```
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**

```
In [ ]: base dir = "../pose-action/features/"
In [ ]: # Load data
         df = pd.read csv(os.path.join(base dir, 'pose keypoints with labels.csv'))
        labels = (df['label'] <= 0).astype(int) #binarize Labels</pre>
         df = df.iloc[:,:-1]
        df.head()
Out[ ]:
                        1
                                 2
                                         3
                                                         5
                                                                 6
                                                                                  8
                                                                                          9 ... 290 291 292 293 294 295 296 297
        0 178.864 180.766 0.968300 198.379 261.046 0.854602 131.697 270.835 0.764647 125.874 ... 0.0
                                                                                                                                  0.0
        1 739.041 147.424 0.908013 758.704 212.014 0.949156 715.641 210.162 0.872539 695.949 ...
                                                                                                0.0
                                                                                                     0.0
                                                                                                          0.0
                                                                                                               0.0
                                                                                                                   0.0
                                                                                                                         0.0
                                                                                                                             0.0
                                                                                                                                  0.0
        2 196.373 180.794 0.960075 206.199 261.029 0.837968 133.707 266.929 0.740769 120.070 ... 0.0 0.0
                                                                                                          0.0
                                                                                                               0.0
                                                                                                                   0.0
                                                                                                                             0.0 0.0
        3 196.522 182.778 0.885628 208.134 259.180 0.857302 133.798 263.053 0.737462 120.051 ...
                                                                                                               0.0
                                                                                                                   0.0
                                                                                                0.0 0.0
                                                                                                                             0.0 0.0
         4 694.069 196.346 0.955753 729.249 214.107 0.902440 686.212 217.955 0.793841 668.633 ... 0.0 0.0 0.0 0.0
                                                                                                                   0.0
                                                                                                                             0.0 0.0
        5 rows × 300 columns
In [ ]: print(len(df), len(labels))
        6935 6935
```

```
In [ ]: labels.value_counts()
             5491
Out[ ]:
             1444
        Name: label, dtype: int64
In [ ]: r_p1 = range(0,24)
        r_p2 = range(75,99)
        r_p3 = range(150, 174)
        r_p4 = range(225,249)
        r_all = np.r_[r_p1, r_p2, r_p3, r_p4]
        df_p1 = df.iloc[:, r_p1]
        df_p2 = df.iloc[:, r_p2]
        df_p3 = df.iloc[:, r_p3]
        df_p4 = df.iloc[:, r_p4]
        df all = df.iloc[:, r all]
In [ ]: df_p1['label'] = labels.values
        df p2['label'] = labels.values
        df_p3['label'] = labels.values
        df p4['label'] = labels.values
        df_all['label'] = labels.values
```

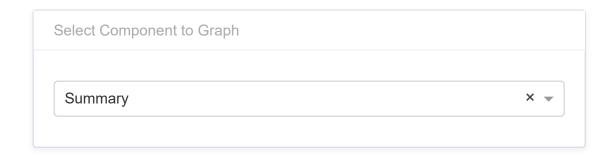
```
C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 9320\2577857134.py:1: SettingWithCopyWarning:
        A value is trying to be set on a copy of a slice from a DataFrame.
        Try using .loc[row indexer,col indexer] = value instead
        See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a
        -view-versus-a-copy
          df p1['label'] = labels.values
        C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel_9320\2577857134.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 p2['label'] = labels.values
        C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 9320\2577857134.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 p3['label'] = labels.values
        C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 9320\2577857134.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 p4['label'] = labels.values
        C:\Users\ASABUNCUOGLU13\AppData\Local\Temp\ipykernel 9320\2577857134.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 all['label'] = labels.values
In [ ]: feature sets = {
            "P1": df p1,
            "P2": df p2,
            "P3": df p3,
            "P4": df p4,
            "All Features": df all
```

```
In [ ]: 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/pose/rule-%s.csv" % 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=2,
                                                           n estimators=550,
                                                            random state=549,
                                                            subsample=0.09857775536929808)) f1 0.8184506891569311
        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=567,
                                                            random state=566,
                                                            subsample=0.09857775536929808)) f1 0.8080967814539559
        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=9,
                                                           n estimators=556,
                                                            random state=555,
                                                            subsample=0.09857775536929808)) f1 0.801881292594165
        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=5,
                                                           n estimators=588,
                                                            random state=587,
                                                            subsample=0.09857775536929808)) f1 0.8360663648295513
        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=2,
                                                           n estimators=546,
                                                            random state=545,
                                                            subsample=0.09857775536929808)) f1 0.8564249976297093
         0.000
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)
        from interpret import show
```

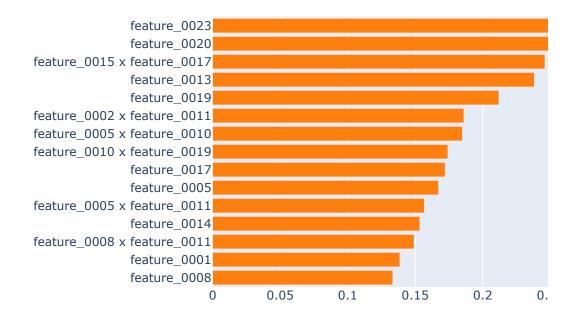
```
from interpret.blackbox import LimeTabular
        from interpret import show
        from interpret.blackbox import ShapKernel
        from interpret.blackbox import MorrisSensitivity
        from interpret.blackbox import PartialDependence
        from interpret.glassbox import ExplainableBoostingClassifier
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[:,:-1])
            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.dash table.Format import Group`
          import dash table as dt
```

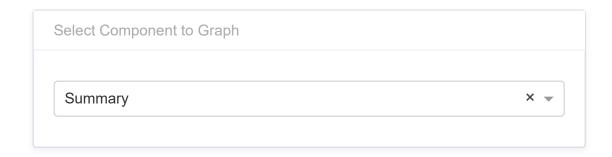
from interpret.perf import ROC



ExplainableBoostingClassifier\_0 (Overall)

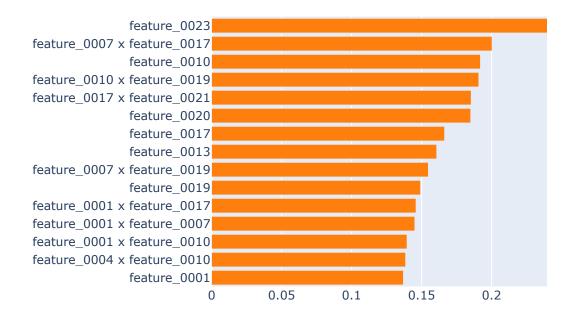
## Overall Importance: Mean Absolute Score

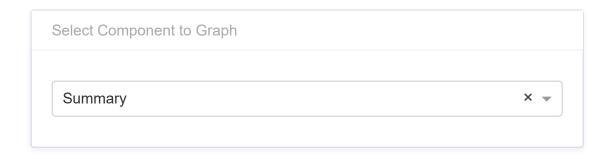




ExplainableBoostingClassifier\_1 (Overall)

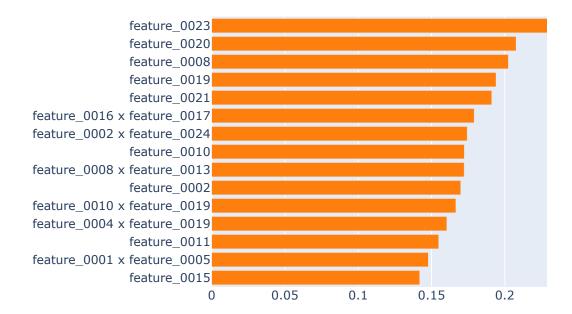
## Overall Importance: Mean Absolute Score





ExplainableBoostingClassifier\_2 (Overall)

## Overall Importance: Mean Absolute Score











```
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[:,:-1])
            X_train, X_test, y_train, y_test = train_test_split(scaled_samples, labels, test_size=0.2, random_state=42, stratif
            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
        zero-size array to reduction operation maximum which has no identity
```

In [ ]: