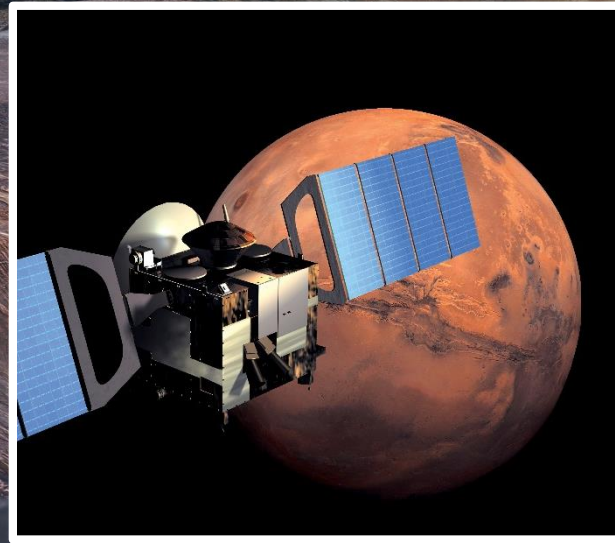
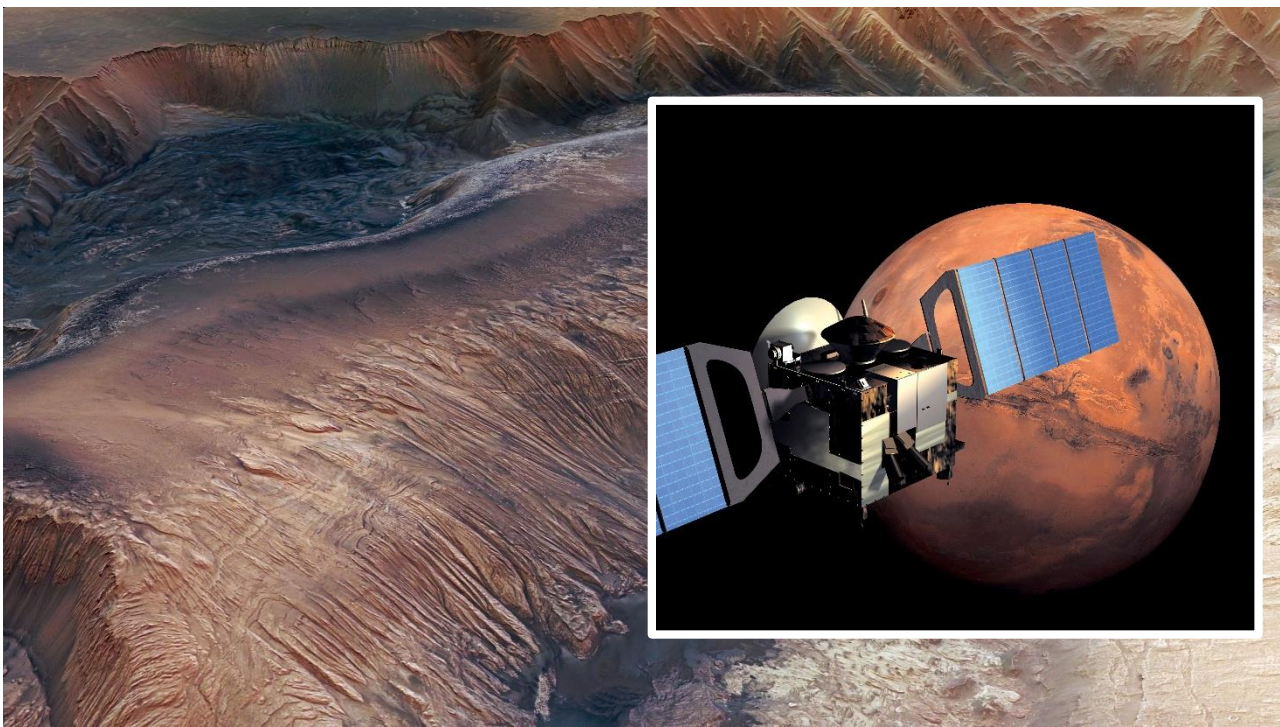
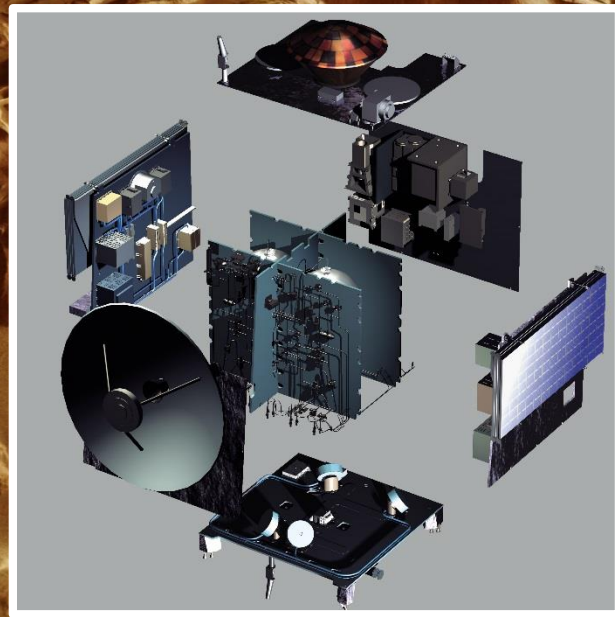
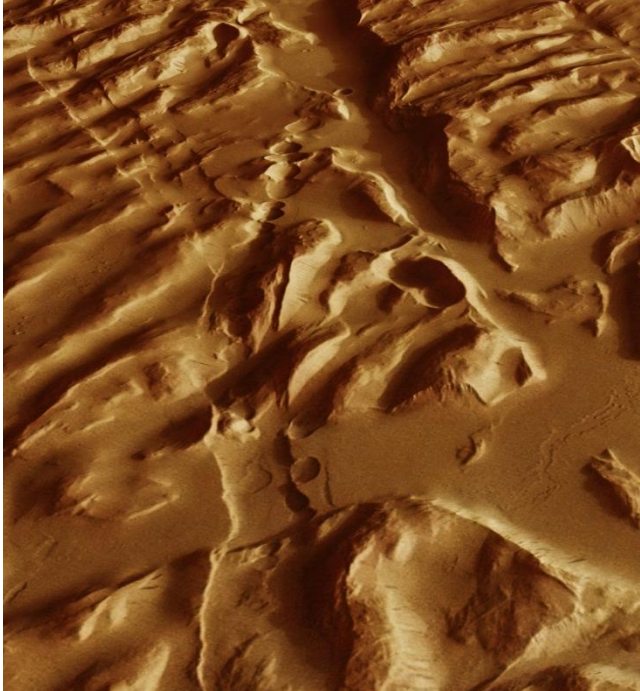
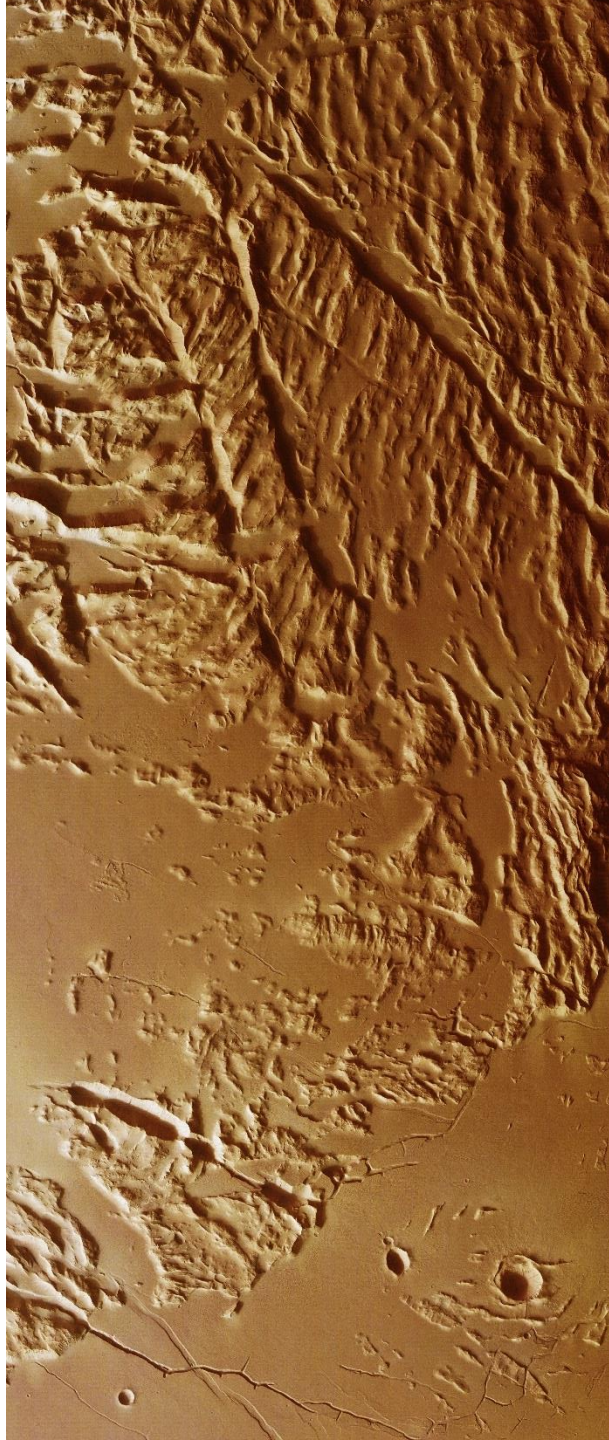
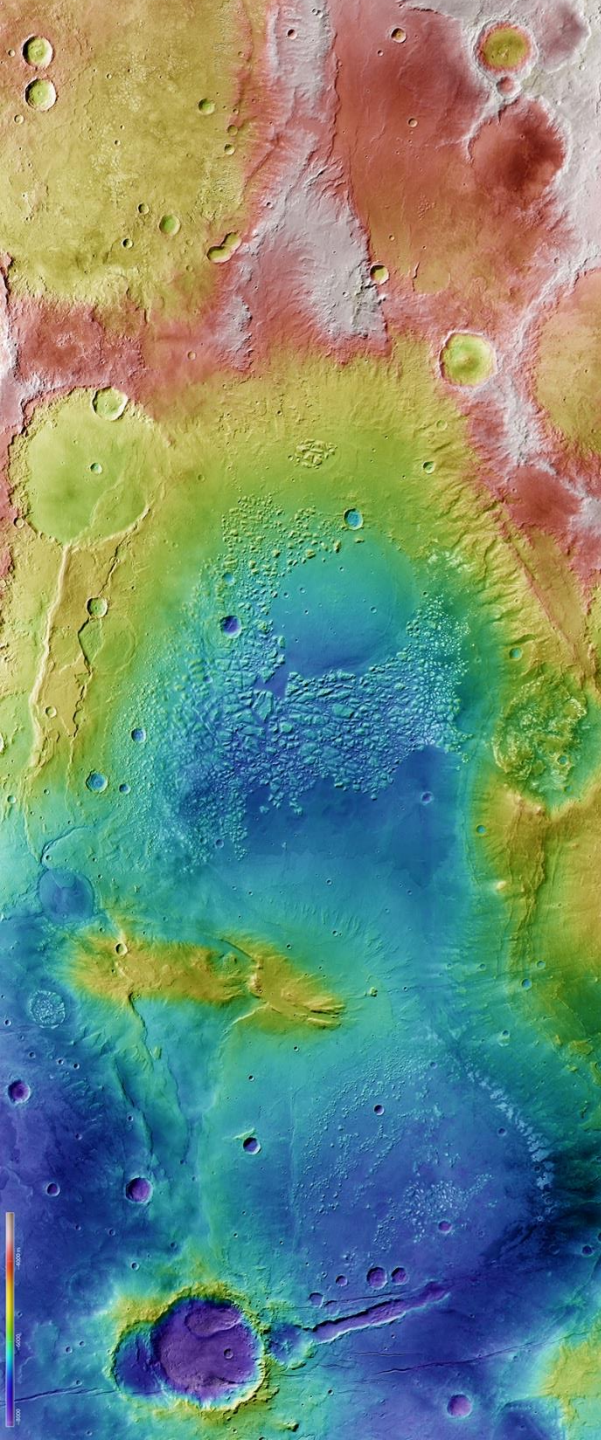


Mars Express Power Challenge

CC71Q - Introducción a la Minería de Datos

Gabriel De La Parra

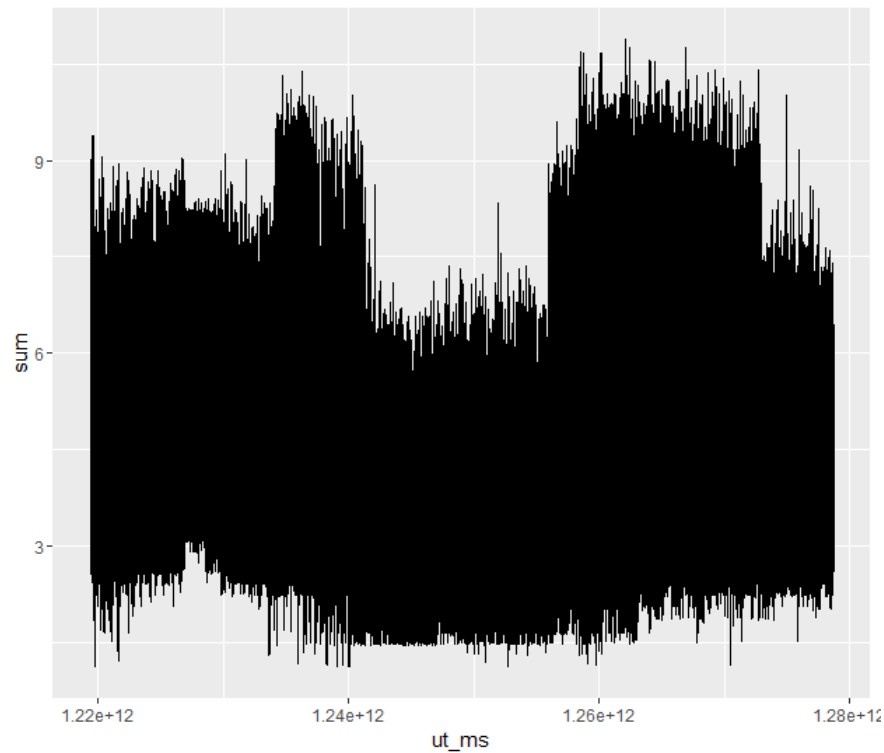
30.Mayo.2016



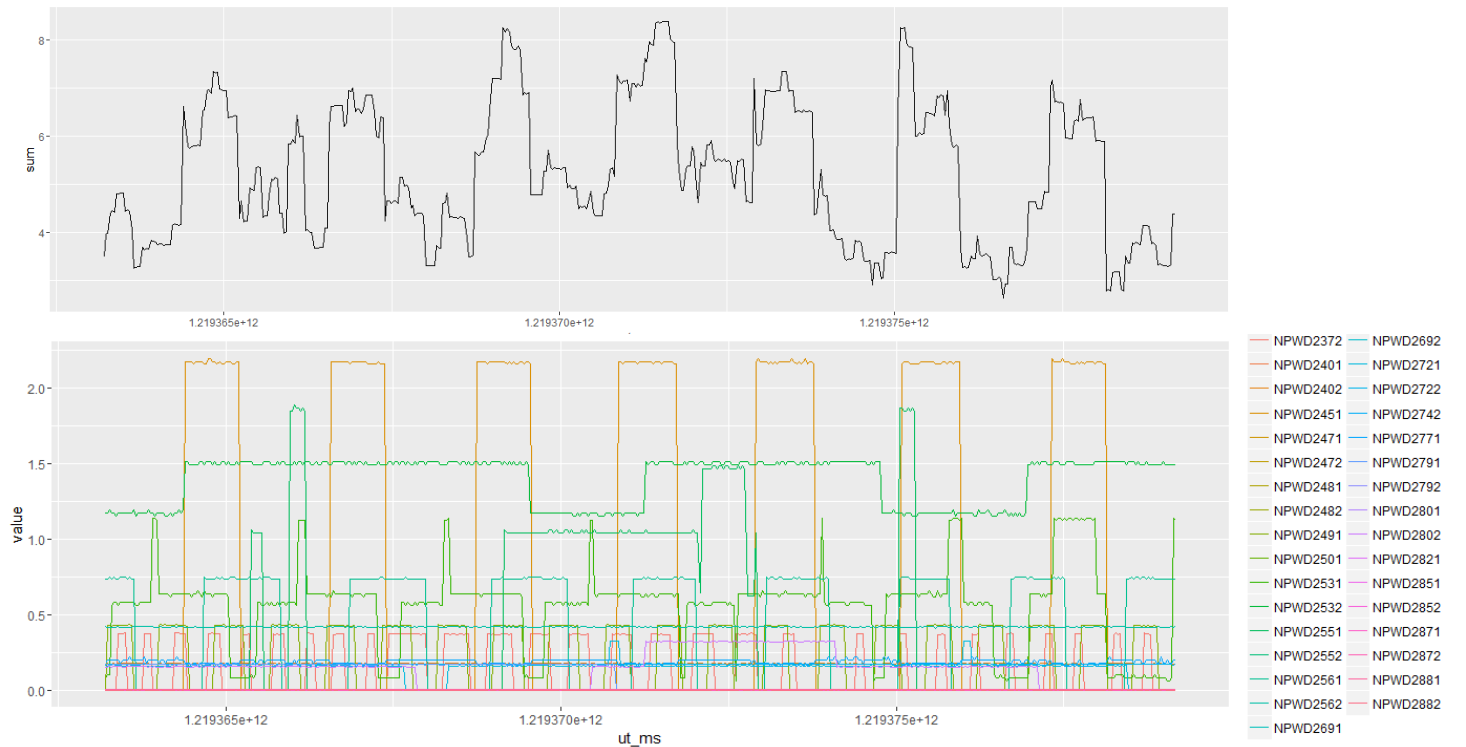
Mars Express Power Challenge

Problema: Predicción del consumo energético del satélite

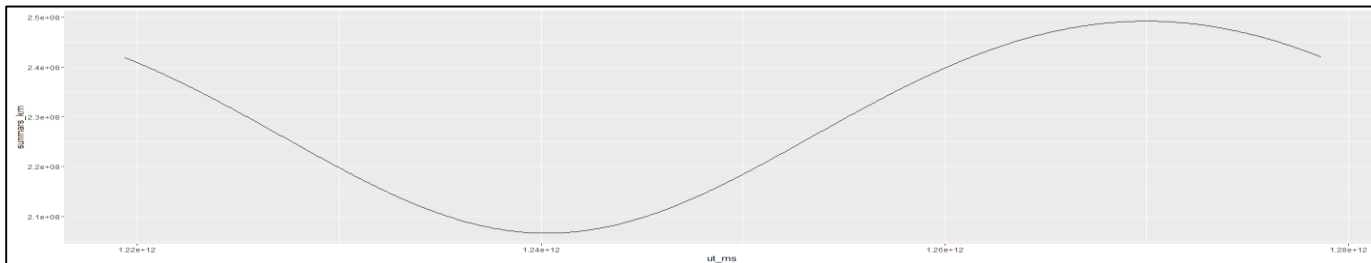
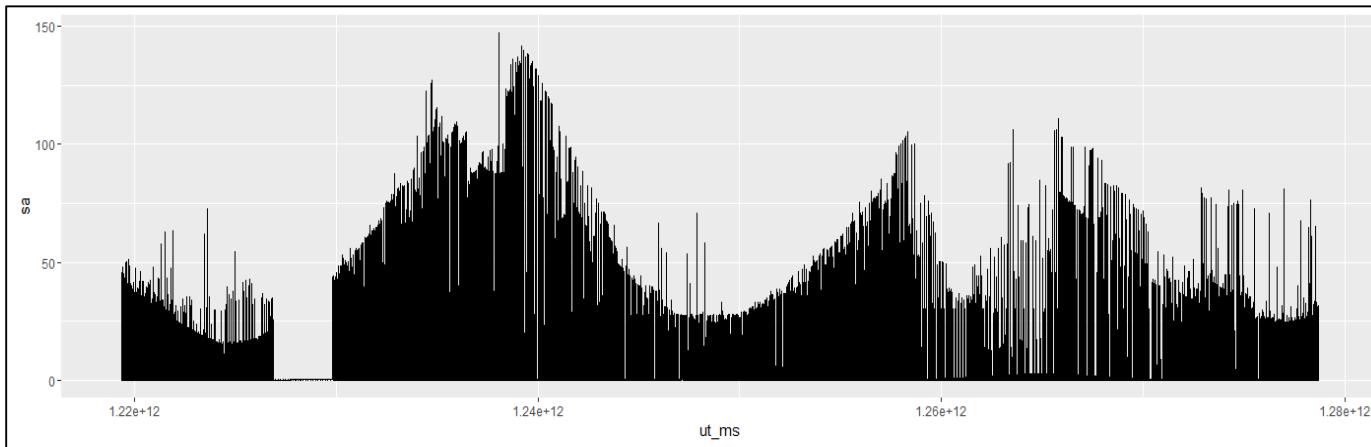
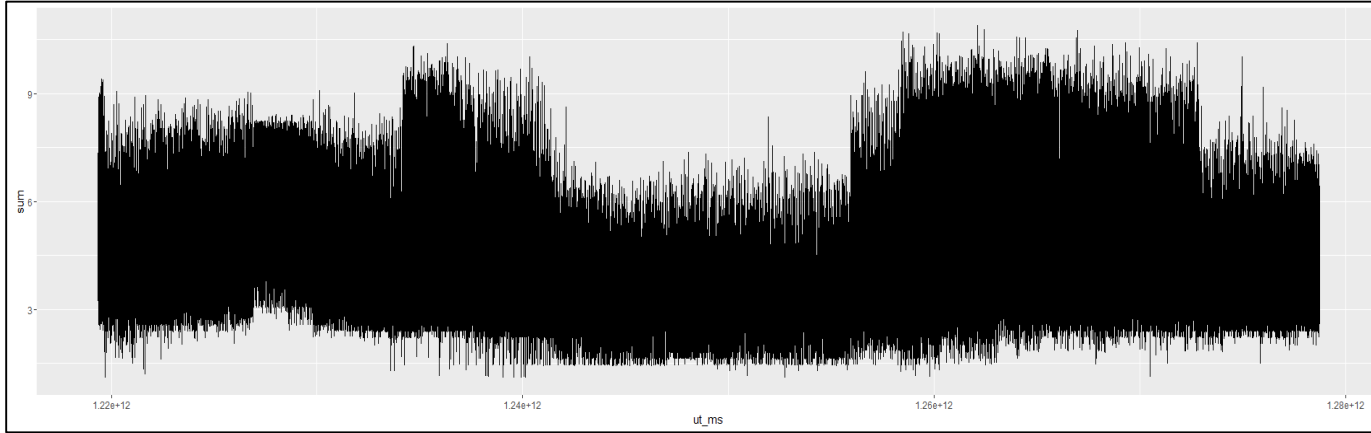
$$E_{\text{Ciencia}} = E_{\text{Solar}} - E_{\text{Nave}} - E_{\text{Climatización}}$$



Consumo: 1 año, Suma circuitos



Consumo [1:500]



Pre-procesamiento

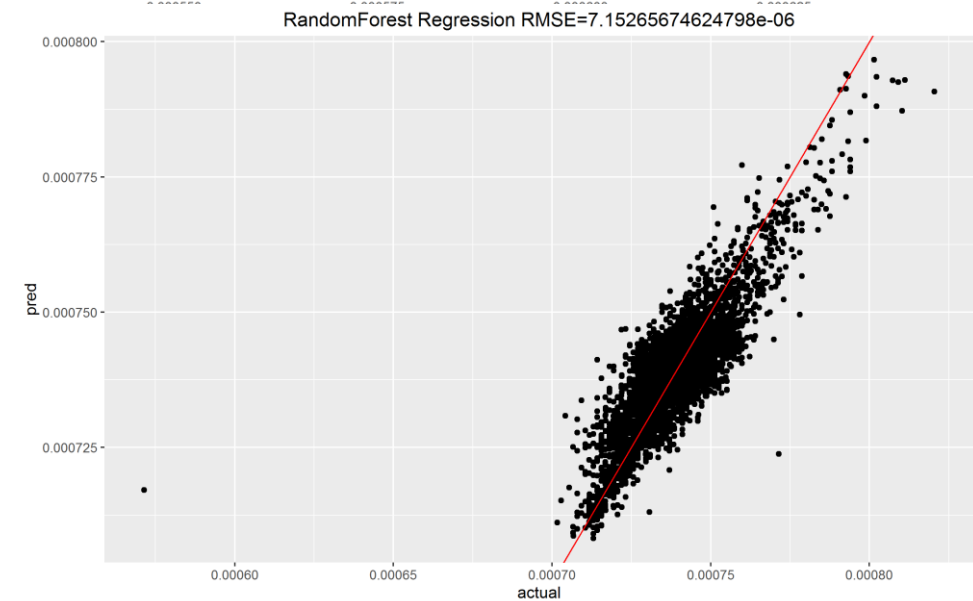
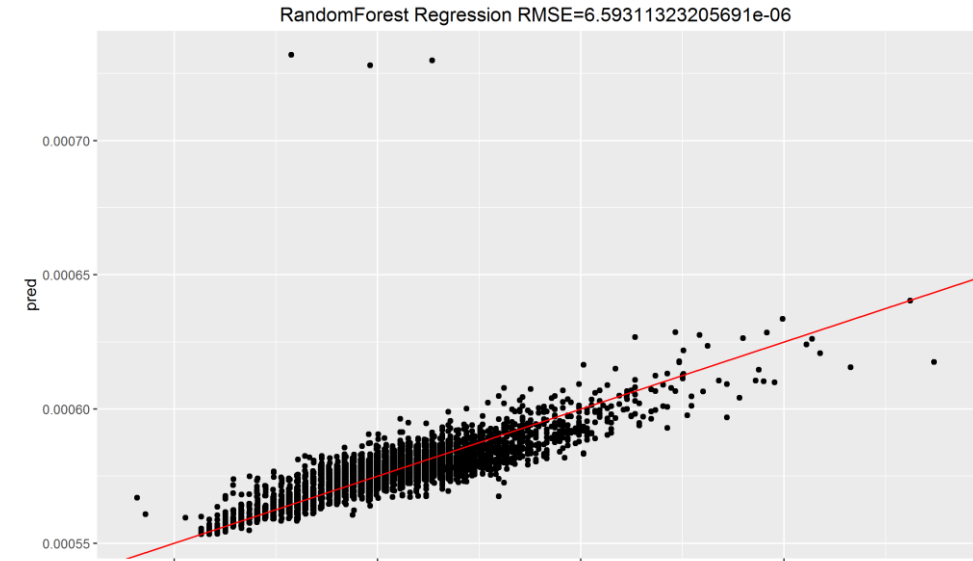
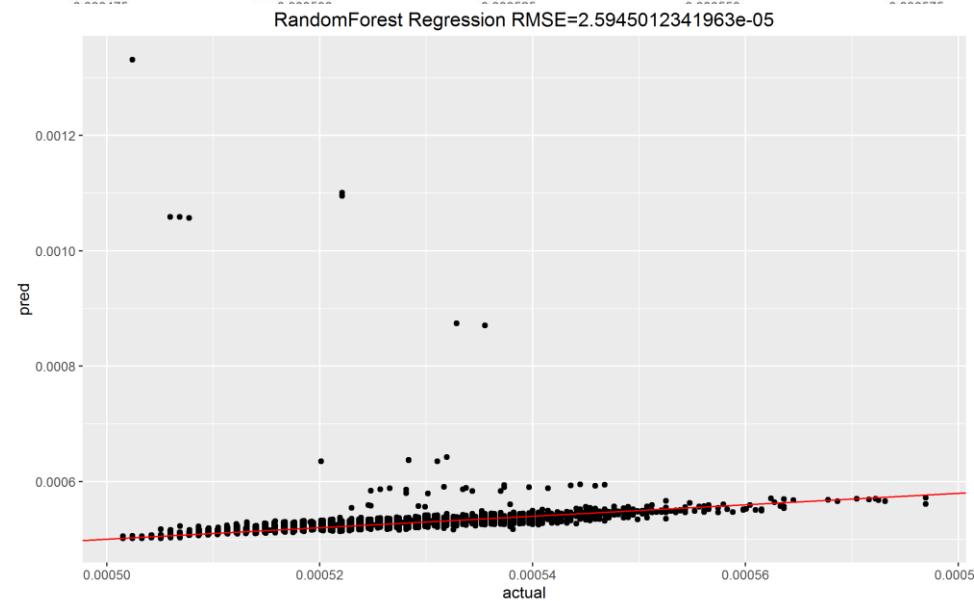
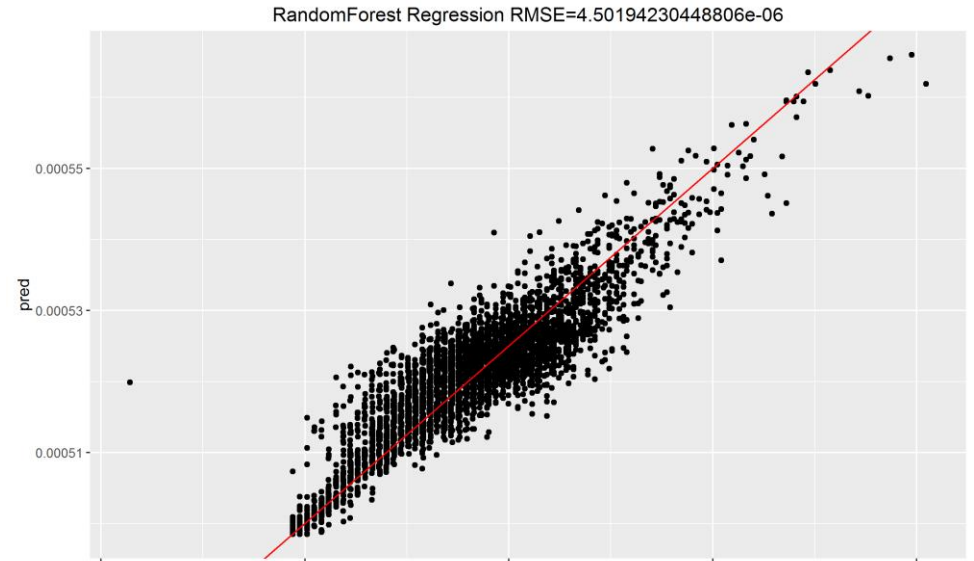
- Ajuste: Escala temporal
- Agrupamiento: Promedio por hora
- Match: Time Scale
- Join *
- Interpolación de valores faltantes
- Split: Train, Test

Procesamiento

- Regresión: Random Forest
- Evaluación: Root Mean Square

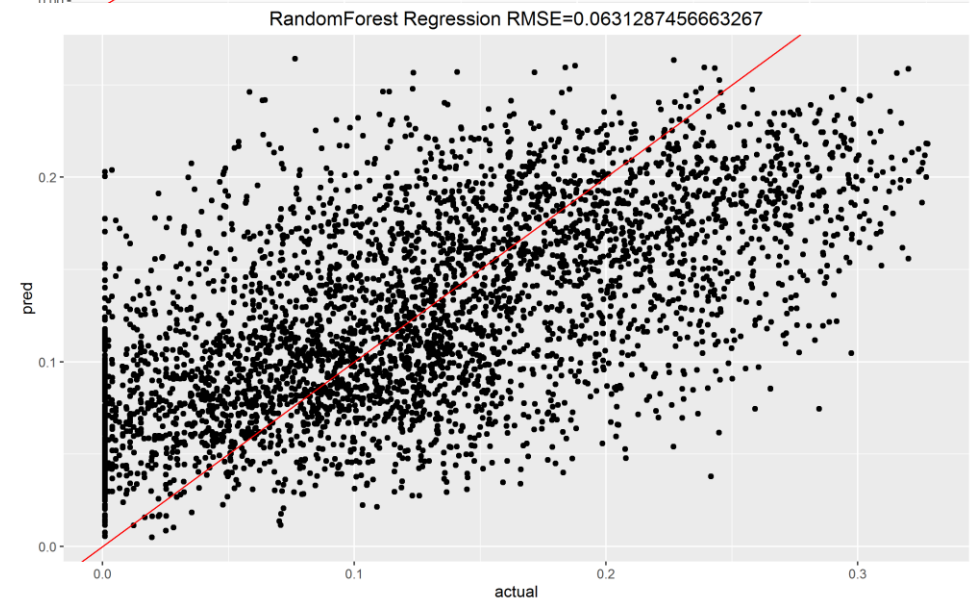
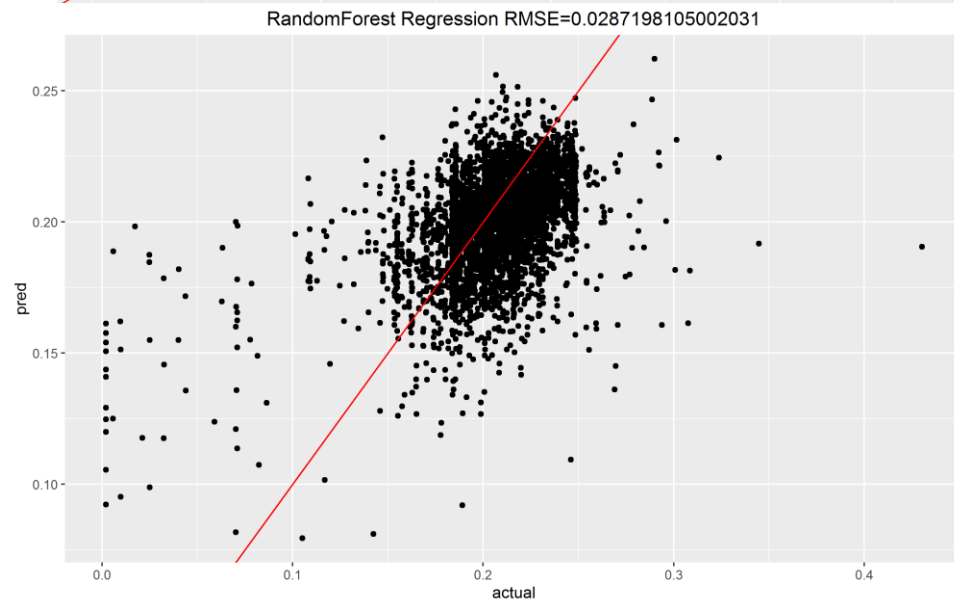
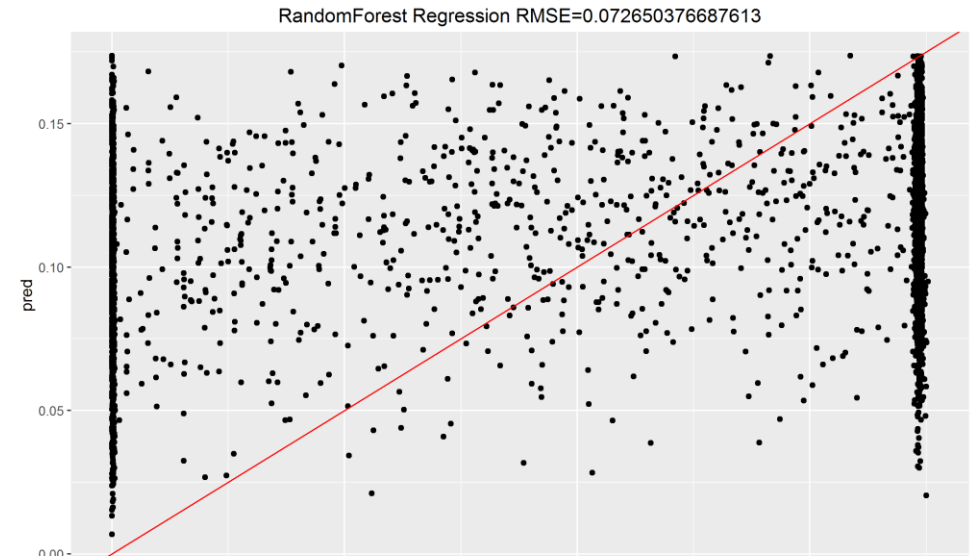
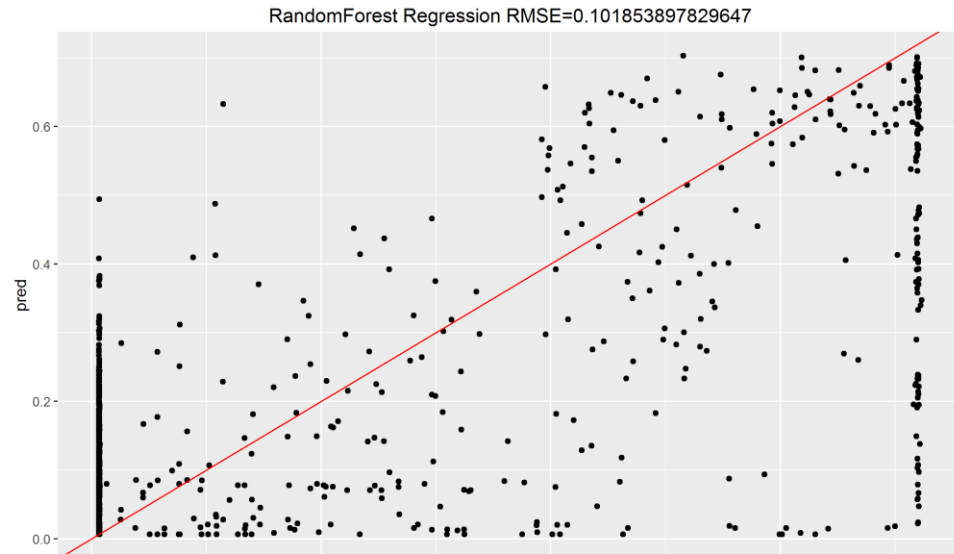
Mars Express Power Challenge

Resultados positivos



Mars Express Power Challenge

Resultados negativos

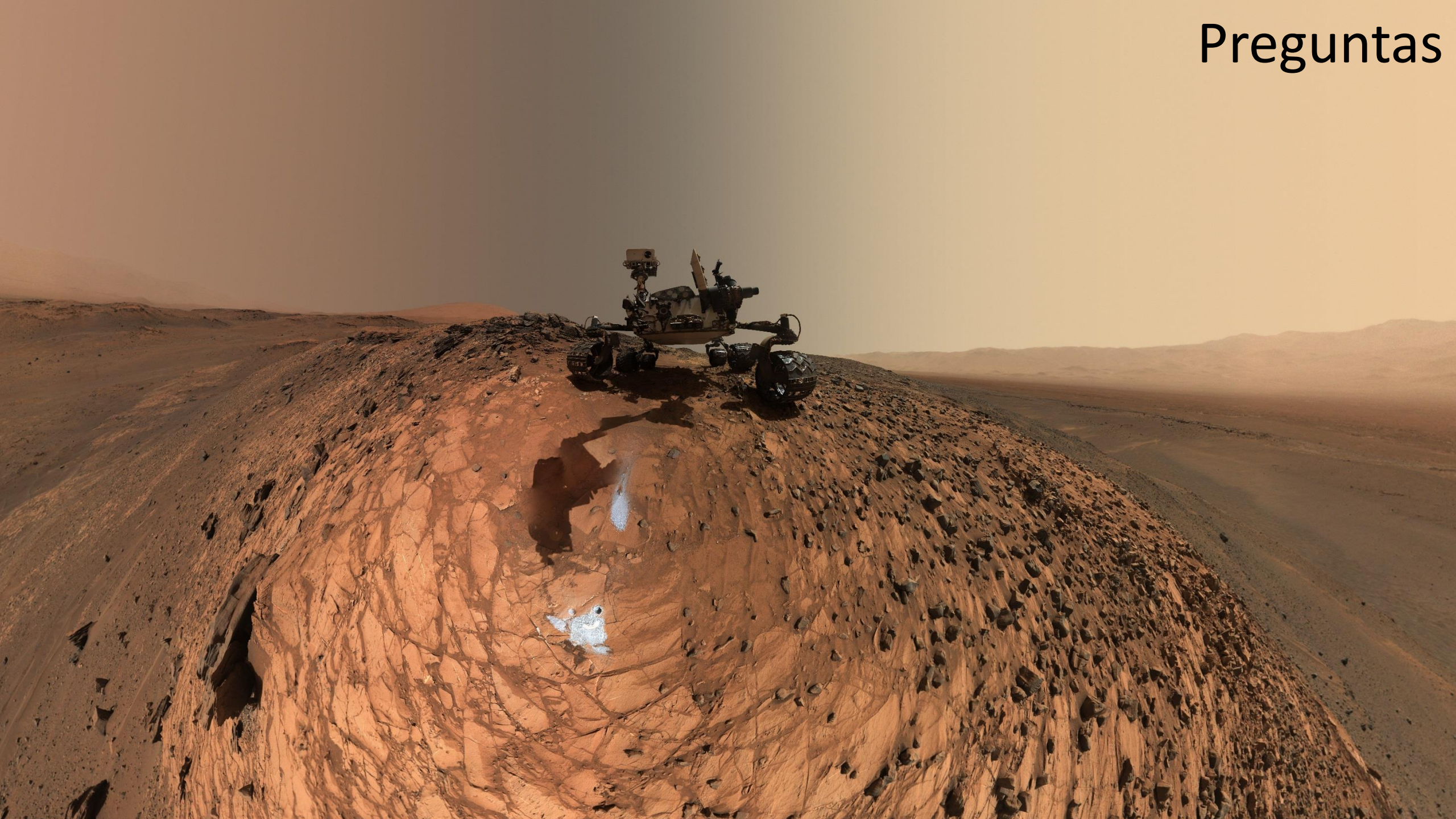


Mars Express Power Challenge

Conclusiones

- Error total: 0.04979785 (5%)
- (Pre)Procesamiento = +Tiempo
- Recursos computacionales
- Generación ~ Consumo
- Predicción macro: Ok
- Predicción micro: Falta procesamiento: DMOP, EVTF, FTL

Preguntas



#Escala Tiempo:

```
power1DT <- power1
power1DT$ut_ms <- as.POSIXct((((power1['ut_ms'])/1000)[,]), origin="1970-01-01")
```

#Agrupamiento: Promedio por hora:

```
power1DT$ut_ms <- cut(power1DT$ut_ms, breaks="hour")
power1DTHourMean <- power1DT %>% group_by(ut_ms) %>% summarise_each(funs(mean))
```

#Match: TimeScale

```
power1DTHourMeanMS <- power1DTHourMean$ut_ms
for (i in 1:nrow(saaf1DTHourMean)) {
  saaf1DTHourMean$ut_ms[i] <- power1DTHourMeanMS[findInterval(saaf1DTHourMean$ut_ms[i],power1DTHourMeanMS)]
}
```

#Merge:

```
power1DTHourMean<-merge(x=power1DTHourMean, y=saaf1DTHourMean, by="ut_ms", all.x=TRUE)
power1DTHourMean<-merge(x=power1DTHourMean, y=ltdata1DTHourMean, by="ut_ms", all.x=TRUE)
```

#Interpolacion:

```
power1DTHourMean$sunmars_km <- na.spline(power1DTHourMean[,grep("sunmars_km",
colnames(power1DTHourMean))],na.rm = FALSE)
power1DTHourMean$earthmars_km <- na.spline(power1DTHourMean[,grep("earthmars_km",
colnames(power1DTHourMean))],na.rm = FALSE)
```

#Regresión

```
rmseSum <- 0
```

```
for(i in 1:33){
```

```
  predictField <- i #Campo a predecir
```

```
  predictCols <- colnames(power1DT[,-1]) #Columnas en juego:
```

```
  #Set de entrenamiento y pruebas
```

```
  train <- power1DTHourMean[1:12000,-1]
```

```
  test <- power1DTHourMean[12001:16000,-1]
```

```
  colName <- predictCols[predictField]
```

```
  #Entrenamiento:
```

```
  rf <- randomForest(as.formula(paste(colName, " ~ .")), data=train, ntree=10)
```

```
  #Prueba:
```

```
  predicted <- predict(rf, test)
```

```
  predCol <- test[,c(colName)]
```

```
  #Medición error:
```

```
  r2 <- RMSE(predCol, predicted)
```

```
  #Graficar Predicción vs. Referencia:
```

```
  p <- ggplot(aes(x=actual, y=pred), data=data.frame(actual=predCol, pred=predict(rf, test)))
```

```
  p <- p + geom_point() + geom_abline(color="red") + ggtitle(paste("RandomForest Regression RMSE=", r2, sep=""))
```

```
  rmseSum <- rmseSum + r2 #Acumulación Error
```

```
  ggsave(paste("Predict",i,".png"), p) #Guardar Imagen
```

```
}
```

```
errorTotal<-rmseSum/33
```