# Mars Express Power Challenge

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Group 6 BAE Systems

## **About Mars Express Orbiter (MEX)**

- A spacecraft launched by the European Space Agency (ESA) in 2003.
- Provides invaluable scientific data.
- Contains an autonomous thermal system for regulating temperature.
- Uses electric power from solar panels (or batteries during eclipses)
  - Science Power = Produced Power Platform Power Thermal Power
- Predicting power consumption is crucial for optimising space operations and ensuring the longevity of MEX.

#### About the Mars Express Challenge

- MEX Flight Control Team from ESA collected three Martian years' (2061 Earth days, from 22/08/2008 to 14/04/2014) worth of MEX's telemetry data.
- Data was then released as part of a machine learning challenge in 2016.
- Develop predictive models that could predict the average electric current of 33 power lines per hour for the fourth Martian year (687 Earth days, from 14/04/2014 to 01/03/2016).

#### **Objectives**

- Develop a machine learning model to predict the average hourly electric current for each of the 33 thermal power lines on MEX.
- Evaluate the model's performance using Root Mean Squared Error (RMSE).
- Present our findings to Mr Gordon and the UoM team.

#### **About the Data**

- Dataset included four Martian years of data, categorised into two main parts:
  - Context data (independent variables):
    - SAAF
    - FTL
    - DMOP
    - EVTF
    - LTDATA
  - Observation data (dependent variable):
    - Hourly electric current measurements for each of the 33 power lines over the first three Martian years.
- The observation data for the fourth Martian year was never released.
  - Our project uses 20% of Year 1-3 data as unseen test data.

# **Project Timeline**

Task	12 Feb - 18 Feb	19 Feb - 25 Feb	26 Feb - 3 Mar	4 Mar - 10 Mar	11 Mar - 17 Mar	8 Apr - 14 Apr	15 Apr - 21 Apr	22 Apr - 28 Apr	29 Apr - 5 May	6 May - End
Background research										
Reading accompanying literature										
Reading existing solution papers										
Sharing and collaborating research										
EDA and pre-processing										
Data cleaning										
EDA										
Feature engineering										
Resampling, interpolation and merging										
Model development and testing										
Discussion on approach										
Identifying important features										
Building model 1 (extra trees)										
Building model 2 (XGBoost)										
Hyperparameter tuning										
Model ensembling										
Writing the report										
Planning										
Creating visualisations										
Introduction										
Pre-processing section										
Machine learning section										
Results section										
Conclusion										
Creating the presentation										
Creating slides										
Script writing and rehearsal										

#### SAAF (Solar Aspect Angles) Files

- Provide information about the spacecraft's orientation with respect to the Sun:
  - ut\_ms: Unix timestamp in milliseconds
  - sa: Solar aspect angle (angle between the spacecraft's solar panels and the Sun-MEX line)
  - sx, sy, sz: Solar angles of the X, Y, and Z axes of the satellite, respectively

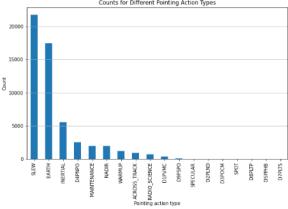
## DMOP (Detailed Mission Operations Plan) Files

- Contain commands sent to various subsystems of the spacecraft:
  - ut\_ms: Unix timestamp in milliseconds
  - subsystem: Name of the operated subsystem command

ut_ms	subsystem
1.21936E+12	AXXX301A
1.21936E+12	AAAAF20C1
1.21936E+12	AAAAF57A1
1.21936E+12	AAAAF23G1
1.21936E+12	AAAAF60A1
1.21937E+12	AXXX305A
1.21937E+12	AXXX380A
1.21937E+12	ASEQ4200
1.21937E+12	ATTTF301E
1.21937E+12	ATTTF310A
1.21937E+12	APSF01A2
1.21937E+12	APSF02A1
1.21937E+12	APSF89A1

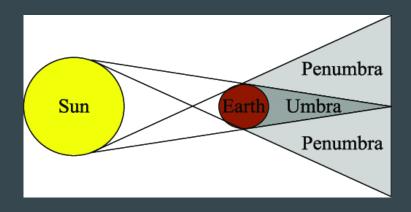
## FTL (Flight Dynamics Timeline) Files

- Contain spacecraft pointing events:
  - utb\_ms: Unix timestamp in milliseconds of the beginning of the pointing period
  - ute\_ms: Unix timestamp in milliseconds of the end of the pointing period
  - type: Type of pointing or action
  - flagcomms: Boolean indicating if any communication device was used



#### **EVTF (Event Files)**

- (Event) files, which contain descriptions of other events such as umbras, penumbras, or changes in altitude:
  - ut\_ms: Unix timestamp in milliseconds
  - description: Short description of the event



```
ut_ms description

1.21936E+12 MLG_LOS_05_/_RTLT_02373

1.21937E+12 NNO_AOS_00_/_RTLT_02373

1.21937E+12 NNO_AOS_05_/_RTLT_02373

1.21937E+12 NNO_AOS_05_/_RTLT_02373

1.21937E+12 NNO_AOS_10_/_RTLT_02374

1.21937E+12 4000_KM_DESCEND

1.21937E+12 MRB_/_RANGE_06000KM_START

1.21937E+12 OCC_MARS_200KM_START_/_RA_181.68_/_DE_-00.08_/_OMP_(296.35,-46.48)_/_SZA_077

1.21937E+12 OCC_MARS_START_/_RA_181.69_/_DE_-00.08_/_OMP_(299.32,-43.44)_/_SZA_076

1.21937E+12 MRB_AOS_00

1.21937E+12 1200_KM_DESCEND
```

#### LTDATA (Long-Term Data) Files

- Provide daily measurements of various parameters related to the spacecraft's environment:
  - ut\_ms: Unix timestamp in milliseconds (one sample per day)
  - sunmars\_km: Distance between the Sun and Mars in kilometers
  - earthmars\_km: Distance between Earth and Mars in kilometers
  - sunmarsearthangle\_deg: Sun-Mars-Earth angle in degrees
  - solarconstantmars: Solar constant at Mars in W/m<sup>2</sup>
  - eclipseduration\_min: Total duration of all eclipses in the day, in minutes
  - occultationduration\_min: Total duration of all occultations in the day, in minutes

## Feature Engineering

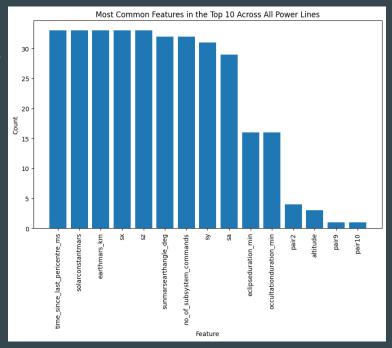
Context File	Features Selected	Comment		
SAAF	sa, sx, sy, sz	No additional feature engineering was necessary.		
DMOP	command, ON/OFF pairs, no_of_subsystem_commands	Various engineering attempts made. Referred to an existing solution on Github for ON/OFF pairs.		
FTL	flagcomms, pointing action types	Dummy variables created for different pointing action type.		
EVTF	altitude, direction, umbra, penumbra, time_since_last_pericentre_ms	Converting strings into various binary and integer features.		
LTDATA	earthmars_km, sunmarsearthangle_deg, solarconstantmars, eclipseduration_min, occultationduration_min	No additional feature engineering was necessary.		

#### Data Resampling and Merging

- Context files initially contained varying time granularities and date ranges.
- Solution: Resampling all files to a consistent hourly time granularity.
  - Initially considered a time granularity of 30 seconds, but found it infeasible.
  - Techniques used:
    - Linear interpolation and mean imputation for numerical features.
    - Forward or backward filling for binary features.
- The resampled files were then merged into a single dataset.

#### **Feature Selection**

- Built 33 simple Extra Trees models for each power line.
- Extracted feature importances.
- Picked the top 20 for each power line.



## **Predicting the Thermal Power Consumption**

- Models:
  - Extra Trees:
    - Well-suited for large, high-dimensional datasets.
    - Effective for feature importance identification.
    - Reduces overfitting through random thresholds.
  - XGBoost:
    - Handles large datasets efficiently.
    - Delivers high predictive performance.
    - Tackles complex relationships within the data.
- Performed hyperparameter tuning using a random search and cross-validation.
- Used ensembling to leverage the strengths of both, creating a third model.

#### Results

- Extra Trees Mean RMSE: 0.0981

- XGBoost Mean RMSE: 0.0978

- Ensemble Mean RMSE: 0.0976

## Challenges and Improvements

- Lack of official documentation

- Incorporation of lag features

- Sheer size of the dataset

#### References

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