XAI2 task3.py

```
# ------ Import libraries ------
1
2
3
   import numpy as np
4
   import pandas as pd
   from sklearn.ensemble import GradientBoostingRegressor
5
   from rulefit import RuleFit
6
7
   import matplotlib.pyplot as plt
8
   from sklearn.linear_model import LinearRegression
   from sklearn.preprocessing import StandardScaler
9
10
11
12
   # ------ Preprocessing data ------
13
   data_task3= pd.read_csv('datasets/day.csv', sep=',')
14
15
   data_task3['spring'] = np.where(data_task3['season'] == 2, 1, 0)
   data_task3['summer'] = np.where(data_task3['season'] == 3, 1, 0)
16
   data_task3['fall'] = np.where(data_task3['season'] == 4, 1, 0)
17
18
   data_task3['misty'] = np.where(data_task3['weathersit'] == 2, 1, 0)
19
20
   data_task3['rain'] = np.where(data_task3['weathersit'].isin([3, 4]), 1, 0)
21
22
   data_task3['temp']
                        = data_task3['temp'].astype(float) * 47 - 8
23
   data_task3['hum']
                        = data_task3['hum'].astype(float) * 100
   data_task3['windspeed']= data_task3['windspeed'].astype(float) * 67
24
25
   data_task3['dteday'] = pd.to_datetime(data_task3['dteday'])
26
27
   data_task3['days_since_2011'] = (data_task3['dteday'] - pd.Timestamp('2011-01-01')).dt.days
28
29
   # ----- Finding f(x) = y ------
30
   X = data_task3[['fall','spring','summer', 'workingday', 'holiday', 'misty','rain', 'temp','hum','windspeed',
31
    'days since 2011']]
32
   y = data_task3['cnt']
33
34
   # ----- Training rulefit -----
35
   gb = GradientBoostingRegressor(n_estimators=500, max_depth=2, learning_rate=0.01, random_state=13)
36
37
38
39
   rf = RuleFit(
                                    # Implicitly gaussian, so no need to specify family="gaussian"
40
       tree generator = gb,
41
       random state = 13
42
   )
43
                                        # Scaling values to take care for the different ranges of the
44
   scaler = StandardScaler()
   features, especially for the temp feature vs days_since_2011
45
   X_scaled = scaler.fit_transform(X)
46
47
   rf.fit(X_scaled, y.values, feature_names=X.columns.tolist())
48
   # ----- Finding the top 4 rules ------
49
50
   rules = rf.get_rules()
51
   rules filtered = rules[
52
       (rules['importance'] != 0) &
53
       (rules['type'] != 'rule')
54
   1
55
56
   rules_top4 = (rules_filtered
          .sort_values(by='importance', ascending=False)
57
58
          .head(4)
          .reset_index(drop=True)
59
```

```
60 )
61
   print("Top 4 rules:")
62
   print(rules_top4[['rule', 'importance']])
63
64
65 # ------ Plotting the top 4 rules -----
66 fig, ax = plt.subplots(figsize=(8,5))
bars = ax.bar(rules_top4['rule'], rules_top4['importance'])
68 ax.set_ylabel('Importance')
69
   ax.set_title('Task 3 - Top 4 rules')
70 plt.tight_layout()
71
   for bar in bars:
72
73
       h = bar.get_height()
74
       ax.annotate(f'{h:.1f}',
75
                  xy=(bar.get_x() + bar.get_width()/2, h),
76
                  xytext=(0, 3),
                  textcoords="offset points",
77
                  ha='center', va='bottom')
78
79
80 plt.show()
```