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Appendices

Protocols to run the data preprocessing and preparation workflow

Below are the procedures to run the script shown on Figure 6. The current protocol is adapted from the data preprocessing pipeline conducted by Harutyunyan et al. (2019). The directory for which the script is run might differ between each local computer or hardware and thus must be adjusted accordingly. Certain scripts might require different modules or packages scattered through the directories, thus it is advisable to run the application on the same directory to prevent errors. Furthermore, the required python package is shown on the top of each script and imported accordingly.

1. Iterates through all the MIMIC-III csv dataset files and extracts redundant data points and removes the pediatric patient category. A directory containing each ICU stays will be generated based on each SUBJECT_ID. The output file is a CSV file of 'stays.csv', 'diagnoses.csv', and 'events.csv'. [**extract_subjects.py**]
2. Fix issues with missing ICUSTAYS_ID stays and HADM_ID [**validate_events.py**]
3. Breaks up per-subject data into separate episodes (pertaining to ICU stays) [**extract_episodes_from_subjects.py**]
4. Splits the whole dataset into training and testing sets [**split_train_and_test.py**]
5. Generate task-specific datasets, which can later be used in models [**create_in_hospital_mortality.py** and **create_length_of_stay.py**]
6. Extract validation set from the training set [**split_train_val.py**]

→ **extract_subjects.py**

```
from __future__ import absolute_import
from __future__ import print_function

import argparse
import yaml
import os

from mimic3benchmark.mimic3csv import *
from mimic3benchmark.preprocessing import
add_hcup_ccs_2015_groups, make_phenotype_label_matrix
from mimic3benchmark.util import *

parser = argparse.ArgumentParser(description='Extract per-
subject data from MIMIC-III CSV files.')
parser.add_argument('mimic3_path', type=str,
help='Directory containing MIMIC-III CSV files.')
parser.add_argument('output_path', type=str,
help='Directory where per-subject data should be written.')
parser.add_argument('--event_tables', '-e', type=str,
nargs='+', help='Tables from which to read events.',
```

```

        default=['CHARTEVENTS', 'LABEVENTS',
'OUTPUTEVENTS'])
parser.add_argument('--phenotype_definitions', '-p',
type=str,

default=os.path.join(os.path.dirname(__file__),
'../resources/hcup_ccs_2015_definitions.yaml'),
        help='YAML file with phenotype
definitions.')
parser.add_argument('--itemids_file', '-i', type=str,
help='CSV containing list of ITEMIDs to keep.')
parser.add_argument('--verbose', '-v', type=int,
help='Level of verbosity in output.', default=1)
parser.add_argument('--test', action='store_true',
help='TEST MODE: process only 1000 subjects, 1000000
events.')
args, _ = parser.parse_known_args()

try:
    os.makedirs(args.output_path)
except:
    pass

patients = read_patients_table(args.mimic3_path)
admits = read_admissions_table(args.mimic3_path)
stays = read_icustays_table(args.mimic3_path)
if args.verbose:
    print('START:', stays.icustay_id.unique().shape[0],
stays.hadm_id.unique().shape[0],
        stays.subject_id.unique().shape[0])

stays = remove_icustays_with_transfers(stays)
if args.verbose:
    print('REMOVE ICU TRANSFERS:',
stays.icustay_id.unique().shape[0],
stays.hadm_id.unique().shape[0],
        stays.subject_id.unique().shape[0])

stays = merge_on_subject_admission(stays, admits)
stays = merge_on_subject(stays, patients)
stays = filter_admissions_on_nb_icustays(stays)
if args.verbose:
    print('REMOVE MULTIPLE STAYS PER ADMIT:',
stays.icustay_id.unique().shape[0],
stays.hadm_id.unique().shape[0],
        stays.subject_id.unique().shape[0])

stays = add_age_to_icustays(stays)
stays = add_inunit_mortality_to_icustays(stays)
stays = add_inhospital_mortality_to_icustays(stays)
stays = filter_icustays_on_age(stays)

```



```

if args.verbose:
    print('REMOVE PATIENTS AGE < 18:',
          stays.icustay_id.unique().shape[0],
          stays.hadm_id.unique().shape[0],
          stays.subject_id.unique().shape[0])

stays.to_csv(os.path.join(args.output_path,
                          'all_stays.csv'), index=False)
diagnoses = read_icd_diagnoses_table(args.mimic3_path)
diagnoses = filter_diagnoses_on_stays(diagnoses, stays)
diagnoses.to_csv(os.path.join(args.output_path,
                              'all_diagnoses.csv'), index=False)
count_icd_codes(diagnoses,
                output_path=os.path.join(args.output_path,
                                          'diagnosis_counts.csv'))

phenotypes = add_hcup_ccs_2015_groups(diagnoses,
                                     yaml.load(open(args.phenotype_definitions, 'r')))
make_phenotype_label_matrix(phenotypes,
                             stays).to_csv(os.path.join(args.output_path,
                                                         'phenotype_labels.csv'),
                                     index=False, quoting=csv.QUOTE_NONNUMERIC)

if args.test:
    pat_idx = np.random.choice(patients.shape[0],
                              size=1000)
    patients = patients.iloc[pat_idx]
    stays = stays.merge(patients[['subject_id']],
                       left_on='subject_id', right_on='subject_id')
    args.event_tables = [args.event_tables[0]]
    print('Using only', stays.shape[0], 'stays and only',
          args.event_tables[0], 'table')

subjects = stays.subject_id.unique()
break_up_stays_by_subject(stays, args.output_path,
                          subjects=subjects, verbose=args.verbose)
break_up_diagnoses_by_subject(phenotypes, args.output_path,
                              subjects=subjects, verbose=args.verbose)
items_to_keep = set(
    [int(itemid) for itemid in
     dataframe_from_csv(args.itemids_file)['itemid'].unique()])
if args.itemids_file else None
for table in args.event_tables:

    read_events_table_and_break_up_by_subject(args.mimic3_path,
                                              table, args.output_path,
                                              items_to_keep=items_to_keep,

                                              subjects_to_keep=subjects, verbose=args.verbose)

```

→ **validate_events.py**

```

from __future__ import absolute_import
from __future__ import print_function

import os
import argparse
import pandas as pd

def is_subject_folder(x):
    return str.isdigit(x)

def main():

    n_events = 0 # total number of events
    empty_hadm = 0 # HADM_ID is empty in
events.csv. We exclude such events.
    no_hadm_in_stay = 0 # HADM_ID does not
appear in stays.csv. We exclude such events.
    no_icustay = 0 # ICUSTAY_ID is empty in
events.csv. We try to fix such events.
    recovered = 0 # empty ICUSTAY_IDs are
recovered according to stays.csv files (given HADM_ID)
    could_not_recover = 0 # empty ICUSTAY_IDs that
are not recovered. This should be zero.
    icustay_missing_in_stays = 0 # ICUSTAY_ID does not
appear in stays.csv. We exclude such events.

    parser = argparse.ArgumentParser()
    parser.add_argument('subjects_root_path', type=str,
                        help='Directory containing subject
subdirectories.')
    args = parser.parse_args()
    print(args)

    subdirectories = os.listdir(args.subjects_root_path)
    subjects = list(filter(is_subject_folder,
subdirectories))

    for (index, subject) in enumerate(subjects):
        if index % 100 == 0:
            print("processed {} / {} {}".format(index+1,
len(subjects), ' '*10))

        stays_df =
pd.read_csv(os.path.join(args.subjects_root_path, subject,
'stays.csv'), index_col=False,
            dtype={'HADM_ID': str,
"ICUSTAY_ID": str})
        stays_df.columns = stays_df.columns.str.upper()

```

```

        # assert that there is no row with empty ICUSTAY_ID
or HADM_ID
        assert(not stays_df['ICUSTAY_ID'].isnull().any())
        assert(not stays_df['HADM_ID'].isnull().any())

        # assert there are no repetitions of ICUSTAY_ID or
HADM_ID
        # since admissions with multiple ICU stays were
excluded
        assert(len(stays_df['ICUSTAY_ID'].unique()) ==
len(stays_df['ICUSTAY_ID']))
        assert(len(stays_df['HADM_ID'].unique()) ==
len(stays_df['HADM_ID']))

        events_df =
pd.read_csv(os.path.join(args.subjects_root_path, subject,
'events.csv'), index_col=False,
                dtype={'HADM_ID': str,
"ICUSTAY_ID": str})
        events_df.columns = events_df.columns.str.upper()
        n_events += events_df.shape[0]

        # we drop all events for them HADM_ID is empty
        # TODO: maybe we can recover HADM_ID by looking at
ICUSTAY_ID
        empty_hadm += events_df['HADM_ID'].isnull().sum()
        events_df = events_df.dropna(subset=['HADM_ID'])

        merged_df = events_df.merge(stays_df,
left_on=['HADM_ID'], right_on=['HADM_ID'],
                                how='left',
suffixes=['', '_r'], indicator=True)

        # we drop all events for which HADM_ID is not
listed in stays.csv
        # since there is no way to know the targets of that
stay (for example mortality)
        no_hadm_in_stay += (merged_df['_merge'] ==
'left_only').sum()
        merged_df = merged_df[merged_df['_merge'] ==
'both']

        # if ICUSTAY_ID is empty in stays.csv, we try to
recover it
        # we exclude all events for which we could not
recover ICUSTAY_ID
        cur_no_icustay =
merged_df['ICUSTAY_ID'].isnull().sum()
        no_icustay += cur_no_icustay
        merged_df.loc[:, 'ICUSTAY_ID'] =
merged_df['ICUSTAY_ID_r'].fillna(merged_df['ICUSTAY_ID_r'])

```

```

        recovered += cur_no_icustay -
merged_df['ICUSTAY_ID'].isnull().sum()
        could_not_recover +=
merged_df['ICUSTAY_ID'].isnull().sum()
        merged_df = merged_df.dropna(subset=['ICUSTAY_ID'])

        # now we take a look at the case when ICUSTAY_ID is
present in events.csv, but not in stays.csv
        # this mean that ICUSTAY_ID in events.csv is not
the same as that of stays.csv for the same HADM_ID
        # we drop all such events
        icustay_missing_in_stays +=
(merged_df['ICUSTAY_ID'] !=
merged_df['ICUSTAY_ID_r']).sum()
        merged_df = merged_df[(merged_df['ICUSTAY_ID'] ==
merged_df['ICUSTAY_ID_r'])]

        to_write = merged_df[['SUBJECT_ID', 'HADM_ID',
'ICUSTAY_ID', 'CHARTTIME', 'ITEMID', 'VALUE', 'VALUEUOM']]

to_write.to_csv(os.path.join(args.subjects_root_path,
subject, 'events.csv'), index=False)

        assert(could_not_recover == 0)
        print('n_events: {}'.format(n_events))
        print('empty_hadm: {}'.format(empty_hadm))
        print('no_hadm_in_stay: {}'.format(no_hadm_in_stay))
        print('no_icustay: {}'.format(no_icustay))
        print('recovered: {}'.format(recovered))
        print('could_not_recover:
{}'.format(could_not_recover))
        print('icustay_missing_in_stays:
{}'.format(icustay_missing_in_stays))

if __name__ == "__main__":
    main()

```

→ **extract_episodes_from_subjects.py**

```

from __future__ import absolute_import
from __future__ import print_function

import argparse

import os
import sys

from mimic3benchmark.subject import read_stays,
read_diagnoses, read_events, get_events_for_stay,
add_hours_elapsed_to_events

```

```

from mimic3benchmark.subject import
convert_events_to_timeseries,
get_first_valid_from_timeseries
from mimic3benchmark.preprocessing import
read_itemid_to_variable_map, map_itemids_to_variables,
read_variable_ranges, clean_events
from mimic3benchmark.preprocessing import transform_gender,
transform_ethnicity, assemble_episodic_data

parser = argparse.ArgumentParser(description='Extract
episodes from per-subject data.')
parser.add_argument('subjects_root_path', type=str,
help='Directory containing subject sub-directories.')
parser.add_argument('--variable_map_file', type=str,

default=os.path.join(os.path.dirname(__file__),
'../resources/itemid_to_variable_map.csv'),
help='CSV containing ITEMID-to-VARIABLE
map.')
parser.add_argument('--reference_range_file', type=str,

default=os.path.join(os.path.dirname(__file__),
'../resources/variable_ranges.csv'),
help='CSV containing reference ranges
for VARIABLES.')
parser.add_argument('--verbose', '-v', type=int,
help='Level of verbosity in output.', default=1)
args, _ = parser.parse_known_args()

var_map =
read_itemid_to_variable_map(args.variable_map_file)
variables = var_map.VARIABLE.unique()

for subject_dir in os.listdir(args.subjects_root_path):
    dn = os.path.join(args.subjects_root_path, subject_dir)
    try:
        subject_id = int(subject_dir)
        if not os.path.isdir(dn):
            raise Exception
    except:
        continue
    sys.stdout.write('Subject {}: '.format(subject_id))
    sys.stdout.flush()

    try:
        sys.stdout.write('reading...')
        sys.stdout.flush()
        stays =
read_stays(os.path.join(args.subjects_root_path,
subject_dir))

```

```

        diagnoses =
read_diagnoses(os.path.join(args.subjects_root_path,
subject_dir))
        events =
read_events(os.path.join(args.subjects_root_path,
subject_dir))
        except:
            sys.stdout.write('error reading from disk!\n')
            continue
        else:
            sys.stdout.write('got {0} stays, {1} diagnoses, {2}
events...'.format(stays.shape[0], diagnoses.shape[0],
events.shape[0]))
            sys.stdout.flush()

        episodic_data = assemble_episodic_data(stays,
diagnoses)

        sys.stdout.write('cleaning and converting to time
series...')
        sys.stdout.flush()
        events = map_itemids_to_variables(events, var_map)
        events = clean_events(events)
        if events.shape[0] == 0:
            sys.stdout.write('no valid events!\n')
            continue
        timeseries = convert_events_to_timeseries(events,
variables=variables)

        sys.stdout.write('extracting separate episodes...')
        sys.stdout.flush()

        for i in range(stays.shape[0]):
            stay_id = stays.ICUSTAY_ID.iloc[i]
            sys.stdout.write(' {}'.format(stay_id))
            sys.stdout.flush()
            intime = stays.INTIME.iloc[i]
            outtime = stays.OUTTIME.iloc[i]

            episode = get_events_for_stay(timeseries, stay_id,
intime, outtime)
            if episode.shape[0] == 0:
                sys.stdout.write(' (no data!)')
                sys.stdout.flush()
                continue

            episode = add_hours_elapsed_to_events(episode,
intime).set_index('HOURS').sort_index(axis=0)
            episodic_data.Weight.ix[stay_id] =
get_first_valid_from_timeseries(episode, 'Weight')

```

```

        episodic_data.Height.ix[stay_id] =
get_first_valid_from_timeseries(episode, 'Height')

episodic_data.ix[episodic_data.index==stay_id].to_csv(os.pa
th.join(args.subjects_root_path, subject_dir,
'episode{}.csv'.format(i+1)), index_label='Icustay')
        columns = list(episode.columns)
        columns_sorted = sorted(columns, key=(lambda x: ""
if x == "Hours" else x))
        episode = episode[columns_sorted]

episode.to_csv(os.path.join(args.subjects_root_path,
subject_dir, 'episode{}_timeseries.csv'.format(i+1)),
index_label='Hours')
        sys.stdout.write(' DONE!\n')

```

→ **split_train_and_test.py**

```

from __future__ import absolute_import
from __future__ import print_function

import os
import shutil
import argparse

def move_to_partition(args, patients, partition):
    if not
os.path.exists(os.path.join(args.subjects_root_path,
partition)):
        os.mkdir(os.path.join(args.subjects_root_path,
partition))
        for patient in patients:
            src = os.path.join(args.subjects_root_path,
patient)
            dest = os.path.join(args.subjects_root_path,
partition, patient)
            shutil.move(src, dest)

def main():
    parser = argparse.ArgumentParser(description='Split
data into train and test sets.')
    parser.add_argument('subjects_root_path', type=str,
help='Directory containing subject sub-directories.')
    args, _ = parser.parse_known_args()

    test_set = set()
    with open(os.path.join(os.path.dirname(__file__),
'../resources/testset.csv'), "r") as test_set_file:
        for line in test_set_file:

```

```

        x, y = line.split(',')
        if int(y) == 1:
            test_set.add(x)

    folders = os.listdir(args.subjects_root_path)
    folders = list((filter(str.isdigit, folders)))
    train_patients = [x for x in folders if x not in
test_set]
    test_patients = [x for x in folders if x in test_set]

    assert len(set(train_patients) & set(test_patients)) ==
0

    move_to_partition(args, train_patients, "train")
    move_to_partition(args, test_patients, "test")

if __name__ == '__main__':
    main()

```

→ **create_in_hospital_mortality.py**

```

from __future__ import absolute_import
from __future__ import print_function

import os
import argparse
import pandas as pd
import random
random.seed(49297)

def process_partition(args, partition, eps=1e-6,
n_hours=48):
    output_dir = os.path.join(args.output_path, partition)
    if not os.path.exists(output_dir):
        os.mkdir(output_dir)

    xy_pairs = []
    patients = list(filter(str.isdigit,
os.listdir(os.path.join(args.root_path, partition))))
    for (patient_index, patient) in enumerate(patients):
        patient_folder = os.path.join(args.root_path,
partition, patient)
        patient_ts_files = list(filter(lambda x:
x.find("timeseries") != -1, os.listdir(patient_folder)))

        for ts_filename in patient_ts_files:
            with open(os.path.join(patient_folder,
ts_filename)) as tsfile:

```



```

        lb_filename =
ts_filename.replace("_timeseries", "")
        label_df =
pd.read_csv(os.path.join(patient_folder, lb_filename))

        # empty label file
        if label_df.shape[0] == 0:
            continue

        mortality =
int(label_df.iloc[0]["Mortality"])
        los = 24.0 * label_df.iloc[0]['Length of
Stay'] # in hours
        if pd.isnull(los):
            print("\n\t(length of stay is
missing)", patient, ts_filename)
            continue

        if los < n_hours - eps:
            continue

        ts_lines = tsfile.readlines()
        header = ts_lines[0]
        ts_lines = ts_lines[1:]
        event_times = [float(line.split(',')[0])
for line in ts_lines]

        ts_lines = [line for (line, t) in
zip(ts_lines, event_times)
                    if -eps < t < n_hours + eps]

        # no measurements in ICU
        if len(ts_lines) == 0:
            print("\n\t(no events in ICU) ",
patient, ts_filename)
            continue

        output_ts_filename = patient + "_" +
ts_filename
        with open(os.path.join(output_dir,
output_ts_filename), "w") as outfile:
            outfile.write(header)
            for line in ts_lines:
                outfile.write(line)

        xy_pairs.append((output_ts_filename,
mortality))

        if (patient_index + 1) % 100 == 0:

```

```

        print("processed {} / {}
patients".format(patient_index + 1, len(patients)),
end='\r')

    print("\n", len(xy_pairs))
    if partition == "train":
        random.shuffle(xy_pairs)
    if partition == "test":
        xy_pairs = sorted(xy_pairs)

    with open(os.path.join(output_dir, "listfile.csv"),
"w") as listfile:
        listfile.write('stay,y_true\n')
        for (x, y) in xy_pairs:
            listfile.write('{} ,{:d}\n'.format(x, y))

def main():
    parser = argparse.ArgumentParser(description="Create
data for in-hospital mortality prediction task.")
    parser.add_argument('root_path', type=str, help="Path
to root folder containing train and test sets.")
    parser.add_argument('output_path', type=str,
help="Directory where the created data should be stored.")
    args, _ = parser.parse_known_args()

    if not os.path.exists(args.output_path):
        os.makedirs(args.output_path)

    process_partition(args, "test")
    process_partition(args, "train")

if __name__ == '__main__':
    main()

```

→ **create_length_of_stay.py**

```

from __future__ import absolute_import
from __future__ import print_function

import os
import argparse
import numpy as np
import pandas as pd
import random
random.seed(49297)

def process_partition(args, partition, sample_rate=1.0,
shortest_length=4.0, eps=1e-6):
    output_dir = os.path.join(args.output_path, partition)

```

```

if not os.path.exists(output_dir):
    os.mkdir(output_dir)

xty_triples = []
patients = list(filter(str.isdigit,
os.listdir(os.path.join(args.root_path, partition))))
for (patient_index, patient) in enumerate(patients):
    patient_folder = os.path.join(args.root_path,
partition, patient)
    patient_ts_files = list(filter(lambda x:
x.find("timeseries") != -1, os.listdir(patient_folder)))

    for ts_filename in patient_ts_files:
        with open(os.path.join(patient_folder,
ts_filename)) as tsfile:
            lb_filename =
ts_filename.replace("_timeseries", "")
            label_df =
pd.read_csv(os.path.join(patient_folder, lb_filename))

            # empty label file
            if label_df.shape[0] == 0:
                print("\n\t(empty label file)",
patient, ts_filename)
                continue

            los = 24.0 * label_df.iloc[0]['Length of
Stay'] # in hours
            if pd.isnull(los):
                print("\n\t(length of stay is
missing)", patient, ts_filename)
                continue

            ts_lines = tsfile.readlines()
            header = ts_lines[0]
            ts_lines = ts_lines[1:]
            event_times = [float(line.split(',')[0])
for line in ts_lines]

            ts_lines = [line for (line, t) in
zip(ts_lines, event_times)
                        if -eps < t < los + eps]
            event_times = [t for t in event_times
                           if -eps < t < los + eps]

            # no measurements in ICU
            if len(ts_lines) == 0:
                print("\n\t(no events in ICU) ",
patient, ts_filename)
                continue

```

```

        sample_times = np.arange(0.0, los + eps,
sample_rate)

        sample_times = list(filter(lambda x: x >
shortest_length, sample_times))

        # At least one measurement
        sample_times = list(filter(lambda x: x >
event_times[0], sample_times))

        output_ts_filename = patient + "_" +
ts_filename
        with open(os.path.join(output_dir,
output_ts_filename), "w") as outfile:
            outfile.write(header)
            for line in ts_lines:
                outfile.write(line)

        for t in sample_times:
            xty_triples.append((output_ts_filename,
t, los - t))

        if (patient_index + 1) % 100 == 0:
            print("processed {} / {}
patients".format(patient_index + 1, len(patients)),
end='\r')

        print(len(xty_triples))
        if partition == "train":
            random.shuffle(xty_triples)
        if partition == "test":
            xty_triples = sorted(xty_triples)

        with open(os.path.join(output_dir, "listfile.csv"),
"w") as listfile:
            listfile.write('stay,period_length,y_true\n')
            for (x, t, y) in xty_triples:
                listfile.write('{} {:.6f} {:.6f}\n'.format(x,
t, y))

def main():
    parser = argparse.ArgumentParser(description="Create
data for length of stay prediction task.")
    parser.add_argument('root_path', type=str, help="Path
to root folder containing train and test sets.")
    parser.add_argument('output_path', type=str,
help="Directory where the created data should be stored.")
    args, _ = parser.parse_known_args()

    if not os.path.exists(args.output_path):

```

```

        os.makedirs(args.output_path)

        process_partition(args, "test")
        process_partition(args, "train")

if __name__ == '__main__':
    main()

```

→ **split_train_val.py**

```

from __future__ import absolute_import
from __future__ import print_function

import shutil
import argparse
import os

def main():
    parser = argparse.ArgumentParser(description="Split
train data into train and validation sets.")
    parser.add_argument('dataset_dir', type=str, help='Path
to the directory which contains the dataset')
    args, _ = parser.parse_known_args()

    val_patients = set()
    with open(os.path.join(os.path.dirname(__file__),
'resources/valset.csv'), 'r') as valset_file:
        for line in valset_file:
            x, y = line.split(',')
            if int(y) == 1:
                val_patients.add(x)

    with open(os.path.join(args.dataset_dir,
'train/listfile.csv')) as listfile:
        lines = listfile.readlines()
        header = lines[0]
        lines = lines[1:]

    train_lines = [x for x in lines if x[:x.find("_")] not
in val_patients]
    val_lines = [x for x in lines if x[:x.find("_")] in
val_patients]
    assert len(train_lines) + len(val_lines) == len(lines)

    with open(os.path.join(args.dataset_dir,
'train_listfile.csv'), 'w') as train_listfile:
        train_listfile.write(header)
        for line in train_lines:
            train_listfile.write(line)

```

```

        with open(os.path.join(args.dataset_dir,
'val_listfile.csv'), 'w') as val_listfile:
            val_listfile.write(header)
            for line in val_lines:
                val_listfile.write(line)

        shutil.copy(os.path.join(args.dataset_dir,
'test/listfile.csv'),
                    os.path.join(args.dataset_dir,
'test_listfile.csv'))

if __name__ == '__main__':
    main()

```

Protocols to run the machine learning models

- In-Hospital Mortality Prediction

```

import numpy as np
import pandas as pd

from sklearn.model_selection import train_test_split
from sklearn.pipeline import Pipeline
from sklearn.compose import ColumnTransformer

from jcopml.pipeline import num_pipe, cat_pipe
from jcopml.utils import save_model, load_model
from jcopml.plot import plot_missing_value
from jcopml.feature_importance import mean_score_decrease

# -----Import Data
X_train = np.load("data/train_X.npy")
y_train = np.load("data/train_y.npy")

X_val = np.load("data/val_X.npy")
y_val = np.load("data/val_y.npy")

X_test = np.load("data/test_X.npy")
y_test = np.load("data/test_y.npy")

# -----Training
from xgboost import XGBClassifier
from sklearn.model_selection import RandomizedSearchCV
from jcopml.tuning import random_search_params as rsp

preprocessor = ColumnTransformer([
    ('numeric', num_pipe(), list(range(X_train.shape[1])))
])

```

```

pipeline = Pipeline([
    ('prep', preprocessor),
    ('algo', XGBClassifier(n_jobs=-1, random_state=42))
])

model = RandomizedSearchCV(pipeline, rsp.xgb_params, cv=4,
    scoring='f1', n_iter=50, n_jobs=-1, verbose=1,
    random_state=42)
model.fit(X_train, y_train)

print(model.best_params_)
print(model.score(X_train, y_train), model.score(X_val,
    y_val), model.score(X_test, y_test))

# -----Evaluation Report
from jcopml.plot import plot_classification_report

plot_classification_report(X_train, y_train, X_test,
    y_test, model, report=True)

# -----Save Models
save_model(model.best_estimator_, "xgboost.pkl")

```

- **Length of Stay**

```

from __future__ import absolute_import
from __future__ import print_function

from sklearn.preprocessing import StandardScaler
from sklearn.impute import SimpleImputer
from sklearn.model_selection import StratifiedKFold
from sklearn.model_selection import cross_val_score
from mimic3benchmark.readers import LengthOfStayReader
from mimic3models import common_utils
from mimic3models.metrics import print_metrics_regression
from mimic3models.length_of_stay.utils import save_results

import xgboost as xgb
import matplotlib as plt

import os
import numpy as np
import argparse
import json
import pandas as pd
import csv

def read_and_extract_features(reader, count, period,
    features):

```

```

    read_chunk_size = 1000
    Xs = []
    ys = []
    names = []
    ts = []
    for i in range(0, count, read_chunk_size):
        j = min(count, i + read_chunk_size)
        ret = common_utils.read_chunk(reader, j - i)
        X =
common_utils.extract_features_from_rawdata(ret['X'],
ret['header'], period, features)
        Xs.append(X)
        ys += ret['y']
        names += ret['name']
        ts += ret['t']
    Xs = np.concatenate(Xs, axis=0)
    return (Xs, ys, names, ts)

def main():
    parser = argparse.ArgumentParser()
    parser.add_argument('--period', type=str,
default='all', help='specifies which period extract
features from',
                        choices=['first4days',
'first8days', 'last12hours', 'first25percent',
'first50percent', 'all'])
    parser.add_argument('--features', type=str,
default='all', help='specifies what features to extract',
                        choices=['all', 'len',
'all_but_len'])
    parser.add_argument('--data', type=str, help='Path to
the data of length-of-stay task',
default=os.path.join(os.path.dirname(__file__),
'../../../../../data/length-of-stay/'))
    parser.add_argument('--output_dir', type=str,
help='Directory relative which all output files are
stored',
                        default='.')
    args = parser.parse_args()
    print(args)

    train_reader =
LengthOfStayReader(dataset_dir=os.path.join(args.data,
'train'),

listfile=os.path.join(args.data, 'train_listfile.csv'))

```



```

    val_reader =
LengthOfStayReader(dataset_dir=os.path.join(args.data,
'train'),

listfile=os.path.join(args.data, 'val_listfile.csv'))

    test_reader =
LengthOfStayReader(dataset_dir=os.path.join(args.data,
'test'),

listfile=os.path.join(args.data, 'test_listfile.csv'))

    print('Reading data and extracting features ...')
    n_train = min(100000,
train_reader.get_number_of_examples())
    n_val = min(100000,
val_reader.get_number_of_examples())

    (train_X, train_y, train_names, train_ts) =
read_and_extract_features(
    train_reader, n_train, args.period, args.features)

    (val_X, val_y, val_names, val_ts) =
read_and_extract_features(
    val_reader, n_val, args.period, args.features)

    (test_X, test_y, test_names, test_ts) =
read_and_extract_features(
    test_reader, test_reader.get_number_of_examples(),
args.period, args.features)

    print('Imputing missing values ...')
    imputer = SimpleImputer(missing_values=np.nan,
strategy='mean')
    imputer.fit(train_X)
    train_X =
pd.DataFrame(np.array(imputer.transform(train_X),
dtype=np.float32))
    val_X = pd.DataFrame(np.array(imputer.transform(val_X),
dtype=np.float32))
    test_X =
pd.DataFrame(np.array(imputer.transform(test_X),
dtype=np.float32))

    print('Normalizing the data to have zero mean and unit
variance ...')
    scaler = StandardScaler()
    scaler.fit(train_X)
    train_X = scaler.transform(train_X)
    val_X = scaler.transform(val_X)
    test_X = scaler.transform(test_X)

```

```

file_name = "{}.{}".format(args.period, args.features)

model = xgb.XGBRegressor(learning_rate=0.01,
random_state=42)
kfold = StratifiedKFold(n_splits=4, random_state=7)

result_dir = os.path.join(args.output_dir, 'results')
common_utils.create_directory(result_dir)

with open(os.path.join(result_dir,
'train_{}.json'.format(file_name)), "w") as res_file:
    ret = print_metrics_regression(train_y,
xgb_model.predict(train_X))
    ret = {k: float(v) for k, v in ret.items()}
    json.dump(ret, res_file)

with open(os.path.join(result_dir,
'val_{}.json'.format(file_name)), 'w') as res_file:
    ret = print_metrics_regression(val_y,
xgb_model.predict(val_X))
    ret = {k: float(v) for k, v in ret.items()}
    json.dump(ret, res_file)

prediction = xgb_model.predict(test_X)

with open(os.path.join(result_dir,
'test_{}.json'.format(file_name)), 'w') as res_file:
    ret = print_metrics_regression(test_y, prediction)
    ret = {k: float(v) for k, v in ret.items()}
    json.dump(ret, res_file)

save_results(test_names, test_ts, prediction, test_y,
            os.path.join(args.output_dir,
'predictions', file_name + '.csv'))

if __name__ == '__main__':
    main()

```

Script to run the dashboard web application platform

```

import dash
import dash_core_components as dcc
import dash_html_components as html
import plotly.express as px
import plotly.graph_objs as go
import pandas as pd
from dash.dependencies import Input, Output

```

```

import dash_bootstrap_components as dbc
import dash_table as dt
import plotly.figure_factory as ff
# import base64

# image_filename =
'/Users/anggi/PycharmProjects/Final_Dash/logo-plotly.svg'
# encoded_image = base64.b64encode(open(image_filename,
'rb').read())

# Styling and Dash Components -----
-----

# external_stylesheets = [
'https://codepen.io/chridryp/pen/bWLwgP.css',
# 'https://codepen.io/chridryp/pen/brPBPO.css'
, [dbc.themes.BOOTSTRAP]]
BS =
"https://stackpath.bootstrapcdn.com/bootstrap/4.4.1/css/bootstrap.min.css"
app = dash.Dash(external_stylesheets=[BS])
app.config.suppress_callback_exceptions = True
app.title = 'MIMIC-III Dashboard Exploration'

# All Diagnosis table on Tab 1 - Files
# -----
-----
tab1_stays = pd.read_csv(
    "{PATH_TO_CSV_DATA}",
    sep=",",
    skiprows=0,
    error_bad_lines=False,
    low_memory=False)

# Nav Bar 2 - Hospital Mortality Files
# -----
-----
tab2_stays = pd.read_csv(
    "{PATH_TO_CSV_DATA}",
    sep=",",
    skiprows=0,
    error_bad_lines=False,
    low_memory=False)
tab2_in_cm_test = pd.read_csv(
    "{PATH_TO_CSV_DATA}",
    sep=";",
    skiprows=0,
    error_bad_lines=False,
    low_memory=False)
tab2_in_cm_test_report = pd.read_csv(
    "{PATH_TO_CSV_DATA}",
    sep=";",

```

```

        skiprows=0,
        error_bad_lines=False,
        low_memory=False)
tab2_in_cm_train_report = pd.read_csv(
    "{PATH_TO_CSV_DATA}",
    sep=";",
    skiprows=0,
    error_bad_lines=False,
    low_memory=False)
tab2_in_cm_train = pd.read_csv(
    "{PATH_TO_CSV_DATA}",
    sep=";",
    skiprows=0,
    error_bad_lines=False,
    low_memory=False)

# Nav Bar 3 - Length of Stay -----
-----
tab3 = pd.read_csv(
    "{PATH_TO_CSV_DATA}",
    sep=",",
    skiprows=0,
    error_bad_lines=False,
    low_memory=False)

tab3_in_cm = pd.read_csv(
    "{PATH_TO_CSV_DATA}",
    sep=",",
    skiprows=0,
    error_bad_lines=False,
    low_memory=False)

# Filter Column used in Filtering Method and Buttons
# Styling Filter Button
# -----
-----
features = ['ETHNICITY', 'DIAGNOSIS', 'GENDER', 'AGE',
'MORTALITY_INUNIT', 'MORTALITY_INHOSPITAL']
opts = [{'label': i, 'value': i} for i in features]

white_button_style = {'background-color': 'white', 'color':
'Blue', }
blue_button_style = {'background-color': 'blue', 'color':
'White'}

# Variables to Call Plotly for Diagrams in Landing Page
# -----
-----
diag_histo = tab2_stays["DIAGNOSIS"]
age_histo = tab2_stays["AGE"]
sex_histo = tab2_stays["GENDER"]

```

```

# Figure for Diagnosis Distributions
fig = go.Figure(
    data=[go.Histogram(x=diag_histo, marker={'color': '#009dff'})],
    layout={
        'title': {'text': 'Disease Diagnosis Occurrence',
                  'font': {'size': 40}},
        'font': {'family': 'Barlow Semi Condensed'},
        'yaxis': {'title': 'Number of Occurrences'},
    }
)

# Figure for Age and Sex Distribution
# -----
histo = px.histogram(tab2_stays,
                    x=age_histo,
                    color=sex_histo,
                    histnorm='percent',
                    marginal='box',
                    color_discrete_sequence=["#009dff",
                    "#a7fad5"],
                    title='Distribution of Sex and Age')

# Figure for confusion matrix in length of stay
# -----
z = [[0, 101746, 12331, 9608, 7015, 4255, 2435, 1358, 817, 85],
     [0, 52694, 9073, 10507, 8387, 4798, 2958, 1328, 688, 44],
     [0, 25084, 5746, 6692, 7549, 5902, 3145, 1387, 755, 33],
     [0, 13295, 3788, 4583, 6161, 5727, 3224, 1390, 619, 8],
     [0, 7353, 2652, 3167, 5188, 5204, 2845, 1590, 508, 35],
     [0, 4577, 1928, 2268, 4195, 4659, 2602, 1547, 440, 10],
     [0, 2875, 1581, 1772, 3567, 3851, 2573, 1333, 519, 6],
     [0, 2221, 1132, 1352, 3008, 3267, 2366, 1315, 460, 24],
     [0, 5095, 3292, 5307, 10971, 13066, 9995, 6328, 1875, 68],
     [0, 3374, 3192, 4012, 9208, 12648, 12994, 11404, 3819, 59]]

x = ['<1', '1-2', '2-3', '3-4', '4-5', '5-6', '6-7', '7-8', '8-14', '14>']
y = ['<1', '1-2', '2-3', '3-4', '4-5', '5-6', '6-7', '7-8', '8-14', '14>']

z_text = [[str(y) for y in x] for x in z]

los_cm = ff.create_annotated_heatmap(z, x=x, y=y,
annotation_text=z_text, colorscale='Viridis')

los_cm.update_layout(title_text='<i><b>Confusion matrix</b></i>',

```

```

        xaxis=dict(title=''),
        yaxis=dict(title='y'))

# add custom xaxis title
los_cm.add_annotation(dict(font=dict(color="black", size=14),
                            x=0.5,
                            y=-0.15,
                            showarrow=False,
                            text="Predicted value",
                            xref="paper",
                            yref="paper"))

# add custom yaxis title
los_cm.add_annotation(dict(font=dict(color="black", size=14),
                            x=-0.35,
                            y=0.5,
                            showarrow=False,
                            text="Real value",
                            textangle=-90,
                            xref="paper",
                            yref="paper"))

# adjust margins to make room for yaxis title
los_cm.update_layout(margin=dict(t=50, l=200))

los_cm['data'][0]['showscale'] = True

# -----
# -----

def generate_table(dataframe, max_rows=10):
    return dbc.Table([
        html.Thead([
            html.Tr([html.Th(col) for col in dataframe.columns],
                    style={'color': '#0000A0'})
        ]),
        html.Tbody([
            html.Tr([
                html.Td(dataframe.iloc[i][col]) for col in
                dataframe.columns
            ]) for i in range(min(len(dataframe), max_rows))
        ], style={'bordered': True, 'hover': True, 'striped':
        True, 'dark': True}
        ),
    ])

# Variables to for Table Figure in Nav 2 - Hospital Mortality
# -----
# -----

```

```

tab2_in_cm_train_report["Precision"],
tab2_in_cm_train_report["Recall"], tab2_in_cm_train_report["f1-
score"], \
tab2_in_cm_train_report["Support"] =
tab2_in_cm_train_report["measures,precision,recall,f1-
score,support"].str.split(",", 3).str
tab2_in_cm_train_report =
tab2_in_cm_train_report.drop(['measures,precision,recall,f1-
score,support'], axis=1)

tab2_in_cm_test_report["Precision"],
tab2_in_cm_test_report["Recall"], tab2_in_cm_test_report["f1-
score"], \
tab2_in_cm_test_report["Support"] =
tab2_in_cm_test_report["measures,precision,recall,f1-
score,support"].str.split(",", 3).str
tab2_in_cm_test_report =
tab2_in_cm_test_report.drop(['measures,precision,recall,f1-
score,support'], axis=1)

# Calling Dash Layout for Web Page
# -----
-----
app.layout = html.Div(children=[
    dbc.Row(html.Div(children=[
        dbc.Col(html.Div(
            html.H1(children='MIMIC-III Exploration Dashboard',
                style={
                    'color': '#1c3763',
                    'display': 'inline-block',
                    'float': 'left',
                    'marginLeft': 30,
                    'marginTop': '1%',
                    'marginBottom': '1%',
                    'marginRight': 50,
                    'backgroundColor': 'white',
                    'fontFamily': 'Barlow Semi Condensed',
                    'fontWeight': 610}
                )
            )
        )
    ]),
    style={
        'width': '100%',
        'backgroundColor': 'white',
        'max-width': '1440px'})),

    html.Div(children=[
        dbc.ButtonGroup(children=[ # Use this to edit all
buttons at once

```

```

        dbc.Button('MIMIC Diagnosis Data', id='btn-nclicks-1', n_clicks=0, color="primary", size='lg'),
        dbc.Button('In-Hospital Mortality', id='btn-nclicks-2', n_clicks=0, color="primary", size='lg'),
        dbc.Button('Length-of-Stay', id='btn-nclicks-3', n_clicks=0, color="primary", size='lg'),
    ],
    id='filter-buttons',
    style={
        'width': '85%',
        'maxWidth': '1410px',
        'marginLeft': '240px',
        'fontFamily': 'Barlow Semi Condensed',
        'fontWeight': 610
    },
    className='row')
],
    id='filter-button-container'),
dbc.Row(html.Div(children=[
    dbc.Col(html.Div(children=[
        html.H1('Filters',
            style={
                'color': '#1c3763',
                'marginLeft': '30px',
                'marginTop': '20px',
                'marginBottom': '20px',
                'fontFamily': 'Barlow Semi Condensed',
                'fontWeight': 610}),
        dbc.Col(html.Div(children=[
            dbc.DropdownMenu(
                [
                    dbc.DropdownMenuItem('ETHNICITY',
id='ETHNICITY'),
                    dbc.DropdownMenuItem(divider=True),
                    dbc.DropdownMenuItem("DIAGNOSIS",
id="DIAGNOSIS"),
                    dbc.DropdownMenuItem(divider=True),
                    dbc.DropdownMenuItem('GENDER',
id='GENDER'),
                    dbc.DropdownMenuItem(divider=True),
                    dbc.DropdownMenuItem('AGE', id='AGE'),
                    dbc.DropdownMenuItem(divider=True),
                    dbc.DropdownMenuItem('MORTALITY_INUNIT',
id='MORTALITY_INUNIT'),
                    dbc.DropdownMenuItem(divider=True),
                    dbc.DropdownMenuItem('MORTALITY_INHOSPITAL',
id='MORTALITY_INHOSPITAL'),
                ],
                id='opt',
                label="Filter 1",

```



```

        bs_size="lg",
        color="primary"
    )],
    style={
        'float': 'left',
        'marginLeft': 12}
    )),
    dbc.Col (html.Div (children=[
        dbc.DropdownMenu (
            [
                dbc.DropdownMenuItem ('ETHNICITY',
id='ETHNICITY1'),
                dbc.DropdownMenuItem (divider=True),
                dbc.DropdownMenuItem ("DIAGNOSIS",
id="DIAGNOSIS1"),
                dbc.DropdownMenuItem (divider=True),
                dbc.DropdownMenuItem ('GENDER',
id='GENDER1'),
                dbc.DropdownMenuItem (divider=True),
                dbc.DropdownMenuItem ('AGE', id='AGE1'),
                dbc.DropdownMenuItem (divider=True),
                dbc.DropdownMenuItem ('MORTALITY_INUNIT',
id='MORTALITY_INUNIT1'),
                dbc.DropdownMenuItem (divider=True),
                dbc.DropdownMenuItem ('MORTALITY_INHOSPITAL',
id='MORTALITY_INHOSPITAL1'),
            ],
            id='opt1',
            label="Filter 2",
            bs_size="lg",
            color="primary",
        )],
        style={
            'float': 'left',
            'marginLeft': 12,
            'paddingTop': 20}
    )),
    dbc.Col (html.Div (children=[
        dbc.DropdownMenu (
            [
                dbc.DropdownMenuItem ('ETHNICITY',
id='ETHNICITY2'),
                dbc.DropdownMenuItem (divider=True),
                dbc.DropdownMenuItem ("DIAGNOSIS",
id="DIAGNOSIS2"),
                dbc.DropdownMenuItem (divider=True),
                dbc.DropdownMenuItem ('GENDER',
id='GENDER2'),
                dbc.DropdownMenuItem (divider=True),
                dbc.DropdownMenuItem ('AGE', id='AGE2'),
            ]

```

```

        dbc.DropoutMenuItem(divider=True),
        dbc.DropoutMenuItem('MORTALITY_INUNIT',
id='MORTALITY_INUNIT2'),
        dbc.DropoutMenuItem(divider=True),

dbc.DropoutMenuItem('MORTALITY_INHOSPITAL',
id='MORTALITY_INHOSPITAL2'),
        ],
        id='opt2',
        label="Filter 3",
        bs_size="lg",
        color="primary",
    )],
    style={
        'float': 'left',
        'marginLeft': 12,
        'paddingTop': 20})),
],
    style={
        'width': '15%',
        'marginRight': 0,
        'marginLeft': '4px',
        'height': '100%',
        'display': 'inline-block',
        'backgroundColor': 'white',
        'float': 'left'
    },
    className="one column")), # Is the block for
"Filter" and dropdown setting panel

    dbc.Col(html.Div(children=[
        html.Div(children=[
            html.Div(
                id='container-button-timestamp'
            ),
        ])
    ])),
],
    style={
        'width': '100%',
        'maxWidth': '1410px',
        'backgroundColor': 'white'
    },
    className='container')),
],
    id='full-layout')

@app.callback(Output('container-button-timestamp', 'children'),
              [Input('btn-nclicks-1', 'n_clicks'),
               Input('btn-nclicks-2', 'n_clicks')],

```

```

        Input('btn-nclicks-3', 'n_clicks'),
        Input('ETHNICITY', "n_clicks"),
        Input('DIAGNOSIS', "n_clicks"),
        Input('GENDER', "n_clicks"),
        Input('AGE', "n_clicks"),
        Input('MORTALITY_INUNIT', "n_clicks"),
        Input('MORTALITY_INHOSPITAL', "n_clicks"),
        Input('ETHNICITY1', "n_clicks"),
        Input('DIAGNOSIS1', "n_clicks"),
        Input('GENDER1', "n_clicks"),
        Input('AGE1', "n_clicks"),
        Input('MORTALITY_INUNIT1', "n_clicks"),
        Input('MORTALITY_INHOSPITAL1', "n_clicks"),
        Input('ETHNICITY2', "n_clicks"),
        Input('DIAGNOSIS2', "n_clicks"),
        Input('GENDER2', "n_clicks"),
        Input('AGE2', "n_clicks"),
        Input('MORTALITY_INUNIT2', "n_clicks"),
        Input('MORTALITY_INHOSPITAL2', "n_clicks")]])
# Define Functions for Button Clicks on Filter Page
# -----
-----
def displayClick(btn1, btn2, btn3, x1, x2, x3, x4, x5, x6, y1,
y2, y3, y4, y5, y6, z1, z2, z3, z4, z5, z6):
    changed_id = [p['prop_id'] for p in
dash.callback_context.triggered][0]
    ctx = dash.callback_context
    # Define Nav bar 1 for "MIMIC Datasets"
    # -----
-----
    if 'btn-nclicks-1' in changed_id:
        data = tab1_stays.to_dict('rows')
        columns = [{"name": i, "id": i} for i in
tab1_stays.columns]
        return (html.Div(
            children=[
                dt.DataTable(
                    data=data,
                    columns=columns,
                    page_size=20,
                    style_table={'overflowX': 'auto'},
                    style_cell={
                        'height': 'auto',
                        'maxWidth': '160px',
                        'overflow': 'hidden',
                        'textOverflow': 'ellipsis',
                        'textAlign': 'left',
                        'padding': '10px',
                        'marginLeft': '50xp',
                        'fontFamily': 'Barlow Semi Condensed',
                        'fontSize': '15px'
                    }
                )
            ]
        ))

```

```

    },
    style_cell_conditional=[{
        'if': {'column_id': 'SUBJECT_ID'},
        'textAlign': 'right'
    }],
    style_as_list_view=True,
    filter_action='native',
    style_header={
        'backgroundColor': 'aliceblue',
        'font_family': 'Barlow Semi Condensed',
        'fontSize': '15px',
        'color': '#1c3763',
        'fontWeight': 'bold'
    },
    style_data_conditional=[{
        'if': {'row_index': 'odd'},
        'backgroundColor': 'rgb(248, 248, 248)'
    }],
    ],
    id='table1',
    style={
        'width': '100%',
        'maxWidth': '1209px',
        'height': '30%',
        'marginLeft': 0,
        'float': 'left',
        'display': 'inline-block',
        'marginRight': 0,
        'position': 'absolute'})
# Define Nav bar 2 for "In-Hospital Mortality"
# -----
-----
elif 'btn-nclicks-2' in changed_id:
    data0 = tab2_stays.to_dict('rows')
    data1 = tab2_in_cm_test_report.to_dict('rows')
    data2 = tab2_in_cm_train_report.to_dict('rows')
    columns0 = [{"name": i, "id": i, } for i in
tab2_stays.columns]
    columns1 = [{"name": i, "id": i, } for i in
tab2_in_cm_test_report.columns]
    columns2 = [{"name": i, "id": i, } for i in
tab2_in_cm_train_report.columns]
    # arr = tab2_stays.to_numpy()
    return (
        dbc.Row(
            html.Div(
                children=[
                    dt.DataTable
                    (data=data0,
                     columns=columns0,
                     page_size=7,

```

```

style_table={'overflowX': 'auto'},
style_cell={
    'height': 'auto',
    'maxWidth': '160px',
    'overflow': 'hidden',
    'textOverflow': 'ellipsis',
    'textAlign': 'left',
    'padding': '10px',
    'marginLeft': '50px',
    'fontFamily': 'Barlow Semi
Condensed',
    'fontSize': '15px'
},
style_cell_conditional=[{
    'if': {'column_id': 'SUBJECT_ID'},
    'textAlign': 'right'}],
style_as_list_view=True,
filter_action='native',
style_header={'backgroundColor':
'aliceblue',
    'font_family': 'Barlow
Semi Condensed',
    'font_size': '15px',
    'color': '#1c3763',
    'fontWeight': 'bold'
},
style_data_conditional=[{
    'if': {'row_index': 'odd'},
    'backgroundColor': 'rgb(248, 248,
248)'}],
)],
style={'width': '100%',
    'maxWidth': '1410px', # Styling for
tab 2 page numbers
    'marginLeft': 0,
    'float': 'left',
    'display': 'inline-block',
    'marginRight': 0})
),
# Model Result Block
# -----
-----
dbc.Row(
    html.Div(
        children=[
            dbc.Button("MODEL RESULT",
                color="primary",
                block=True,
                disabled=True,
                style={'color': '#ffffff',

```

```

'font_family': 'Barlow
Semi Condensed',
'fontWeight':
'bold'})),
],
style={'width': '100%'}
),
# Test Score Heatmap
# -----
-----
dbc.Row(
  html.Div(
    children=[
      html.Div(children=[
        dcc.Graph(
          figure={
            'data': [
              go.Heatmap(
                x=['Predicted 0',
'Predicted 1'],
                y=['Actual 1',
'Actual 0'],
                z=[[.1, .4],
[1.0, .07]],
                colorscale='bugn',
                text=[['241',
'133'],
['2772',
'90']]
              )],
            'layout': go.Layout(
              # hovermode="x unified",
              title={
                'text': 'Test Score:
0.446',
                'font_family':
'Barlow Semi Condensed',
                'font_size': 20,
                'x': 0.5,
                'xanchor': 'center',
                'yanchor': 'top'},
            ))
          ]),
style={'float': 'left',
'display': 'inline-block',
'width': '42%',
'marginLeft': '10px'},
className='three columns'),
html.Div(children=[

```

```

        dcc.Graph(
            figure={
                'data': [
                    go.Heatmap(
                        x=['Predicted 0',
'Predicted 1'],
                        y=['Actual 1',
'Actual 0'],
                        z=[[.1, .4],
[1.0, .07]],
                        colorscale=[[0,
[1,
'rgb(224, 243, 248)'],
'rgb(12, 51, 131)']],
                        text=[['759',
[1228'],
[122']]]
                    )],
                'layout': go.Layout(
                    # hovermode="x unified",
                    title={'text': 'Train
Score: 0.736', 'font_family': 'Barlow Semi Condensed',
'font_size': 20,
'x': 0.5,
'xanchor':
'center',
'yanchor':
'top'},
                    xaxis={'ticksuffix':
''},
                    yaxis={'title': ''},
                    # annotation=[['241',
[133'],
[90']]]
                    # [2772',
                    ])
            ], style={'float': 'right',
'display': 'inline-block',
'width': '42%',
'marginLeft': '10px'},
            className='three columns'),
        ],
        id='heatmaps',
        style={'width': '90%',
'max-width': '1410px'},
        className='container')),

```

```
# Test and Train Label
```

```

        dbc.Row(
            html.Div(
                children=[
                    html.Div(children=[
                        html.H4('Test',
                            style={'color': ' #0000A0',
                                'font_family':
'Barlow Semi Condensed'})),
                    dt.DataTable(
                        data=data1,
                        columns=columns1,
                        page_size=20,
                        style_cell={'textAlign': 'left',
                            'padding': '10px',
                            'marginLeft':
'80xp',
                                'font_family':
'Barlow Semi Condensed',
                                'font_size':
'15px'},
                        style_as_list_view=True,
                        filter_action='native',
                        style_header={
                            'backgroundColor':
'aliceblue',
                            'font_family': 'Barlow Semi
Condensed',
                            'font_size': '20px',
                            'color': '#0000A0',
                            'fontWeight': 'bold'},
                        style_data_conditional=[
                            {
                                'if': {'row_index':
'odd'},
                                'backgroundColor':
'rgb(248, 248, 248)'
                            }
                        ]
                    ),
                ],
                style={'float': 'left',
                    'display': 'inline-block',
                    'width': '42%',
                    'marginRight': 100},
                className='three columns'),
            html.Div(children=[
                html.H4('Train',
                    style={'color': ' #0000A0',
                        'font_family':
'Barlow Semi
Condensed'})),

```



```

        dt.DataTable(
            data=data2,
            columns=columns2,
            page_size=20,
            style_cell={'textAlign': 'left',
'padding': '10px', 'marginLeft': '80xp',
'font_family':
'Barlow Semi Condensed',
'font_size':
'15px'},
            style_as_list_view=True,
            filter_action='native',
            style_header={
'backgroundColor':
'aliceblue',
'font_family': 'Barlow Semi
Condensed',
'font_size': '20px',
'color': '#1c3763',
'fontWeight': 'bold'},
            style_data_conditional=[
{
'if': {'row_index':
'odd'},
'backgroundColor':
'rgb(248, 248, 248)'
}
]),
],
style={'float': 'right',
'display': 'inline-block',
'width': '42%', 'marginLeft':
'10px'},
className='three columns'),
],
id='test-train-label',
style={'width': '90%',
'max-width': '1410px'},
className='container')),
)
# Navigation page for tab 3
# -----
elif 'btn-nclicks-3' in changed_id:
    data0 = tab2_stays.to_dict('rows')
    columns0 = [{"name": i, "id": i, } for i in
tab2_stays.columns]
    data3 = tab3_in_cm.to_dict('rows')
    columns3 = [{"name": i, "id": i, } for i in
tab3_in_cm.columns]

```

```

return (
  dbc.Row(
    html.Div(
      children=[
        dt.DataTable(
          data=data0,
          columns=columns0,
          page_size=10,
          style_table={'overflowX': 'auto'},
          style_cell={'textAlign': 'left',
                     'padding': '10px',
                     'marginLeft': '50xp',
                     'font_family': 'Barlow
Semi Condensed',
                               'font_size': '15px'
                               },
          style_as_list_view=True,
          filter_action='native',
          style_header={
            'backgroundColor': 'aliceblue',
            'font_family': 'Barlow Semi
Condensed',
                          'font_size': '15px',
                          'color': '#1c3763',
                          'fontWeight': 'bold'},
          style_cell_conditional=[{
            'if': {'column_id':
'SUBJECT_ID'},
                          'textAlign': 'right'}],
          style_data_conditional=[
            {
              'if': {'row_index': 'odd'},
              'backgroundColor': 'rgb(248,
248, 248)'
            }
          ],
          id='tab3_table',
          style={'width': '100%',
                'maxWidth': '1410px', # Styling for
tab 3 page numbers
                'marginLeft': 0,
                'float': 'left',
                'display': 'inline-block',
                'marginRight': 0})),

        # Model ResultPane
        # -----
        -----
        dbc.Row(
          html.Div(
            children=[

```

```

        dbc.Button("MODEL RESULT",
                    color="primary",
                    block=True,
                    disabled=True,
                    style={'color': '#ffffff',
                           'fontFamily': 'Barlow
Semi Condensed',
                           'fontWeight':
'bold'})),
    ],
    style={
        'width': '100%'
    }
)),

# Model For Table 3
# -----
-----

    dbc.Row(
        html.Div(
            children=[
                html.Div(children=[
                    html.H4('Test Result',
                            style={'color': '#0000A0',
                                   'font_family':
'Barlow Semi Condensed'})),
                    dt.DataTable(
                        data=data3,
                        columns=columns3,
                        page_size=20,
                        style_cell={'textAlign': 'left',
                                   'padding': '10px',
                                   'marginLeft':
'80px',
                                   'minWidth': '100px',
                                   'width': '100px',
                                   'whiteSpace':
'normal',
                                   'height': 'auto',
                                   'maxWidth': '100px',
                                   'font_family':
'Barlow Semi Condensed',
                                   'font_size':
'15px'}),
                        style_as_list_view=True,
                        filter_action='native',
                        style_header={
                            'height': 'auto',
                            'backgroundColor':
'aliceblue',

```

```

Condensed',
'font_family': 'Barlow Semi
'font_size': '15px',
'color': '#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{
'if': {'row_index':
'backgroundColor':
'rgb(248, 248, 248)'}
}
])
],
style={'float': 'left',
'display': 'inline-block',
'width': '42%',
'paddingTop': '30px',
'marginRight': 30},
className='three columns'),
html.Div(children=[
dcc.Graph(
id='los_cm',
figure=los_cm
),
], style={'float': 'right',
'display': 'inline-block',
'width': '55%',
'paddingTop': '30px',
'marginLeft': '10px'},
className='three columns'),
],
id='los',
style={'width': '100%',
'max-width': '1410px'},
className='container'))
)

```

```

# This triggers the whole button filters in the Side Panel
# -----
-----

```

```

elif ctx.triggered:
    button_id = ctx.triggered[0]["prop_id"].split(".")[0]
    # First Filter Button for Ethnicity
    if button_id in ['ETHNICITY']:
        df = tab2_stays[['ETHNICITY']]
        data = tab2_stays.to_dict('rows')
        columns = [{"name": i, "id": i, } for i in (df)]

```

```

        return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'fontWeight': 'bold'},
style_data_conditional=[
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
], )],
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})

elif button_id in ['DIAGNOSIS']:
    df = tab2_stays[['DIAGNOSIS']]
    data = tab2_stays.to_dict('rows')
    columns = [{"name": i, "id": i, } for i in (df)]

    return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',

```

```

'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'color':
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{
'if':
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
}
], )],
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})

elif button_id in ['GENDER']:
df = tab2_stays[['GENDER']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]

return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',

```

```

'font_size': '15px'}
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
],
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})
elif button_id in ['AGE']:
df = tab2_stays[['AGE']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]
return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
style_as_list_view=True,

```

```

filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'fontWeight': 'bold'},
style_data_conditional=[
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)']
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0}
elif button_id in ['MORTALITY_INUNIT']:
df = tab2_stays[['MORTALITY_INUNIT']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]
return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}],
style_as_list_view=True,
filter_action='native',

```



```

style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'fontWeight': 'bold'},
'color': '#0000A0',
style_data_conditional=[
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)']
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})
elif button_id in ['MORTALITY_INHOSPITAL']:
df = tab2_stays[['MORTALITY_INHOSPITAL']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]
return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',

```

```

'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)']
style={
'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0}),
elif button_id in ['ETHNICITY1']:
df = tab2_stays[['ETHNICITY']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]
return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',

```

```

'color':
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{
    'if':
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
}
], )],
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})

elif button_id in ['DIAGNOSIS1']:
df = tab2_stays[['DIAGNOSIS']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]

return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'color':
'#0000A0',

```

```

'fontWeight': 'bold'},
style_data_conditional=[
    {
        'if':
        {'row_index': 'odd'},
        'backgroundColor': 'rgb(248, 248, 248)'
    }
], )],
style={'width': '20%',
       'height': '30%',
       'marginLeft': 0,
       'float': 'left',
       'display': 'inline-block',
       'marginRight': 0})

elif button_id in ['GENDER1']:
    df = tab2_stays[['GENDER']]
    data = tab2_stays.to_dict('rows')
    columns = [{"name": i, "id": i, } for i in (df)]

    return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'color':
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[

```

```

{
    'if':
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
}
], )],
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})

elif button_id in ['AGE1']:
    df = tab2_stays[['AGE']]
    data = tab2_stays.to_dict('rows')
    columns = [{"name": i, "id": i, } for i in (df)]

    return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'color':
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{
    'if':
{'row_index': 'odd'},

```

```

'backgroundColor': 'rgb(248, 248, 248)'
    ], )],
    style={'width': '20%', 'height':
'30%', 'marginLeft': 0, 'float': 'left',
    'display': 'inline-block',
'marginRight': 0})

    elif button_id in ['MORTALITY_INUNIT1']:
        df = tab2_stays[['MORTALITY_INUNIT']]
        data = tab2_stays.to_dict('rows')
        columns = [{"name": i, "id": i, } for i in (df)]

        return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'color':
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{
'if':
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
}
], )],
    style={'width': '20%',

```

```

        'height': '30%',
        'marginLeft': 0,
        'float': 'left',
        'display': 'inline-block',
        'marginRight': 0})

    elif button_id in ['MORTALITY_INHOSPITAL1']:
        df = tab2_stays[['MORTALITY_INHOSPITAL']]
        data = tab2_stays.to_dict('rows')
        columns = [{"name": i, "id": i, } for i in (df)]

        return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,

style_cell={'textAlign': 'left', 'padding': '10px',

'marginLeft': '50xp',

'font_family': 'Barlow Semi Condensed',

'font_size': '15px'}

,

style_as_list_view=True,

filter_action='native',

style_header={

'backgroundColor': 'aliceblue',

'font_family': 'Barlow Semi Condensed',

'font_size': '15px',

'color':

'#0000A0',

'fontWeight': 'bold'},

style_data_conditional=[

{

'if':

{'row_index': 'odd'},

'backgroundColor': 'rgb(248, 248, 248)'

}

], )],

style={'width': '20%',

'height': '30%',

'marginLeft': 0,

'float': 'left',

'display': 'inline-block',

```

```

'marginRight': 0})

# Button Filter for Ethnicity
elif button_id in ['ETHNICITY2']:
    df = tab2_stays[['ETHNICITY']]
    data = tab2_stays.to_dict('rows')
    columns = [{"name": i, "id": i, } for i in (df)]

    return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'color':
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{
'if':
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
}
], )],
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})

```



```

# Button Filter for Diagnosis
elif button_id in ['DIAGNOSIS2']:
    df = tab2_stays[['DIAGNOSIS']]
    data = tab2_stays.to_dict('rows')
    columns = [{"name": i, "id": i, } for i in (df)]

    return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,

style_cell={'textAlign': 'left', 'padding': '10px',

'marginLeft': '50xp',

'font_family': 'Barlow Semi Condensed',

'font_size': '15px'}

,

style_as_list_view=True,

filter_action='native',

style_header={

'backgroundColor': 'aliceblue',

'font_family': 'Barlow Semi Condensed',

'font_size': '15px',

'color':

'#0000A0',

'fontWeight': 'bold'}},

style_data_conditional=[

{

'if':

{'row_index': 'odd'},

'backgroundColor': 'rgb(248, 248, 248)'

}

], )],

style={'width': '20%',

'height': '30%',

'marginLeft': 0,

'float': 'left',

'display': 'inline-block',

'marginRight': 0})

# Button Filter for Genders
elif button_id in ['GENDER2']:

```

```

df = tab2_stays[['GENDER']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]

return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'color':
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{
'if':
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
}
], ),
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})

# Button Filter for Age
elif button_id in ['AGE2']:
df = tab2_stays[['AGE']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]

```

```

        return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'color':
'#0000A0',
'fontWeight': 'bold'},
style_data_conditional=[
{
'if':
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
}
], )],
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})

# Button Filter for Mortality in Unit
elif button_id in ['MORTALITY_INUNIT2']:
    df = tab2_stays[['MORTALITY_INUNIT']]
    data = tab2_stays.to_dict('rows')
    columns = [{"name": i, "id": i, } for i in (df)]

    return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,

```

```

style_cell={'textAlign': 'left', 'padding': '10px',
'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,

style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'fontWeight': 'bold'},
'color':
'#0000A0',
style_data_conditional=[
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
},
], ),
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})

# Button Filter for Mortality in Hospital
elif button_id in ['MORTALITY_INHOSPITAL2']:
df = tab2_stays[['MORTALITY_INHOSPITAL']]
data = tab2_stays.to_dict('rows')
columns = [{"name": i, "id": i, } for i in (df)]

return html.Div(children=[dt.DataTable(data=data,
columns=columns, page_size=20,
style_cell={'textAlign': 'left', 'padding': '10px',

```

```

'marginLeft': '50xp',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px'}
,
style_as_list_view=True,
filter_action='native',
style_header={
'backgroundColor': 'aliceblue',
'font_family': 'Barlow Semi Condensed',
'font_size': '15px',
'fontWeight': 'bold'},
'color':
'#0000A0',
style_data_conditional=[
{'row_index': 'odd'},
'backgroundColor': 'rgb(248, 248, 248)'
], )],
style={'width': '20%',
'height': '30%',
'marginLeft': 0,
'float': 'left',
'display': 'inline-block',
'marginRight': 0})
# Shows landing page
# -----
-----
else:
    data = tab2_stays.to_dict('rows')
    columns = [{"name": i, "id": i, } for i in
(tab2_stays.columns)]
    return (
        dbc.Row(
            html.Div(
                children=[
                    dt.DataTable(
                        data=data,
                        columns=columns,

```

```

        page_size=10,
        style_table={'overflowX': 'auto'},
        style_cell={'textAlign': 'left',
                    'padding': '10px',
                    'marginLeft': '50xp',
                    'font_family': 'Barlow
Semi Condensed',
                    'font_size': '15px'
                    },
        style_as_list_view=True,
        style_cell_conditional=[{
'SUBJECT_ID'},
            'if': {'column_id':
                    'textAlign': 'right'}}],
        filter_action='native',
        style_header={
            'backgroundColor': 'aliceblue',
            'font_family': 'Barlow Semi
Condensed',
            'font_size': '15px',
            'color': '#0000A0',
            'fontWeight': 'bold'},
        style_data_conditional=[
            {
                'if': {'row_index': 'odd'},
                'backgroundColor': 'rgb(248,
248, 248)'
            }
        ],
    ],
    style={'width': '100%',
          'maxWidth': '1410px',
          'marginLeft': 0,
          'float': 'left',
          'display': 'inline-block',
          'marginRight': 0})
),
# Diagrams Name
dbc.Row(
    html.Div(
        children=[
            dbc.Button("MODEL RESULT",
                       color="primary",
                       block=True,
                       disabled=True,
                       style={'color': '#ffffff',
                              'font_family': 'Barlow
Semi Condensed',
                              'fontWeight':
'bold'})),
        ],
        style={'width': '100%',

```

```

        'maxWidth': '1410px'})
    ),
    # Graph for Diagnosis
    # -----
-----
    dbc.Row(
        html.Div(
            dcc.Graph(
                id='diag',
                figure=fig
            ),
            style={
                'width': '100%',
                'maxWidth': '1440px'
            }
        ),
    ),
    dbc.Row(
        html.Div(
            dcc.Graph(
                id='age_sex',
                figure=histo
            ),
            style={
                'width': '100%',
                'maxWidth': '1440px'
            }
        )
    )
)

```

```

# App Callbacks to make the tabs Interactive
@app.callback(Output('btn-nclicks-1', 'style'),
              [Input('btn-nclicks-1', 'n_clicks'),
               Input('btn-nclicks-2', 'n_clicks'),
               Input('btn-nclicks-3', 'n_clicks')])
def change_button_style(n_clicks, n_clicks1, n_clicks2):
    changed_id = [p['prop_id'] for p in
dash.callback_context.triggered][0]
    if 'btn-nclicks-1' in changed_id:
        return white_button_style

```

```

@app.callback(Output('btn-nclicks-2', 'style'),
              [Input('btn-nclicks-1', 'n_clicks'),
               Input('btn-nclicks-2', 'n_clicks'),
               Input('btn-nclicks-3', 'n_clicks')])
def change_button_style(n_clicks, n_clicks1, n_clicks2):
    changed_id = [p['prop_id'] for p in
dash.callback_context.triggered][0]

```

```

    if 'btn-nclicks-2' in changed_id:
        return white_button_style

@app.callback(Output('btn-nclicks-3', 'style'),
              [Input('btn-nclicks-1', 'n_clicks'),
               Input('btn-nclicks-2', 'n_clicks'),
               Input('btn-nclicks-3', 'n_clicks')])
def change_button_style(n_clicks, n_clicks1, n_clicks2):
    changed_id = [p['prop_id'] for p in
dash.callback_context.triggered][0]
    if 'btn-nclicks-3' in changed_id:
        return white_button_style

if __name__ == '__main__':
    app.run_server(debug=True)

```