputed data

5. SO WHAT?

- When using Environmental data captured by sensors having large gaps, this procedure will increase the accuracy of the results obtained through downstream analyses

- Such methodologies developed to deal with missing values could be extended to a wide range of applications influenced by environmental conditions

6. WHAT'S NEXT?

- Cleaned data will be linked with NSW Admitted Patient Data Collection from 2015 to 2020 and apply spatiotemporal modelling techniques to identify vulnerability environmental profiles for diabetes in NSW
- The findings of this research would be of great importance to policymakers to manage and mitigate the burden of diabetes 

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