```
In []: %matplotlib inline
    import os
    from sklearn.decomposition import PCA
    import matplotlib.pyplot as plt
    import seaborn as sns
    import numpy as np
    import pandas as pd
In []: eng_levels = [0, 1]
```

Face Features

```
person name = "yagmur"
In [ ]:
                               base dir = "../face/features/%s" % person name
                               labels path = "../scores/all levels binary.csv"
In [ ]: labels = pd.read csv(labels path)[person name]
In [ ]: # Load data
                                df = pd.read csv(os.path.join(base dir, 'features.csv'))
                                # Remove empty spaces in column names.
                                df.columns = [col.replace(" ", "") for col in df.columns]
                                # Print few values of data.
                                df.head()
                                          frame face_id_timestamp_confidence_success_gaze_0_x_gaze_0_y_gaze_0_z_gaze_1_x_gaze_1_y ... AU12_c_AU14_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_AU15_c_A
Out[ ]:
                               0
                                                        1
                                                                                   0
                                                                                                                   0.0
                                                                                                                                                       0.98
                                                                                                                                                                                           1 0.309688 0.217516 -0.925624 -0.078749
                                                                                                                                                                                                                                                                                                                                              0.124520 ...
                                                                                                                                                                                                                                                                                                                                                                                                      1.0
                                                                                                                                                                                                                                                                                                                                                                                                                                   1.0
                                                                                                                                                                                                                                                                                                                                                                                                                                                              0.0
                                                        2
                                                                                  0
                                                                                                                   0.0
                                                                                                                                                       0.88
                                                                                                                                                                                           1 -0.744108 -0.003582 -0.668050 -0.731003
                                                                                                                                                                                                                                                                                                                                              0.035917 ...
                                                                                                                                                                                                                                                                                                                                                                                                       0.0
                                                                                                                                                                                                                                                                                                                                                                                                                                   0.0
                                                                                                                                                                                                                                                                                                                                                                                                                                                              0.0
                               2
                                                        3
                                                                                   0
                                                                                                                   0.0
                                                                                                                                                                                           1 -0.730967 -0.056477 -0.680072 -0.739313 -0.004669 ...
                                                                                                                                                                                                                                                                                                                                                                                                                                                              0.0
                                                                                                                                                        0.98
                                                                                                                                                                                                                                                                                                                                                                                                       0.0
                                                                                                                                                                                                                                                                                                                                                                                                                                  1.0
```

1 -0.688489 -0.079700 -0.720854 -0.715481

1 -0.623707 -0.017633 -0.781459 -0.733528

0.028073 ...

0.004099 ...

0.0

0.0

1.0

1.0

0.0

1.0

5 rows × 714 columns

5

0

0

0.0

0.0

0.98

0.98

```
print(len(df), len(labels))
In [ ]:
         6969 6969
         df.describe()
In [ ]:
Out[]:
                      frame face id timestamp
                                                  confidence
                                                                                                                                gaze_1_y ...
                                                                  success
                                                                             gaze_0_x
                                                                                          gaze_0_y
                                                                                                       gaze_0_z
                                                                                                                   gaze_1_x
          count 6969.000000
                              6969.0
                                         6969.0
                                                 6969.000000
                                                              6969.000000
                                                                          6969.000000
                                                                                       6969.000000
                                                                                                    6969.000000
                                                                                                                6969.000000
                                                                                                                             6969.000000
                                                                                                                                             696
                 215.962548
                                 0.0
                                            0.0
                                                    0.766288
                                                                 0.772421
                                                                             0.130164
                                                                                          0.126324
                                                                                                      -0.715448
                                                                                                                   -0.048672
                                                                                                                                0.084677 ...
          mean
                  157.337614
                                 0.0
                                            0.0
                                                    0.362595
                                                                 0.419300
                                                                             0.215496
                                                                                          0.165907
                                                                                                       0.392075
                                                                                                                   0.196111
                                                                                                                                0.131894 ...
            std
                   1.000000
                                 0.0
                                             0.0
                                                    0.000000
                                                                 0.000000
                                                                             -0.799789
                                                                                         -0.466460
                                                                                                      -1.000000
                                                                                                                   -0.776111
                                                                                                                               -0.508339
           min
                                 0.0
                                            0.0
           25%
                   86.000000
                                                    0.880000
                                                                 1.000000
                                                                             0.000000
                                                                                          0.000000
                                                                                                      -0.966496
                                                                                                                   -0.150220
                                                                                                                                0.000000 ...
                  182.000000
                                             0.0
           50%
                                 0.0
                                                    0.980000
                                                                 1.000000
                                                                             0.122265
                                                                                          0.122958
                                                                                                      -0.913610
                                                                                                                    0.000000
                                                                                                                                0.075680
           75%
                 327.000000
                                 0.0
                                             0.0
                                                    0.980000
                                                                 1.000000
                                                                             0.291115
                                                                                          0.257790
                                                                                                      -0.789262
                                                                                                                    0.035816
                                                                                                                                0.183527 ...
                 666.000000
                                 0.0
                                            0.0
                                                    0.980000
                                                                 1.000000
                                                                              0.847260
                                                                                          0.578685
                                                                                                       0.000000
                                                                                                                   0.776535
                                                                                                                                0.501084 ...
           max
        8 \text{ rows} \times 714 \text{ columns}
         high conf ind = ~np.logical or(df['confidence'] < 0.5, df['success'] == 0)
         df = df.loc[high conf ind]
         labels = labels.loc[high_conf_ind]
         print(len(df), len(labels))
In [ ]:
         5381 5381
         labels.value counts()
In [ ]:
               5176
Out[ ]:
                205
         Name: yagmur, dtype: int64
         # Define Feature Series Ranges
In [ ]:
          r au intensities = range(df.columns.get loc("AU01 r"), df.columns.get loc("AU45 r"))
         r au class = range(df.columns.get loc("AU01 c"), df.columns.get loc("AU45 c"))
         r_3d_eye_landmarks = range(df.columns.get_loc("eye_lmk_X_0"), df.columns.get_loc("eye_lmk_Z_55"))
         r gaze directions = range(df.columns.get loc("gaze 0 x"), df.columns.get loc("gaze angle y"))
```

```
C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 18012\819994795.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a
-view-versus-a-copy
  df au intensities['label'] = labels.values
C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 18012\819994795.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a
-view-versus-a-copy
  df au class['label'] = labels.values
C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 18012\819994795.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer, col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a
-view-versus-a-copy
  df 3d eye landmarks['label'] = labels.values
C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 18012\819994795.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a
-view-versus-a-copy
  df gaze directions['label'] = labels.values
C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 18012\819994795.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a
-view-versus-a-copy
  df pose['label'] = labels.values
C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 18012\819994795.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a
-view-versus-a-copy
  df 3d face landmarks['label'] = labels.values
```

```
df_all = pd.concat([df_3d_eye_landmarks.iloc[:, :-1],
        df au intensities.iloc[:, :-1],
        df gaze directions.iloc[:, :-1],
        df 3d face landmarks.iloc[:, :-1],
        df pose],axis=1)
        feature sets = {
             "AU Intensity": df au intensities,
             "3D Eye Landmark": df 3d eye landmarks,
            "3D Face Landmark": df 3d face landmarks,
             "Gaze Directions": df gaze directions,
            "Head Pose": df_pose,
            "3D Face and Head Pose": df face and pose,
            "All OpenFace Fts": df all
        from sklearn.preprocessing import StandardScaler
        from sklearn.model selection import train test split
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.linear_model import LogisticRegression
        from sklearn.svm import SVC
        from sklearn.ensemble import AdaBoostClassifier
        from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
        from sklearn.decomposition import PCA
        from sklearn.pipeline import Pipeline
        from sklearn.model_selection import cross_val_score
        from sklearn.metrics import f1 score
In [ ]: #!pip install git+https://github.com/christophM/rulefit.git
        from rulefit import RuleFit
In [ ]: for title in feature_sets:
            dfc = feature sets[title]
            not zero ind = \sim(dfc == 0).all(axis=1)
            dfc = dfc.loc[not zero ind]
            labels = dfc['label'].loc[not zero ind]
            scaler = StandardScaler()
            scaled samples = scaler.fit transform(dfc.iloc[:,:-2])
```

```
X_train, X_test, y_train, y_test = train_test_split(scaled_samples, labels, test_size=0.2, random_state=42, stratif

features = dfc.columns

rf = RuleFit(model_type='r', rfmode='classify', max_iter=5000, n_jobs=-1) ## Classification task with only rule-bas

rf.fit(X_train, y_train, feature_names=features)

y_pred = rf.predict(X_test)

res = f1_score(y_test, y_pred, average='weighted')

print(rf, "f1", res)

rules = rf.get_rules()

rules = rules[rules.coef != 0].sort_values("support", ascending=False)

rules.to_csv("reports/interpret/%s/rule-%s.csv" % (person_name, title))
```

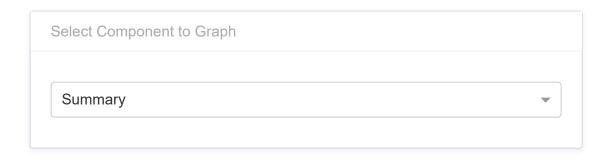
```
RuleFit(max iter=1000, model type='r', n jobs=-1, rfmode='classify',
        tree generator=GradientBoostingClassifier(learning rate=0.01,
                                                   max depth=100,
                                                   max leaf nodes=8,
                                                   n estimators=565,
                                                   random state=564,
                                                   subsample=0.11469081487094829)) f1 0.9536517177833821
RuleFit(max iter=1000, model type='r', n jobs=-1, rfmode='classify',
        tree_generator=GradientBoostingClassifier(learning_rate=0.01,
                                                   max depth=100,
                                                   max leaf nodes=3,
                                                   n estimators=555,
                                                   random state=554,
                                                   subsample=0.11469081487094829)) f1 0.9418720599171825
RuleFit(max iter=1000, model type='r', n jobs=-1, rfmode='classify',
        tree generator=GradientBoostingClassifier(learning rate=0.01,
                                                   max depth=100,
                                                   max leaf nodes=6,
                                                   n estimators=559,
                                                   random state=558,
                                                   subsample=0.11469081487094829)) f1 0.9506059829805349
RuleFit(max iter=1000, model type='r', n jobs=-1, rfmode='classify',
        tree generator=GradientBoostingClassifier(learning rate=0.01,
                                                   max depth=100,
                                                   max leaf nodes=3,
                                                   n estimators=548,
                                                   random state=547,
                                                   subsample=0.11469081487094829)) f1 0.9435081896662996
RuleFit(max iter=1000, model type='r', n jobs=-1, rfmode='classify',
        tree generator=GradientBoostingClassifier(learning rate=0.01,
                                                   max depth=100,
                                                   max leaf nodes=4,
                                                   n estimators=571,
                                                   random state=570,
                                                   subsample=0.11469081487094829)) f1 0.9475060643816446
RuleFit(max iter=1000, model type='r', n jobs=-1, rfmode='classify',
        tree generator=GradientBoostingClassifier(learning rate=0.01,
                                                   max depth=100,
                                                   max leaf nodes=4,
                                                   n estimators=549,
                                                   random state=548,
                                                   subsample=0.11469081487094829)) f1 0.9469428623909881
RuleFit(max iter=1000, model type='r', n jobs=-1, rfmode='classify',
        tree generator=GradientBoostingClassifier(learning rate=0.01,
                                                   max depth=100,
```

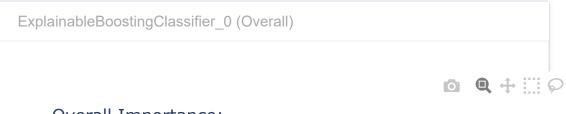
```
max_leaf_nodes=3,
n_estimators=583,
random_state=582,
subsample=0.11469081487094829)) f1 0.9442806902594673
```

```
In [ ]:
        pca = PCA()
        rf = RandomForestClassifier(n estimators=100, n jobs=-1)
        blackbox_model = Pipeline([('pca', pca), ('rf', rf)])
        blackbox model = SVC(gamma=2, C=1, probability=True)
In [ ]:
        from interpret import show
        from interpret.perf import ROC
        from interpret.blackbox import LimeTabular
        from interpret import show
        from interpret.blackbox import ShapKernel
        from interpret.blackbox import MorrisSensitivity
        from interpret.blackbox import PartialDependence
        from interpret.glassbox import ExplainableBoostingClassifier
In [ ]: for title in feature sets:
            ebm = ExplainableBoostingClassifier()
            dfc = feature sets[title]
            not zero ind = \sim(dfc == 0).all(axis=1)
            dfc = dfc.loc[not zero ind]
            labels = dfc['label'].loc[not zero ind]
            scaler = StandardScaler()
            scaled samples = scaler.fit transform(dfc.iloc[:,:-2])
            X train, X test, y train, y test = train test split(scaled samples, labels, test size=0.2, random state=42, stratif
             ebm.fit(X train, y train)
            ebm global = ebm.explain global()
             show(ebm global)
```

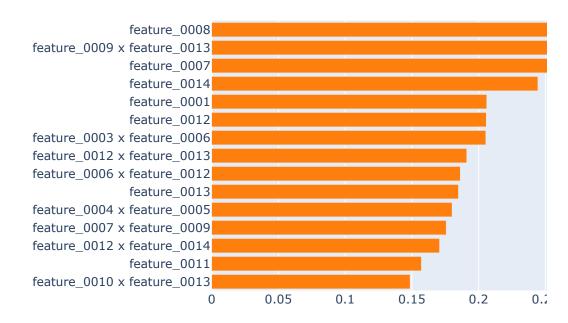
```
c:\Users\ASABUNCUOGLU13\Anaconda3\lib\site-packages\interpret\visual\udash.py:5: UserWarning:
The dash_html_components package is deprecated. Please replace
`import dash_html_components as html` with `from dash import html`
    import dash_html_components as html
c:\Users\ASABUNCUOGLU13\Anaconda3\lib\site-packages\interpret\visual\udash.py:6: UserWarning:
The dash_core_components package is deprecated. Please replace
`import dash_core_components as dcc` with `from dash import dcc`
    import dash_core_components as dcc
c:\Users\ASABUNCUOGLU13\Anaconda3\lib\site-packages\interpret\visual\udash.py:7: UserWarning:
The dash_table package is deprecated. Please replace
`import dash_table` with `from dash import dash_table`

Also, if you're using any of the table format helpers (e.g. Group), replace
`from dash_table.Format import Group` with
`from dash_table.Format import Group` with
import dash table as dt
```



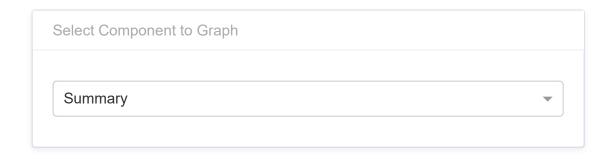


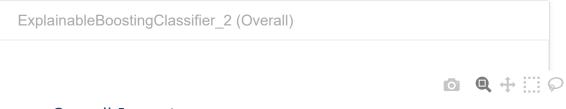
Overall Importance: Mean Absolute Score



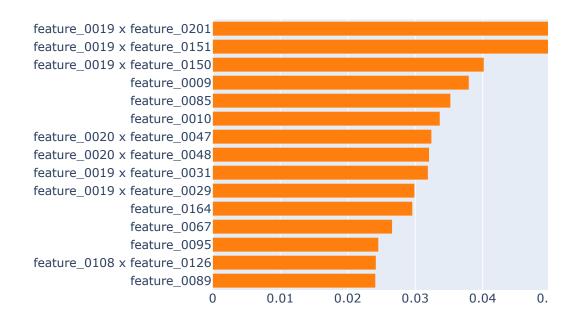


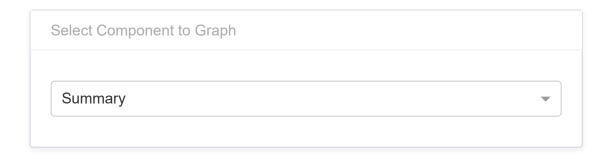


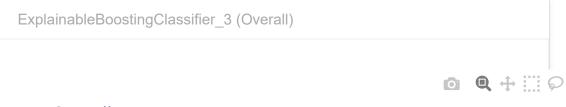




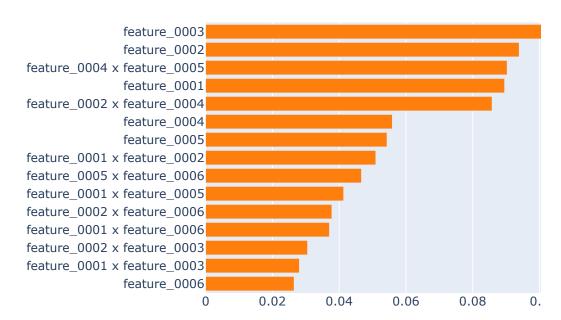
Overall Importance: Mean Absolute Score







Overall Importance: Mean Absolute Score















```
In []: for title in feature_sets:
    dfc = feature_sets[title]
    not_zero_ind = ~(dfc == 0).all(axis=1)

    dfc = dfc.loc[not_zero_ind]
    labels = dfc['label'].loc[not_zero_ind]

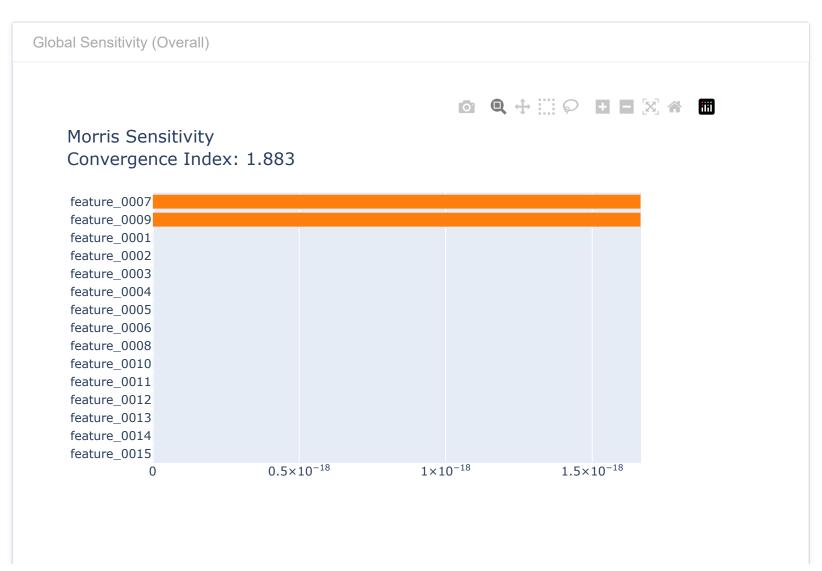
    scaler = StandardScaler()
    scaled_samples = scaler.fit_transform(dfc.iloc[:,:-2])

    X_train, X_test, y_train, y_test = train_test_split(scaled_samples, labels, test_size=0.2, random_state=42, stratify
    blackbox_model.fit(X_train, y_train)
    try:
        sensitivity = MorrisSensitivity(predict_fn=blackbox_model.predict_proba, data=X_train)
        sensitivity_global = sensitivity.explain_global(name="Global Sensitivity")

        show(sensitivity_global)

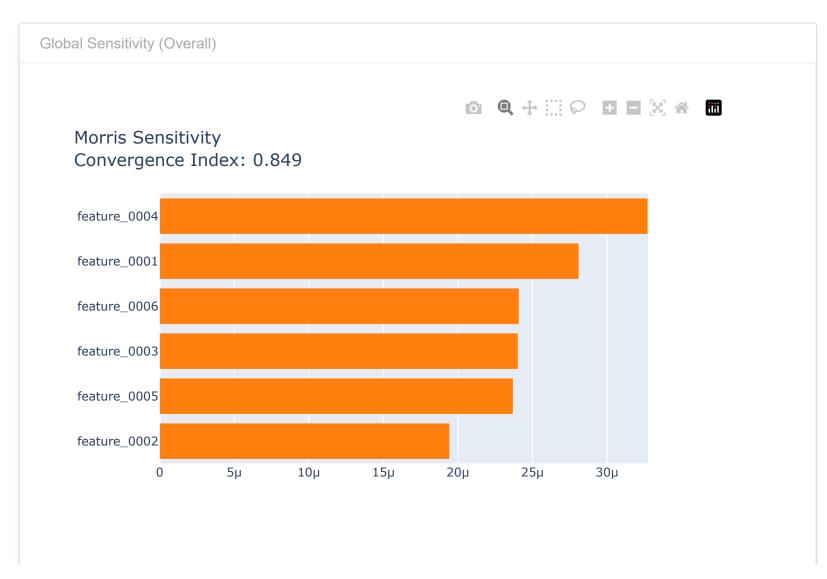
except ValueError:
        print("zero-size array to reduction operation maximum which has no identity")
```



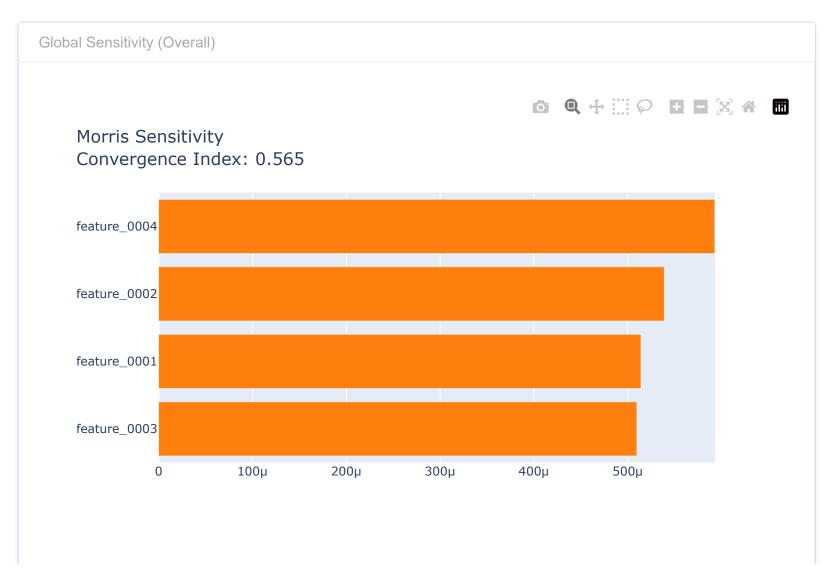


zero-size array to reduction operation maximum which has no identity zero-size array to reduction operation maximum which has no identity









zero-size array to reduction operation maximum which has no identity zero-size array to reduction operation maximum which has no identity

In []: