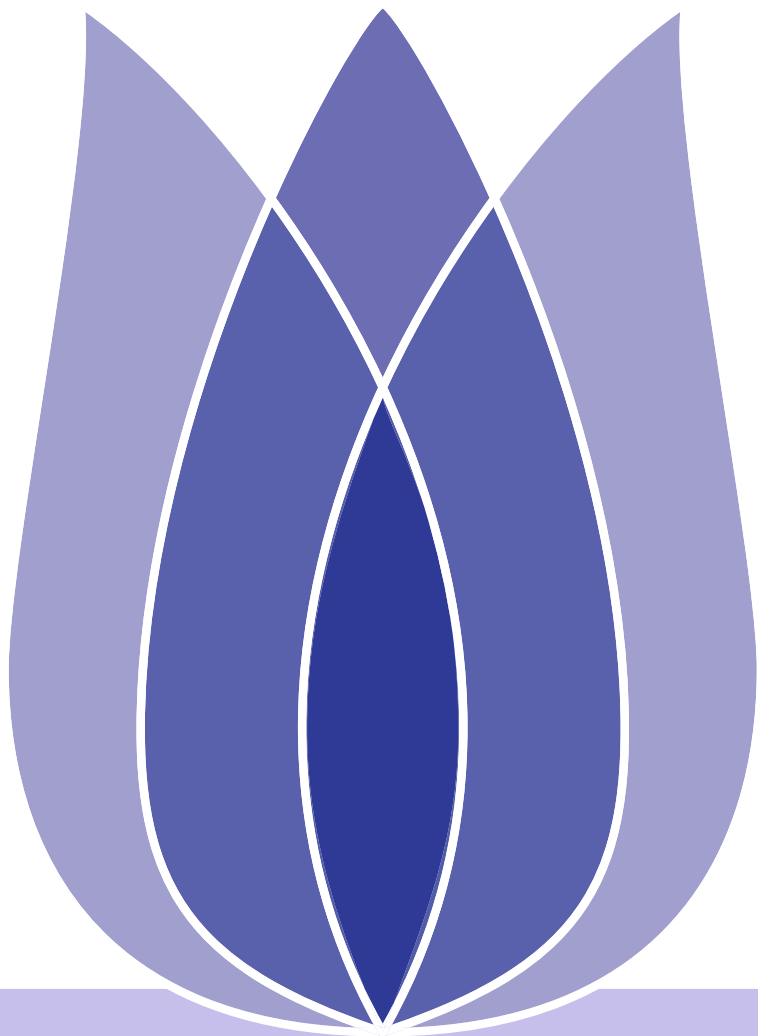


Air Pollution Prediction based on multicollinearity

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03/06/2022





Overview

- [Problem Definition](#)
- [Challenges](#)
- [Proposed Model](#)
- [Evaluation Results](#)
- [Conclusion](#)

Problem Definition

Air pollution predictors

Challenges

Challenges

Proposed Model

Step One -Preprocessing

Step Two - Model training

Step Three - Model Prediction and Evaluation

Evaluation Results

Conclusion



Problem Definition

Air pollution predictors

Challenges

Proposed Model

Evaluation Results

Conclusion

Problem Definition



Air pollution predictors

Problem Definition
Air pollution predictors
Challenges
Proposed Model
Evaluation Results
Conclusion

Defn

- Predict air pollution composition in atmosphere in the future.
- Use predictors such as Temperature, Humidity, Sensor data.
 - Response variables are carbon monoxide, benzene, notrous oxide



Linear Regression vs Rigid Regrsson

Problem Definition
Air pollution predictors
Challenges
Proposed Model
Evaluation Results
Conclusion

Linear Regression

- Linear regression presents relationship as a straight line.
- Show correlation between two variables (one predictor for response variable/variables).
- Response should be continuous and independent variable(s) (predictor variables) can be continuous or discrete.

Rigid Linear Regression

- Use to implement multicollinearity of predictor variables which are highly correlated each other **objects** in the whole dataset.
- Add penalty values to reduce the loss or error of linear regression cause by bias and/or variance of the variables.



- [Problem Definition](#)
- [Challenges](#)**
- [Challenges](#)
- [Proposed Model](#)
- [Evaluation Results](#)
- [Conclusion](#)

Challenges



- [Problem Definition](#)
- [Challenges](#)
- [Challenges](#)
- [Proposed Model](#)
- [Evaluation Results](#)
- [Conclusion](#)

Group Outlying Aspects Mining

- Focus on differences between **groups**.
- **Multiple** points.

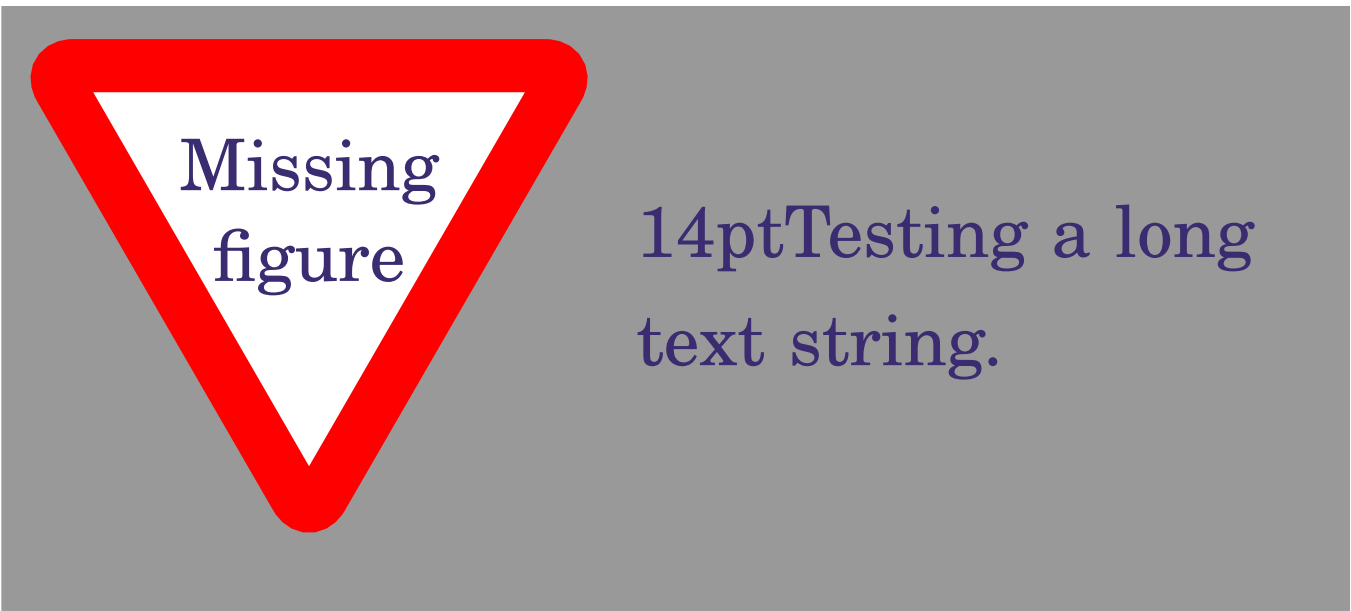


Figure 1: Group Outlying Aspects Target

Outlying Aspects Mining

- Concentrates on differences between **objects**.
- **One** point.

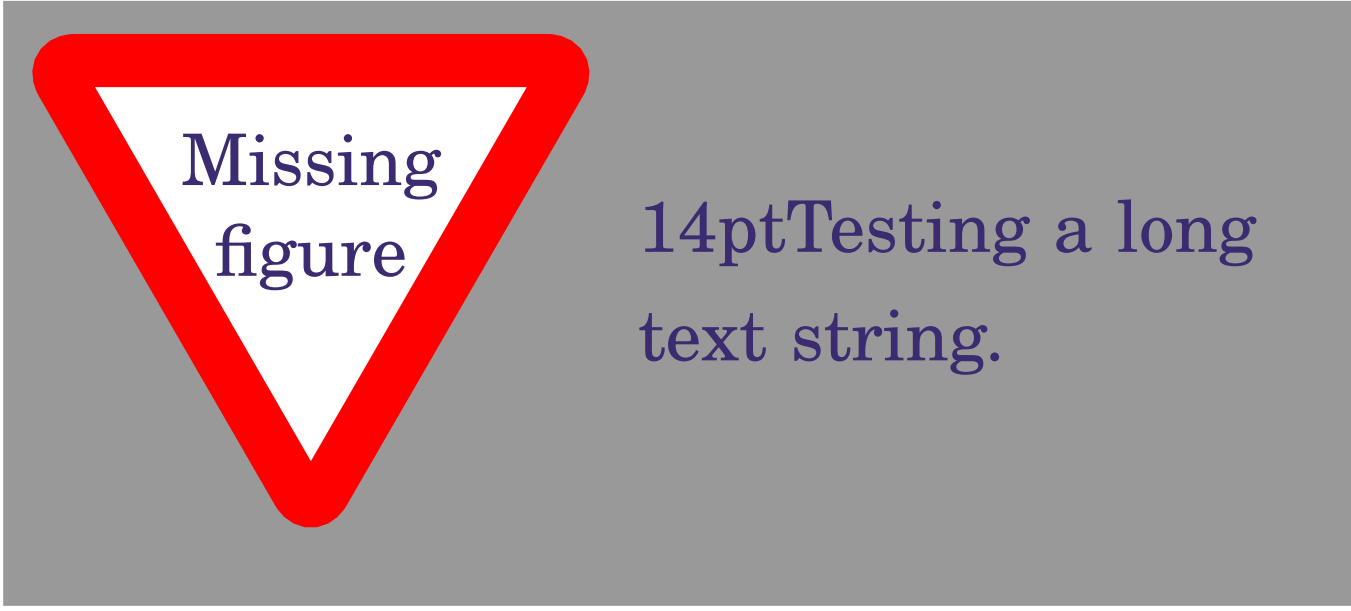


Figure 2: Outlying Aspects Target



Challenges

Problem Definition

Challenges

Challenges

Proposed Model

Evaluation Results

Conclusion

- Reasons for **multicollinearity** in predictors.
 - ◆ Inaccurate use of different types of variables.
 - ◆ Poor selection of questions or null hypothesis.
 - ◆ Variable repetition.
 - ◆ A dependent variable selection.
 - ◆ High correlation.
 - ◆ Use of dummy variables.



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[Problem Definition](#)

[Challenges](#)

[Proposed Model](#)

[Step One -Preprocessing](#)

[Step Two - Model training](#)

[Step Three - Model Prediction and Evaluation](#)

[Evaluation Results](#)

[Conclusion](#)

Proposed Model



Problem Definition
Challenges
Proposed Model
Step One -Preprocessing
Step Two - Model training
Step Three - Model Prediction and Evaluation
Evaluation Results
Conclusion

Framework of Proposed Model:

Figure 3: Framework of Proposed Model



Step One -Preprocessing

- Problem Definition
- Challenges
- Proposed Model
- Step One -Preprocessing**
- Step Two - Model training
- Step Three - Model Prediction and Evaluation
- Evaluation Results
- Conclusion

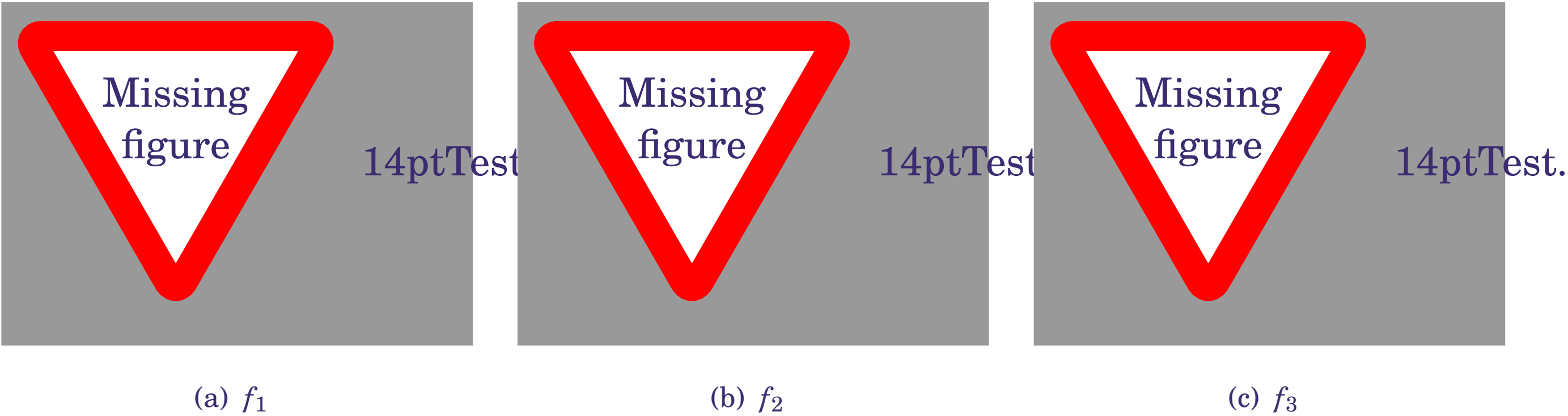


Figure 4: Histogram of G_q on three features

Step Two - Model training

Problem Definition

Challenges

Proposed Model

Step One -Preprocessing

Step Two - Model training

Step Three - Model Prediction and
Evaluation

Evaluation Results

Conclusion

- Calculate Earth Mover Distance
 - ◆ Represent one feature among different groups
 - ◆ Purpose: calculate the minimum mean distance

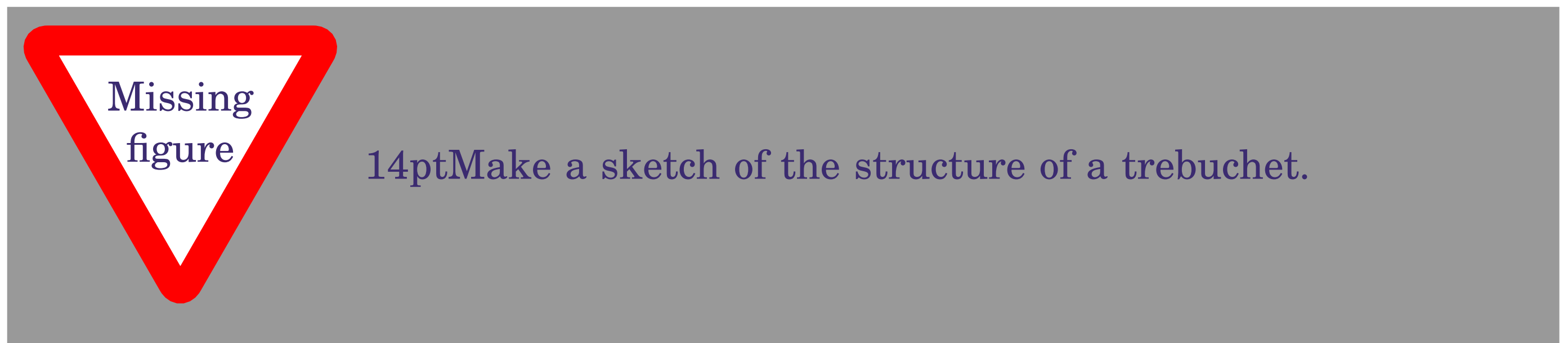


Figure 5: EMD of one feature



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Step Three - Model Prediction and Evaluation

- [Problem Definition](#)
- [Challenges](#)
- [Proposed Model](#)
- [Step One -Preprocessing](#)
- [Step Two - Model training](#)
- [Step Three - Model Prediction and Evaluation](#)
- [Evaluation Results](#)
- [Conclusion](#)

- Identify group outlying aspects mining based on the value of outlying degree.
- The greater the outlying degree is, the more likely it is group outlying aspect.



Illustration

- [Problem Definition](#)
- [Challenges](#)
- [Proposed Model](#)
- [Step One -Preprocessing](#)
- [Step Two - Model training](#)
- [Step Three - Model Prediction and Evaluation](#)
- [Evaluation Results](#)
- [Conclusion](#)



- [Problem Definition](#)
- [Challenges](#)
- [Proposed Model](#)
- [Evaluation Results](#)**
- [Conclusion](#)

Evaluation Results



- Problem Definition
- Challenges
- Proposed Model
- Evaluation Results
- Conclusion

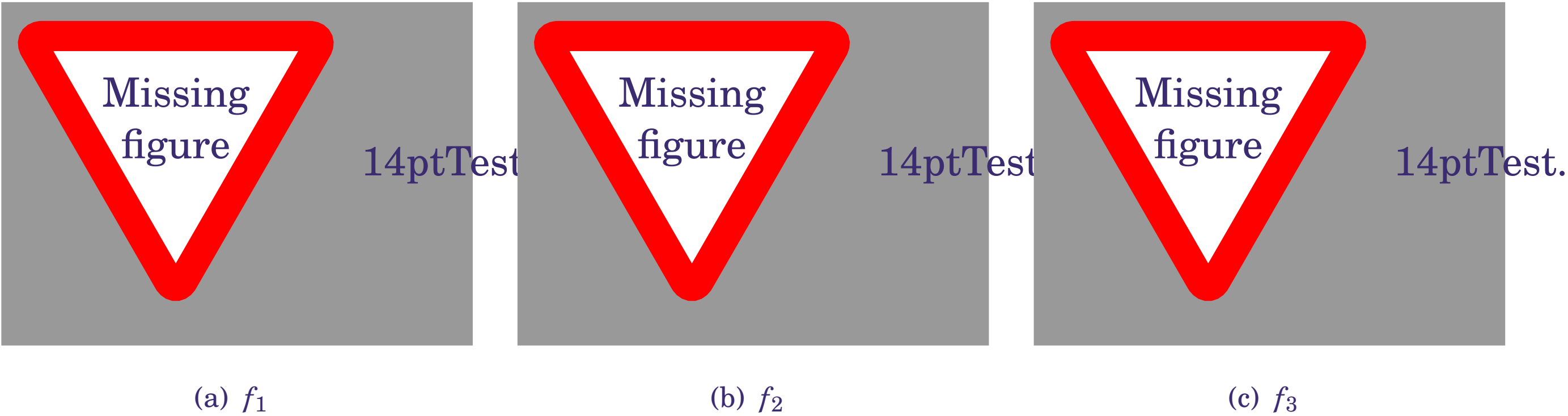


Figure 6: Histogram of G_q on three features



- [Problem Definition](#)
- [Challenges](#)
- [Proposed Model](#)
- [Evaluation Results](#)
- [Conclusion](#)

Conclusion



Conclusion

Problem Definition
Challenges
Proposed Model
Evaluation Results
Conclusion

- **Problem Definition:** we propose prediction model to predict air pollution components which are specified as carbon monoxide, benzene and nitrous oxide. Algorithm Consequently, we use rigid linear regression instead general linear regression method.
- **Strategies :**We uses multiple predictors such as temperature, absolute and relative humidity and five sensor data. We identify that these predictors are correlated among each other. Therefore, we reveals the need of handling multicollinearity. We clearly show the performance efficiency of proposed models using determination of coefficient, bias and variance scores for each models.
- **Recommendations :** Evaluate how well these prediction models behave on adding more noisy data. How to affect prediction accuracy by noisy data. Increase the amount of predictors to predict air pollution



Questions?

- [Problem Definition](#)
- [Challenges](#)
- [Proposed Model](#)
- [Evaluation Results](#)
- [Conclusion](#)



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