

# Using Survey Data as a Predictor of Pandemic Vaccination

## Jupyter Dash Results Dashboard

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### Displaying Classificaion Modeling Results

This interactive dashboard allows a user to select between 3 different data preparations, and see the associated model results for 6 models. This is a work in progress and I hope to add on additional elements to display, for other parts of the project and an EDA chart explorer.

#### Import the relevant libraries and data

In [1]: `from jupyter_dash import JupyterDash`

In [2]: `# Import the libraries
import dash
import dash_core_components as dcc
import dash_html_components as html
import pandas as pd
import plotly.express as px
from jupyter_dash import JupyterDash
from dash.dependencies import Input, Output`

In [4]: `# When running in JupyterHub or Binder, call the infer_jupyter_config function to detect the proxy configuration.
JupyterDash.infer_jupyter_proxy_config()`

In [ ]:

#### Creating the tables from classification modeling results

In [3]: `import dash_table`

In [4]: `# Base modeling
data1 = [['XGBoost', 0.85, 0.68], ['Random Forest', 0.84, 0.68], ['SVC', 0.84, 0.67], ['Logistic Regression', 0.84, 0.66], ['KNN', 0.81, 0.54], ['Decision Trees', 0.75, 0.40]]
dfa_results = pd.DataFrame(data1, columns = ['Classification_model', 'Accuracy', 'Precision1'])
dfa_results`

Out[4]:

	Classification_model	Accuracy	Precision1
0	XGBoost	0.85	0.68
1	Random Forest	0.84	0.68
2	SVC	0.84	0.67
3	Logistic Regression	0.84	0.66
4	KNN	0.81	0.54
5	Decision Trees	0.75	0.40

In [5]: `# Base modeling with SMOTE
data2 = [['XGBoost', 0.83, 0.61], ['Random Forest', 0.83, 0.64], ['SVC', 0.80, 0.52], ['Logistic Regression', 0.77, 0.48], ['KNN', 0.67, 0.35], ['Decision Trees', 0.74, 0.41]]
dfB_results = pd.DataFrame(data2, columns = ['Classification_model', 'Accuracy', 'Precision1'])
dfB_results`

Out[5]:

	Classification_model	Accuracy	Precision1
0	XGBoost	0.83	0.61
1	Random Forest	0.83	0.64
2	SVC	0.80	0.52
3	Logistic Regression	0.77	0.48
4	KNN	0.67	0.35
5	Decision Trees	0.74	0.41

In [6]: `# Approach B - Base modeling
data3 = [['XGBoost', 0.84, 0.71], ['Random Forest', 0.84, 0.72], ['SVC', 0.83, 0.70], ['Logistic Regression', 0.83, 0.69], ['KNN', 0.80, 0.59], ['Decision Trees', 0.76, 0.45]]
dfC_results = pd.DataFrame(data3, columns = ['Classification_model', 'Accuracy', 'Precision1'])
dfC_results`

Out[6]:

	Classification_model	Accuracy	Precision1
0	XGBoost	0.84	0.71
1	Random Forest	0.84	0.72
2	SVC	0.83	0.70
3	Logistic Regression	0.83	0.69
4	KNN	0.80	0.59
5	Decision Trees	0.76	0.45

#### Creating the code to display the dashboard in this notebook

```
In [8]: # One basic table - for modeling results
# Build App
external_stylesheets = ['https://codepen.io/chriddyp/pen/bWLwgP.css']
app = JupyterDash(__name__, external_stylesheets=external_stylesheets)

Table1 = dash_table.DataTable(
    id='tableA',
    columns=[{"name": i, "id": i} for i in dfA_results.columns],
    data=dfA_results.to_dict('records'),
    style_cell=dict(textAlign='left'),
    style_header=dict(background-color="paleturquoise"),
    style_data=dict(background-color="lavender")
)

Table2 = dash_table.DataTable(
    id='tableB',
    columns=[{"name": i, "id": i} for i in dfB_results.columns],
    data=dfB_results.to_dict('records'),
    style_cell=dict(textAlign='left'),
    style_header=dict(background-color="paleturquoise"),
    style_data=dict(background-color="lavender")
)

Table3 = dash_table.DataTable(
    id='tableC',
    columns=[{"name": i, "id": i} for i in dfC_results.columns],
    data=dfC_results.to_dict('records'),
    style_cell=dict(textAlign='left'),
    style_header=dict(background-color="paleturquoise"),
    style_data=dict(background-color="lavender")
)

# This is the layout of the dashboard
app.layout = html.Div([
    html.H2("Classification Model Results"),
    html.Label([
        "Select data preparation to see results of 6 models",
        dcc.Dropdown(
            id='column-dropdown', clearable=False,
            value = 'A-Basic-preprocessing', options=[
                {'label': c, 'value': c}
                for c in ['A-Basic-preprocessing', 'B-Basic-with-SMOTE', 'C-One-Hot-Encoded']
            ])
    ]),
    html.Div(id='results_table', style={'padding': 20})
])

# Define callback to update graph
@app.callback(
    Output('results_table', 'children'),
    [Input("column-dropdown", "value")]
)
def update_figure(model):
    if model == 'A-Basic-preprocessing':
        return Table1
    if model == 'B-Basic-with-SMOTE':
        return Table2
    if model == 'C-One-Hot-Encoded':
        return Table3

# Run app and display result inline in the notebook
app.run_server(mode='inline')
```

# Classification Model Results

Select data preparation to see results of 6 models

C-One-Hot-Encoded

Classification_model	Accuracy	Precision1
XGBoost	0.84	0.71
Random Forest	0.84	0.72
SVC	0.83	0.7
Logistic Regression	0.83	0.69
KNN	0.8	0.59
Decision Trees	0.76	0.45

**Observations on the dashboard**

This is really just a proof of concept at this point. There are some changes and enhancements I would like to make, including:

- figure out why excess whitespace is included below the table
- add a title and description above the table - this would need to change with each table
- better control the width of the table columns - I know how to change the proportion, but also need to change the overall size of the table
- improve the size and font display
- Change the color fill within the table