STATS 202: DATA MINING AND ANALYSIS FINAL

Instructor: Linh Tran

Final Project

Due Date: August 2, 2023

Stanford University

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Kaggle Reference. Team Name: Adam Kainikara.

Work done by Adam Kainikara.

1 Treatment Effect. Using Python I loaded the data from all available CSV sets. Created a function called find_patients which took the input of the loaded data and returned every unique patient id. After collecting every patient id, I created a blank dictionary. The dictionary would have the patient id as the key. For values, I stored all the associated values of each patient in an array. For patients that had one or more visit days, the information would also be stored. The following is one example of how the data would be stored for a particular patient.

dtype=[('study', '<U10'), (country, '<U10'), ('txgroup', '<U10'), ('assesmentid', '<f8'), ('patientid', '<f8'), ('visitday', '<i4'), ('xvalues', '<f8', (30,)), ('panss', '<f8'), ('leadstatus', '<U10')]), 30951.0: array([('"C"', '"China"', '"Control"', 304958., 30951., 0, [4., 3., 2., 1., 3., 4., 2., 4., 5., 4., 5., 4., 5., 3., 1., 2., 1., 1., 4., 1., 4., 4., 3., 3., 4., 6., 3., 1., 5., 2.], 94., '"Assign to'), ('"C"', '"China"', '"Control"', 301327., 30951., 7, [4., 3., 2., 1., 2., 4., 2., 4., 5., 4., 5., 4., 5., 3., 1., 1., 1., 1., 4., 1., 4., 4., 3., 3., 4., 6., 3., 1., 4., 2.], 91., '"Assign to'), ('"C"', '"China"', '"Control"', 303725., 30951., 14, [4., 3., 2., 1., 1., 4., 2., 4., 5., 4., 5., 4., 5., 3., 1., 1., 1., 1., 4., 2., 4., 4., 3., 3., 4., 6., 3., 1., 4., 2.], 91., '"Passed"'), ('"C"', '"China"', '"Control"', 304954., 30951., 42, [4., 3., 4., 1., 1., 4., 2., 4., 5., 5., 5., 4., 5., 3., 1., 2., 1., 1., 4., 3., 4., 4., 3., 3., 4., 6., 3., 2., 5., 2.], 98., '"Passed"'), ('"C"', '"China"', '"Control"', 307645., 30951., 70, [4., 2., 2., 3., 1., 4., 3., 5., 5., 6., 6., 5., 5., 4., 2., 3., 1., 1., 4., 3., 5., 5., 3., 3., 4., 6., 3., 3., 5., 2.], 108., '"Passed"')

After this was done, patients were further filtered into a control group and a treatment group. In addition, patients with fewer than one visit day were removed. Below is a graph of treatment and control groups. The X axis is visit day and Y axis is PANSS score. To reduce clutter on the graph, only Study E is shown.

After, I calculated the difference between each patient's final PANNS score and their initial score. The distribution of the difference in scores can be seen bellow.

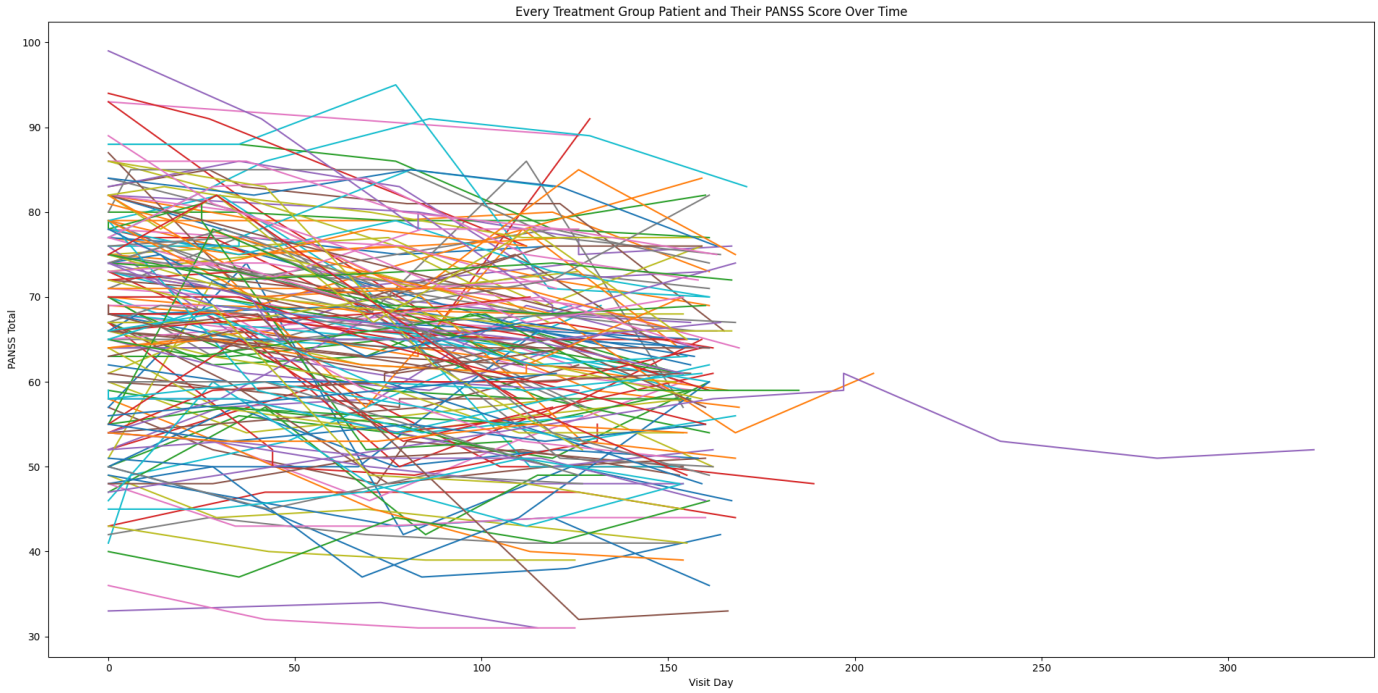


FIGURE 0.1. Treatment Group

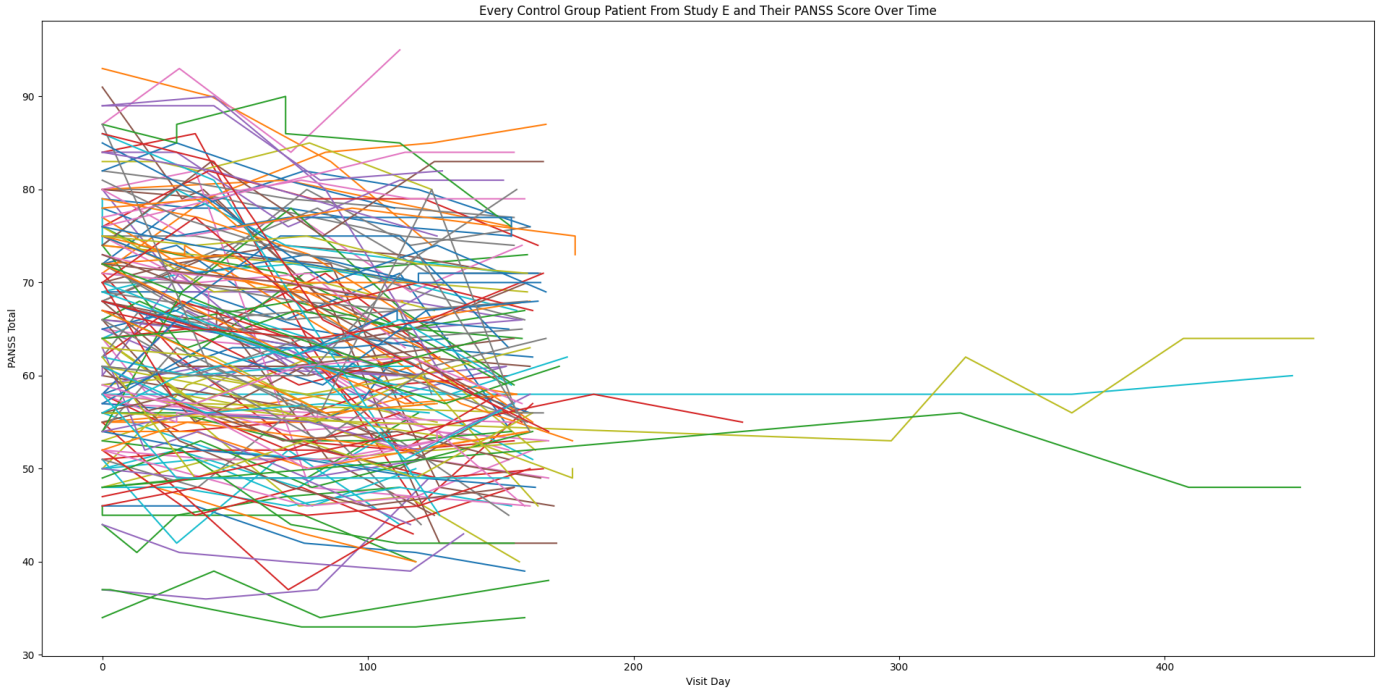


FIGURE 0.2. Control Group

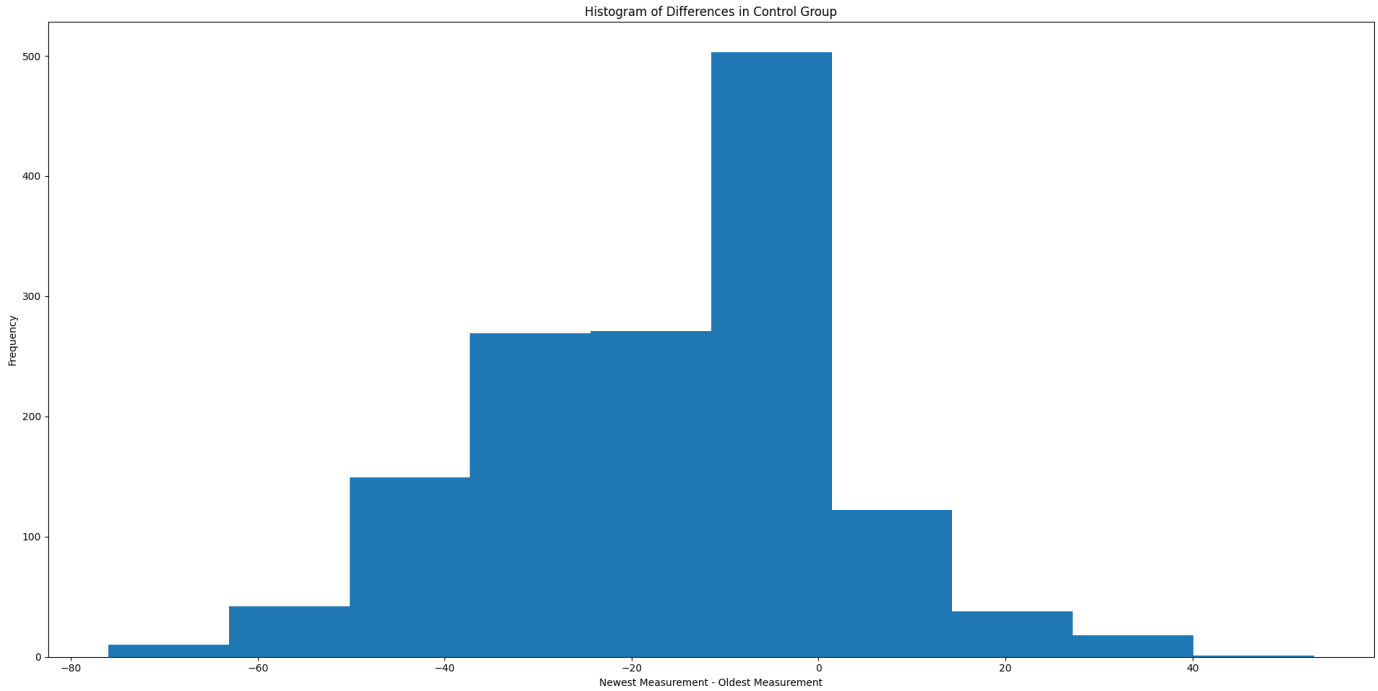


FIGURE 0.3. Differences in Control Group

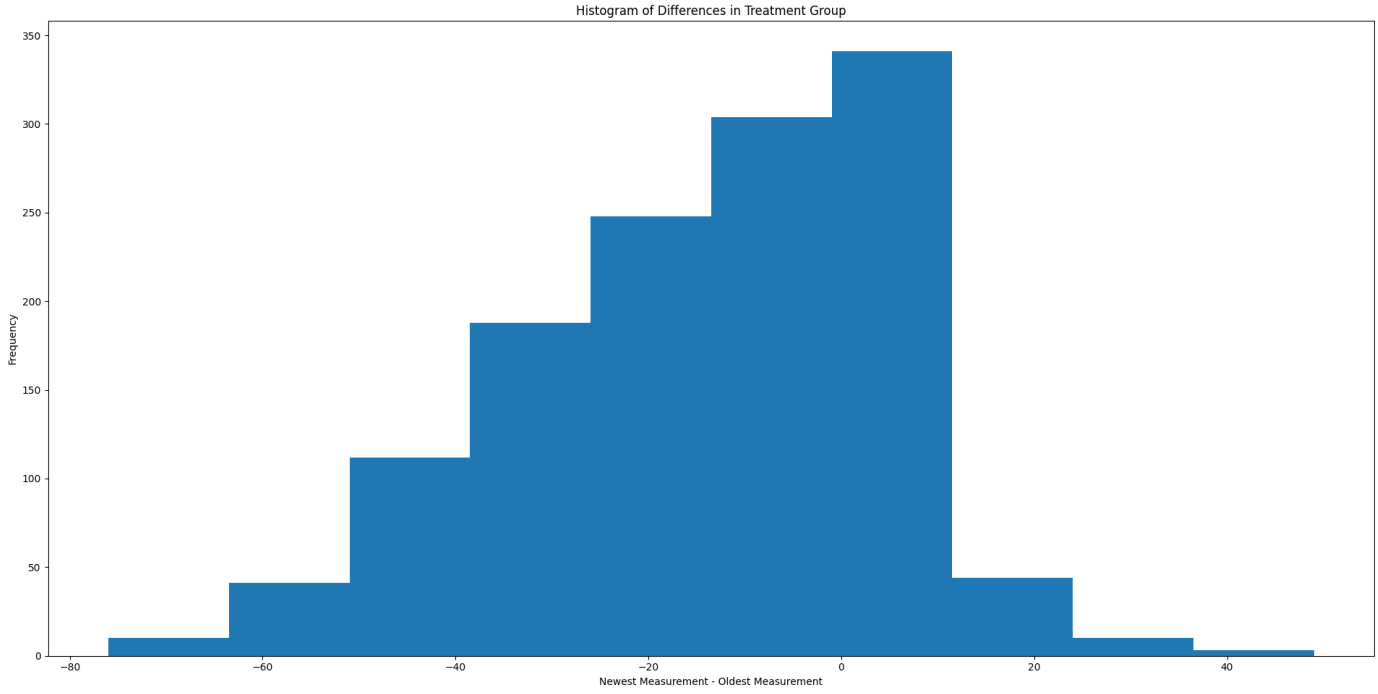


FIGURE 0.4. Differences in Treatment Group

The mean difference of patients in the control group was -16.13 with a standard deviation of 19.40. That is on average, patients in the control group on average saw their PANSS score decrease by 16 at the end of the study. The mean difference of patients in the treatment group was -15.98 with a standard deviation of 19.24.

To see if the treatment had any affect, a t test will be conducted where H_0 is there is no significant difference in the mean change in score between the treatment and control groups and H_a is there is a significant difference in the mean change in score between the treatment and control groups. With $\alpha = 0.05$ a test statistic of 0.2105 and 2945 egress of freedom. The p value is 0.416. The result is not significant. The treatment affect does not make a significant impact on the decrease in PANSS Scores.

2 Patient Segmentation. For this problem I used two forms of clustering. One being k means clustering, and the other being hierarchical clustering so I could see a dendogram. For K means clustering, I began by clustering on various components that made up the PANSS score. For example I tried P1 with N1, P1 with G1 etc. I choose 3 centrists to fill. I also used cross validation to choose the best k. In the scatter plot below, it shows the locations of the centorids of the clusters when clustering using P1 and N1. The centroids are in orange. The blue dots are the xvalue combinations with P1 and N1.

I then wanted to know what a dendogram of this would look like, so one was constructed. There where two clusters as I clustered by P1 and N1. One thing I found quite interesting is that one of the clusters (green one which is N1) is larger than the other cluster. I thought that the clusters would be of similar size. Also the clusters don't meet for a while.

3 Forecasting.

Kaggle Reference. Team Name: Adam Kainikara.

Work done by Adam Kainikara.

There were many data wrangling steps involved in this part of the project. I used data from set E for this part. At first, I created a structured array which was organized by study, country, txgroup, assesment id, patient id, visit day, xvalues, panss score and lead status. Then I made a function that found every unique patient id. This was to figure out how many patients were in the study. After collecting every patient id, I created a blank dictionary. The dictionary would have the patient id as the key. For values, I stored all the associated values of each patient in an array. For patients that had one or more visit days, the information would also be stored. Patients who did not have at least two visit days where removed from the study. In a spread sheet, I did some basic calculations. These calculations included figuring out the average % change for each patient from the initial reading in the study to the last reading in the study.

For my first prediction, I uploaded a csv that contained the most recent visit day. I did this to see if the most recent visit PANSS score would be similar to the real 18th week score. This resulted in a score of 6.61736.

My next submission was the score decreased by the average decrease calculated in the spread sheet. To do this, I calculated each patients change in PANSS score from the first week to the last week. I found the average % drop across the data set. I then changed the initial reading (first day's reading) by the average % change.The average % change across the data set was about 10%. This did not change my score significantly.

For my final submission I used linear interpolation where:

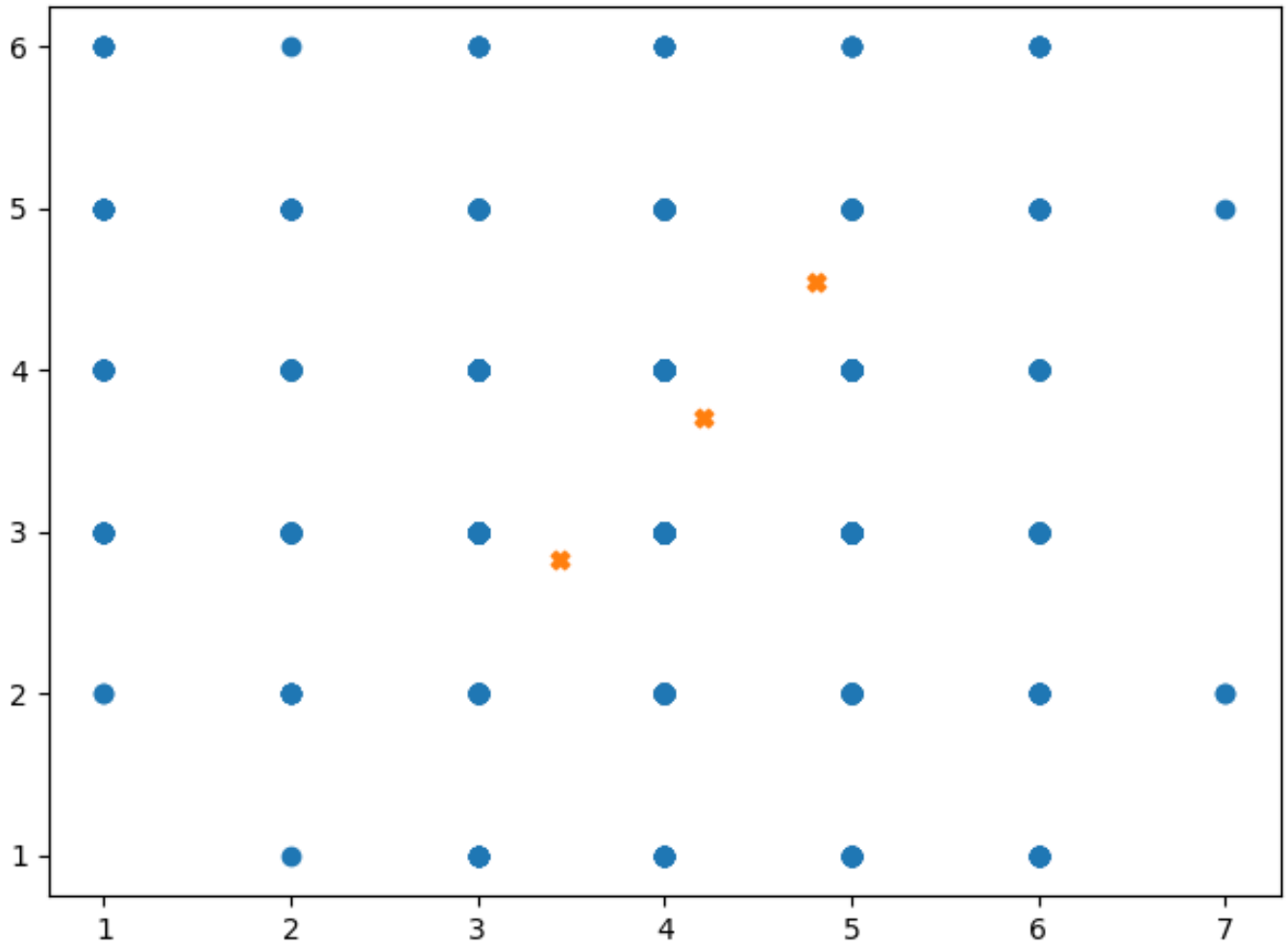


FIGURE 0.5. K Means Clustering

$\alpha = \frac{x-x_1}{x_2-x_1}$ *where* $0 \leq \alpha \leq 1$ and
$\hat{y} = (1-\alpha)y_1 + \alpha y_2$

For this, I used day 126 as the time to solve for the 18th week score. Although day 126 does not necessarily mean that that was the patient's 18th week, I assumed that most patients would be around this time period. This assumption was made because I calculated the average number of weeks each patient was in the study for based off of visit days and this was 17 weeks. The resulted csv that I submitted changed my score to

4 Binary Classification.

Kaggle Reference. Team Name: Adam Kainikara.

Work done by Adam Kainikara.

There were many data wrangling steps involved in this part of the project. I used data sets A,B,C for training and E as the test set. At first, I created a structured array which was organized by study, country, txgroup, assesment id, patient id, visit day, xvalues, panss score and lead status. Then made a function that found every unique patient id. This was to figure out how many patients were in the study. After collecting every patient id, I created

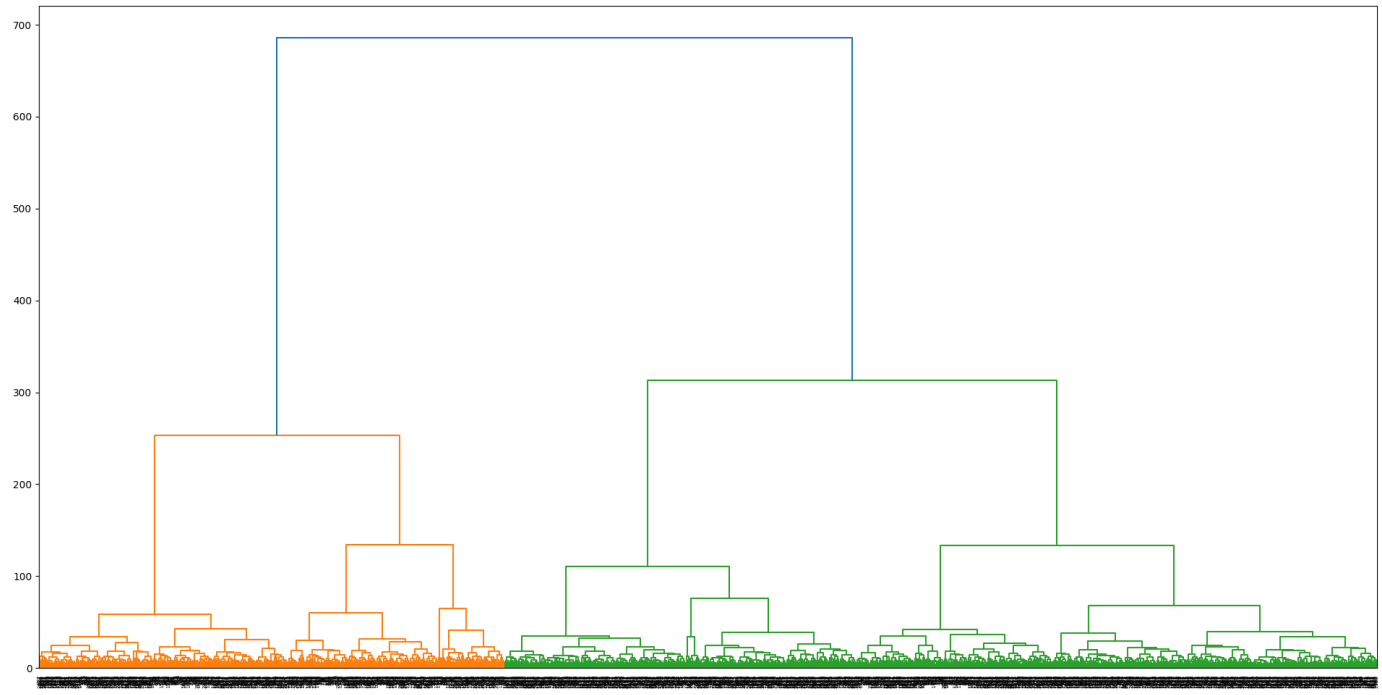


FIGURE 0.6. Dendogram

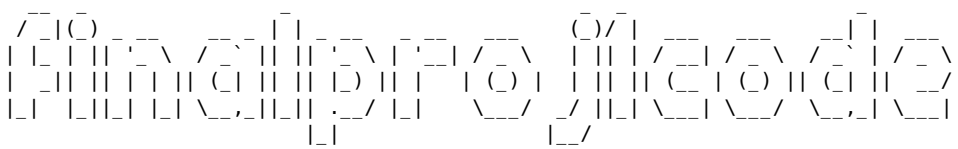
a blank dictionary. The dictionary would have the patient id as the key. For values, I stored all the associated values of each patient in an array. For patients that had one or more visit days, the information would also be stored. Patients who did not have at least two visit days where removed from the study.

For classification I used many classification methods. These include knn, naive bayes, decision trees, random forests and support vector machines. When using KNN, I choose many different K values, however my accuracy was never very good so I decided not to use it. My score on kaggle was 16.

I then used naieve bayes and decision trees, both of which did not improve my score. My best attempt was using a support vector machine. The support vector machine gave me a score of 1.3.

K=4, Knn, The training accuracy is: 0.8297683204622479 The test accuracy is 0.15366350067842605.

Support vector machine, The training accuracy is: 0.7872659592199567 The test accuracy is 0.621438263229307



finalp3.py

```
from numpy import *
import sys
import matplotlib.pyplot as plt
patient_visit_dt = dtype([('study','U10'),('country','U10'),('txgroup','U10'),('patientid', float64),('visitday', int32),('xvalues',float64,(30)),('panss',float64)])

def data_loader(fname):
    #Study      Country PatientID       SiteID  RaterID AssessmentID    TxGroup VisitDay        P1      P2      P3      P4      P5      P6      P7      N1      N2      N3      N4      N5
    # 0     1       2           3       4       5               6       7           8   9   10  11  12  13  14  15  16  17  18  19  20  21  22  23  24  25  26  27
    # G7        G8      G9      G10     G11     G12     G13     G14     G15     G16     PANSS_Total
    # 28    29  30  31  32  33  34  35  36  37  38

    #data_a = loadtxt(fname,skiprows=1, usecols=(0,1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37,38), delimiter=',')
    cata_data_a = loadtxt(fname,skiprows=1, usecols=(0,1,6), delimiter=',', dtype=str )
    quant_data_a = loadtxt(fname,skiprows=1, usecols=(2,3,4,5,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37,38), delimiter=',', dtype=float64)
    #print(cata_data_a, quant_data_a)
    #print(cata_data_a.shape, quant_data_a.shape)
    #print(type(cata_data_a), type(quant_data_a))


    data_a = empty(quant_data_a.shape[0], dtype=patient_visit_dt)
    #print(data_a.shape)

    #raise SystemExit
    #visit_a = array((visit_a[['study', 'country', 'txgroup']],visit_a['xvalues'],visit_a['yvalues']), dtype = patient_visit_dt)
    #x_v = quant_data_a[:,5:35]

    #y_v = quant_data_a[:,35]
    #print(data_a[['study', 'country', 'txgroup']].shape, cata_data_a.shape)
    data_a['study']  = cata_data_a[:, 0]
    data_a['country']  = cata_data_a[:, 1]
    data_a['txgroup']  = cata_data_a[:, 2]
    data_a['patientid'] = quant_data_a[:,0]
    data_a['visitday'] = quant_data_a[:,4]
    #print("eeeee", quant_data_a[:,4])
    #print("aaaa", quant_data_a[:,0])

    #print(quant_data_a[:,5:35])
    #print(quant_data_a[:,35])

    data_a['xvalues'] = quant_data_a[:,5:35]
    data_a['panss'] = quant_data_a[:,35]
    #print(data_a['xvalues'].shape, data_a['panss'].shape)
    #print(data_a)

    #data_a[['study', 'country', 'txgroup']] = tuple(cata_data_a[:, i] for i in range(3))
    #print('!!', cata_data_a[:2])
    #data_a[['study', ] = cata_data_a[0, :2]
    #cata_data_a[:, :2]

    #visit_a['xvalues'] = x_v
    #visit_a['yvalues'] = y_v
    #visit_a[['country', 'study']] = cata_data_a[:,0], cata_data_a[:,1]
    return data_a, cata_data_a, quant_data_a

def trial():
    num_l = []
    for i in range(100):
        num_l.append(i)
        #print(i)
    return num_l

def find_patients(data_a):
    d = {}
    #rows = range(data_a.shape[0])
    for i in range(data_a.shape[0]):
        key = data_a[i]['patientid']
        if key in d:
            d[key].append(i)
        else:
            d[key] = [i]

    patient_d = {}

    for patientid, index_l in d.items():
        patient_a = data_a[index_l]
        index_v = patient_a['visitday'].argsort()
        patient_d[patientid] = patient_a[index_v]

    print(type(patient_d))
    print("test", patient_d)
    return patient_d

def seperate_control_treatment(patient_d):
    control_l = []
    treatment_l = []

    for patient_a in patient d.values():
        if patient_a[0]['txgroup'] == '"Control"':
            control_l.append(patient_a)
        else:
            treatment_l.append(patient_a)
    #print(control_l, treatment_l)
    return control_l, treatment_l

def plot_patient_data(data_l):
    print(type(data_l))
    print(len(data_l))

    fig, ax = plt.subplots()

    for patient_a in data_l:

        x_l = patient_a['visitday']
        y_l = patient_a['panss']
        ax.plot(x l, y l)
    plt.xlabel('Visit Day')
    plt.ylabel('PANSS Total')
    plt.title('Every Treatment Group Patient and Their PANSS Score Over Time')
    plt.show()
def filter_patients(patient_d, day_limit=126):
    print(type(patient_d))
    print(patient_d)
    #delete_patient_l = [patientid for patientid, patient_a in patient_d.items() if patient_a[-1]['visitday'] < day_limit]
    delete_patient_l = [patientid for patientid, patient_a in patient_d.items() if patient_a[-1]['visitday'] - patient_a[0]['visitday'] < day_limit]
    for patientid in delete_patient_l:
        del patient_d[patientid]
def fit_linearmodel():
    pass
def main():
    #data_a = hstack([data_loader(fname) for fname in sys.argv[1:]])

    data_a,cata_data_a, quant data a = data_loader("Study_E.csv")
    patientid_v = data_a['patientid']
    upatientid v = unique(patientid_v)
    print('Patient id:', upatientid_v.shape[0])

    print("just viewing", patientid_v)
    #print(da)

```

```
    # day_v = data_a['visitday']
    # # uday_v = unique(day_v)
    # week_v = uday_v/7
    # print(uday_v)
    # print('------------')
    # print('------------')
    # print(week_v)
    # print('------------')

    # print(f'data a:10 {data_a[10:]}')
    m_v = data_a['txgroup'] ==   '"Treatment'
    day_v = data a[m v]['visitday']
    print('treatment day', sorted(day_v))
    uday_v = unique(day_v)
    # week_v = uday_v//7
    # print(f'uday_v treatments: {uday_v}')
    # print(f'week_v treatments: {week_v}')
    # #week_v = int32(week_v)
    # print('week count:', list(zip(arange(week_v.shape[0]), bincount(week_v))))
    patient d = find patients(data_a)
    print('Before delete:', len(patient_d))
    filter_patients(patient_d, day_limit=105)
    print('After delete:', len(patient_d))

    #print(len(a))
    print('------------')
    #print(d)

    control_l, treatment_l = seperate_control_treatment(patient_d)
    print("control", control_l)
    print('------------')
    print('------------')
    print('------------')
    print('------------')
    print('------------')
    print("treatment", treatment_l)

    plot_patient_data(treatment_l)
    plot_patient_data(control_l)
    raise SystemExit
    plt.xlabel('Visit Day')
    plt.ylabel('PANSS Total')
    plt.title('Every Treatment Group Patient and Their PANSS Score Over Time')
if __name__ == '__main__':
    main()
```

p3class.py

```
from cgi import test
from numpy import *
import sys
import matplotlib.pyplot as plt
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive_bayes import CategoricalNB
from sklearn.naive_bayes import GaussianNB
from sklearn import tree
from sklearn.ensemble import RandomForestClassifier
from sklearn import svm
patient_visit_dt = dtype([('study','U10'),('country','U10'),('txgroup','U10'),('assesmentid', float64),('patientid', float64),('visitday', int32),('xvalues',float64,(31)),('panss',float64),

def data_loader(fname):
    #Study      Country PatientID       SiteID  RaterID AssessmentID    TxGroup VisitDay        P1      P2      P3      P4      P5      P6      P7      N1      N2      N3      N4      N5
    # 0     1       2           3       4       5               6       7           8   9   10  11  12  13  14  15  16  17  18  19  20  21  22  23  24  25  26  27
    # G7        G8      G9      G10     G11     G12     G13     G14     G15     G16     PANSS_Total
    # 28    29  30  31  32  33  34  35  36  37  38

    if 'Study_E.csv' in fname:
        cata_data_a = loadtxt(fname,skiprows=1, usecols=(0,1,6), delimiter=',', dtype=str )
    else:
        cata_data_a = loadtxt(fname,skiprows=1, usecols=(0,1,6,39), delimiter=',', dtype=str )
    quant_data_a = loadtxt(fname,skiprows=1, usecols=(2,3,4,5,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37,38), delimiter=',', dtype=float64)



    data_a = empty(quant_data_a.shape[0], dtype=patient_visit_dt)
    #print(data_a.shape)


    data_a['study']  = cata_data_a[:, 0]
    data_a['country']  = cata_data_a[:, 1]
    data_a['txgroup']  = cata_data_a[:, 2]
    data_a['assesmentid'] = quant_data_a[:,3]
    data_a['patientid'] = quant_data_a[:,0]
    data_a['visitday'] = quant_data_a[:,4]


    data_a['xvalues'] = quant_data_a[:,5:36]
    data_a['panss'] = quant_data_a[:,35]
    if 'Study_E.csv' in fname:
        data_a['leadstatus'] =  0
    else:
        data_a['leadstatus'] = cata_data_a[:,3]
    return data_a

def trial():
    num_l = []
    for i in range(100):
        num_l.append(i)
    return num_l

def find_patients(data_a):
    d = {}
    for i in range(data_a.shape[0]):
        key = data_a[i]['patientid']
        if key in d:
            d[key].append(i)
        else:
            d[key] = [i]

    patient_d = {}
    for patientid, index_l in d.items():
        patient_a = data_a[index_l]
        index_v = patient_a['visitday'].argsort()
        patient_d[patientid] = patient_a[index_v]

    print(type(patient_d))
    return patient_d

def seperate_control_treatment(patient_d):
    control_d = {}
    treatment_d = {}

    for patient_a in patient_d.values():

        if patient_a[0]['txgroup'] == '"Control"':
            control_d[patient_a[0]['patientid']] = patient_a
        else:
            #print('test')
            treatment_d[patient_a[0]['patientid']] = patient_a

    return control_d, treatment_d


def plot_patient_data(data_l):
    print(type(data_l))
    print(len(data_l))

    fig, ax = plt.subplots()

    for patient_a in data_l:

        x_l = patient_a['visitday']
        y_l = patient_a['panss']
        ax.plot(x_l, y_l)
    plt.show()

def filter_patients(patient_d, day_limit=126):
    print(type(patient_d))
    print(patient_d)
    #delete_patient_l = [patientid for patientid, patient_a in patient_d.items() if patient_a[-1]['visitday'] < day_limit]
    delete_patient_l = [patientid for patientid, patient_a in patient_d.items() if patient_a[-1]['visitday'] - patient_a[0]['visitday'] < day_limit]
    for patientid in delete_patient_l:
        del patient_d[patientid]

def difference_in_fields(patient_d, field_name):
    difference_values_l = []
    for patient_a in patient_d.values():
        dif1 = patient_a[-1][field_name]
        dif2 = patient_a[0][field_name]
        dif = dif1-dif2
        #dif = patient_a[-1]['panss'] - patient_a[1]['panss']
        difference_values_l.append(dif)
    difference_values_a = array(difference_values_l)
    return difference_values_a

def difference_in_scores(patient_d):
    assert 0
    difference_values_l = []
    for patient_a in patient_d.values():
        dif1 = patient_a[-1]['panss']
        dif2 = patient_a[0]['panss']
        dif = dif1-dif2
        difference_values_l.append(dif)
    return difference_values_l

def difference_in_scores_stats(difference_values_l):
    mean_difference = mean(difference_values_l)
```

```
    standev_difference = std(difference_values_l)
    print(f'The mean difference is {mean_difference} and has a standard deviation of {standev_difference}')
    return mean_difference, standev_difference

def knn_pred(xtrain_a, ytrain_v, xtest_a, ytest_v):
    knn_model = KNeighborsClassifier(n_neighbors = 5)
    knn_model.fit(xtrain_a, ytrain_v)
    ytrainpred_v = knn_model.predict(xtrain_a)
    ypred_v = knn_model.predict(xtest_a)
    ypred_proba_a = knn_model.predict_proba(xtest_a)
    accuracy_train = ((ytrain_v==ytrainpred_v).sum())/ytrainpred_v.shape[0]
    accuracy_test = ((ytest_v==ypred_v).sum())/ytest_v.shape[0]

def class_pred(clf, xtrain_a, ytrain_v, xtest_a, ytest_v):
    clf.fit(xtrain_a, ytrain_v)
    ytrainpred_v = clf.predict(xtrain_a)
    ypred_v = clf.predict(xtest_a)
    ypred_proba_a = clf.predict_proba(xtest_a)
    accuracy_train = ((ytrain_v==ytrainpred_v).sum())/ytrainpred_v.shape[0]
    accuracy_test = ((ytest_v==ypred_v).sum())/ytest_v.shape[0]



    # m_v = ypred_v == ytest_v
    # print('match samples:', ypred_v[m_v])
    # print('mismatch samples:', list(zip(ypred_v[logical_not(m_v)], ytest_v[logical_not(m_v)])))

    print(f'The training accuracy is: {accuracy_train} ')
    print(f'The test accuracy is {accuracy_test}')
    return ypred_proba_a

def z_score_conver(data_a):
    train_data_a = data_a['xvalues']
    #print(train_data_a)
    mean_a = train_data_a.mean(axis=0)
    standev_a = train_data_a.std(axis=0)
    zscore = (train_data_a - mean_a)/standev_a
    #print(zscore)
    return zscore

def z_score_convert2(data_a, input_test_data_a):
    train_data_a = data_a['xvalues']
    test_data_a = input_test_data_a['xvalues']
    #print(train_data_a)
    train_mean_v = train_data_a.mean(axis=0)
    train_standev_v = train_data_a.std(axis=0)
    train_zscore_a = (train_data_a - train_mean_v)/train_standev_v
    test_zscore_a = (test_data_a - train_mean_v)/train_standev_v
    #print(zscore)
    return train_zscore_a, test_zscore_a

def time_shifter(patient_d):
    #adjusted_d = {}
    for patient_id, visitday in patient_d.items():
        time_zero = visitday[0]['visitday']
        visitday[0] = 0
        for other_days in visitday[1:]:
            other_days['visitday']-=time_zero



def desired_data(patient_d):
    predicted_data_l = []
    for patient_id, array_values_a in patient_d.items():
        last_array = array_values_a[-1]
        output_data = last_array['panss']
        predicted_data_l.append((patient_id, output_data))
    submitted_data_a = array(predicted_data_l, output_data)
    return submitted_data_a

def class_data_merge(test_data_a, maxproba_v):
    #print("see assesment shapes", test_data_a['assesmentid'].shape)
    class_proba_l = list(zip(test_data_a['assesmentid'],maxproba_v))
    class_proba_a = array (class_proba_l)

    return class_proba_a

def save_file(patient_data_a, fname):
    savetxt(fname, patient_data_a, delimiter=',')

def main():
    data_a = hstack([data_loader(fname) for fname in sys.argv[1:]])
    train_data_a = hstack([data_loader(fname) for fname in sys.argv[1:-1]])
    test_data_a = data_loader(sys.argv[-1])

    m_v = logical_not(train_data_a['leadstatus'] == '"Passed"')
    train_data_a = train_data_a[m_v]
    m_v = logical_not(test_data_a['leadstatus'] == '"Passed"')
    test_data_a = test_data_a[m_v]


    # print("seeing train after load", train_data_a.shape)
    #print("seeing test after load", test_data_a.shape)

    patientid_v = data_a['patientid']
    upatientid_v = unique(patientid_v)
    #print('Patient id:', upatientid_v.shape[0])

    #print("just viewing", patientid_v)

    #print(da)
    patient_d = find_patients(data_a)

    #print('------------')

    control_d, treatment_d = seperate_control_treatment(patient_d)
    #print("view control people", control_d)
    #print(type(control_d))
    #print("view treatment people", treatment_d)
    #print(type(control_d))
    #print('------------')

    control_patientdiff = difference_in_fields(control_d, 'panss')
    control_patientdiffstats = difference_in_scores_stats(control_patientdiff)
    #print(type(control_patientdiff))

    treatment_patientdiff = difference_in_fields(treatment_d, 'panss')
    treatment_patientdiffstats = difference_in_scores_stats(treatment_patientdiff)

    #print('------------')


    print(f'Control Patients: {control_patientdiffstats}')
    print(f'Treatment Patients: {treatment_patientdiffstats}')

    print('------------')
    #print(patient_d)
    fname = "upload1.csv"
    upload_data_a = desired_data(patient_d)
    print(upload_data_a)
    save_file(upload_data_a, fname)

    print('------------')
    zscore_train_a, zscore_test_a = z_score_convert2(train_data_a, test_data_a)
```

```
264
265
266
267
268     #ypredproba_a = knn_pred(zscore_train_a, train_data_a['leadstatus'], zscore_test_a, test_data_a['leadstatus'])
269     #ypredproba_a = knn_pred(train_data_a['xvalues'], train_data_a['leadstatus'], test_data_a['xvalues'], test_data_a['leadstatus'])
270
271     #clf = KNeighborsClassifier(n_neighbors = 5)
272     #clf = CategoricalNB(force_alpha=True)
273     #clf = GaussianNB()
274     #clf = tree.DecisionTreeClassifier(max_depth=8, min_samples_leaf=10, ccp_alpha=0.005, criterion='entropy')
275     #clf = RandomForestClassifier(max_depth=8, min_samples_leaf=10, ccp_alpha=0.005, random_state=0)
276     clf = svm.SVC(probability=True)
277     ypredproba_a = class_pred(clf, train_data_a['xvalues'], train_data_a['leadstatus'], test_data_a['xvalues'], test_data_a['leadstatus'])
278
279
280     print(ypredproba a)
281     print('------------')
282
283     max_proba_v = ypredproba_a.max(axis=1)
284     class_data_a = class_data_merge(test_data_a, max_proba_v)
285
286     fname = "classupload1.csv"
287     save_file(class_data_a, fname)
288
289     raise SystemExit
290
291     # day_v = data_a['visitday']
292     # # uday_v = unique(day_v)
293     # week_v = uday_v/7
294     # print(uday_v)
295     # print('------------')
296     # print('------------')
297     # print(week_v)
298     # print('------------')
299
300     # print(f'data_a:10 {data_a[10:]}')
301     m_v = data_a['txgroup'] ==   '"Treatment'
302     day_v = data a[m v]['visitday']
303     print('treatment day', sorted(day_v))
304     uday_v = unique(day_v)
305     # week_v = uday_v//7
306     # print(f'uday_v treatments: {uday_v}')
307     # print(f'week_v treatments: {week_v}')
308     # #week_v = int32(week_v)
309     # print('week count:', list(zip(arange(week_v.shape[0]), bincount(week_v))))
310     patient d = find patients(data_a)
311     print('Before delete:', len(patient_d))
312     filter_patients(patient_d, day_limit=105)
313     print('After delete:', len(patient_d))
314
315     #print(len(a))
316     print('------------')
317     #print(d)
318     print(patient_d)
319
320     control_d, treatment_d = seperate_control_treatment(patient_d)
321     #print("control", control_l)
322     print('------------')
323     print('------------')
324     print('------------')
325     print('------------')
326     print('------------')
327     #print("treatment", treatment_l)
328     #plot_patient_data(control_l)
329     #plot_patient_data(treatment_l)
330 if __name__ == '__main__':
331     main()
```

prob2.py

```
from cgi import test
from numpy import *
import sys
import matplotlib.pyplot as plt
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive_bayes import CategoricalNB
from sklearn.naive_bayes import GaussianNB
from sklearn import tree
from sklearn.ensemble import RandomForestClassifier
from sklearn import svm
from sklearn.cluster import KMeans
from scipy.cluster.hierarchy import complete, fcluster
from scipy.cluster import hierarchy
from scipy.spatial.distance import pdist
from sklearn.model_selection import KFold

patient_visit_dt = dtype([('study','U10'),('country','U10'),('txgroup','U10'),('assesmentid', float64),('patientid', float64),('visitday', int32),('xvalues',float64,(31)),('panss',float64),

def data_loader(fname):
    #Study      Country PatientID       SiteID  RaterID AssessmentID    TxGroup VisitDay        P1      P2      P3      P4      P5      P6      P7      N1      N2      N3      N4      N5
    # 0     1     2         3       4       5              6       7          8   9   10  11  12  13  14  15  16  17  18  19  20  21  22  23  24  25  26  27
    # G7        G8      G9      G10     G11     G12     G13     G14     G15     G16     PANSS_Total
    # 28    29  30  31  32  33  34  35  36  37  38

    if 'Study_E.csv' in fname:
        cata_data_a = loadtxt(fname,skiprows=1, usecols=(0,1,6), delimiter=',', dtype=str )
    else:
        cata_data_a = loadtxt(fname,skiprows=1, usecols=(0,1,6,39), delimiter=',', dtype=str )
    quant_data_a = loadtxt(fname,skiprows=1, usecols=(2,3,4,5,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37,38), delimiter=',', dtype=float64)



    data_a = empty(quant_data_a.shape[0], dtype=patient_visit_dt)
    #print(data_a.shape)


    data_a['study']  = cata_data_a[:, 0]
    data_a['country']  = cata_data_a[:, 1]
    data_a['txgroup']  = cata_data_a[:, 2]
    data_a['assesmentid'] = quant_data_a[:,3]
    data_a['patientid'] = quant_data_a[:,0]
    data_a['visitday'] = quant_data_a[:,4]


    data_a['xvalues'] = quant_data_a[:,5:36]
    data_a['panss'] = quant_data_a[:,35]
    if 'Study_E.csv' in fname:
        data_a['leadstatus'] =  0
    else:
        data_a['leadstatus'] = cata_data_a[:,3]
    return data_a

def trial():
    num_l = []
    for i in range(100):
        num_l.append(i)
    return num_l

def find_patients(data_a):
    d = {}
    for i in range(data_a.shape[0]):
        key = data_a[i]['patientid']
        if key in d:
            d[key].append(i)
        else:
            d[key] = [i]

    patient_d = {}
    for patientid, index_l in d.items():
        patient_a = data_a[index_l]
        index_v = patient_a['visitday'].argsort()
        patient_d[patientid] = patient_a[index_v]

    print(type(patient_d))
    return patient_d

def seperate_control_treatment(patient_d):
    control_d = {}
    treatment_d = {}

    for patient_a in patient_d.values():

        if patient_a[0]['txgroup'] == '"Control"':
            control_d[patient_a[0]['patientid']] = patient_a
        else:
            #print('test')
            treatment_d[patient_a[0]['patientid']] = patient_a

    return control_d, treatment_d


def plot_patient_data(data_l):
    print(type(data_l))
    print(len(data_l))

    fig, ax = plt.subplots()

    for patient_a in data_l:

        x_l = patient_a['visitday']
        y_l = patient_a['panss']
        ax.plot(x_l, y_l)
    plt.show()

def filter_patients(patient_d, day_limit=126):
    print(type(patient_d))
    print(patient_d)
    #delete_patient_l = [patientid for patientid, patient_a in patient_d.items() if patient_a[-1]['visitday'] < day_limit]
    delete_patient_l = [patientid for patientid, patient_a in patient_d.items() if patient_a[-1]['visitday'] - patient_a[0]['visitday'] < day_limit]
    for patientid in delete_patient_l:
        del patient_d[patientid]

def difference_in_fields(patient_d, field_name):
    difference_values_l = []
    for patient_a in patient_d.values():
        dif1 = patient_a[-1][field_name]
        dif2 = patient_a[0][field_name]
        dif = dif1-dif2
        #dif = patient_a[-1]['panss'] - patient_a[1]['panss']
        difference_values_l.append(dif)
    difference_values_a = array(difference_values_l)
    return difference_values_a

def difference_in_scores(patient_d):
    assert 0
    difference_values_l = []
    for patient_a in patient_d.values():
        dif1 = patient_a[-1]['panss']
        dif2 = patient_a[0]['panss']
```

```
        dif = dif1-dif2
        difference_values_l.append(dif)
    return difference_values_l

def difference_in_scores_stats(difference_values_l):
    mean_difference = mean(difference_values_l)
    standev difference = std(difference_values_l)
    print(f'The mean difference is {mean_difference} and has a standard deviation of {standev_difference}')
    return mean_difference, standev_difference

def knn_pred(xtrain_a, ytrain_v, xtest_a, ytest_v):
    knn_model = KNeighborsClassifier(n_neighbors = 5)
    knn_model.fit(xtrain_a, ytrain_v)
    ytrainpred_v = knn_model.predict(xtrain_a)
    ypred_v = knn_model.predict(xtest_a)
    ypred_proba_a = knn_model.predict_proba(xtest_a)
    accuracy_train = ((ytrain_v==ytrainpred_v).sum())/ytrainpred_v.shape[0]
    accuracy_test = ((ytest_v==ypred_v).sum())/ytest_v.shape[0]

def class_pred(clf, xtrain_a, ytrain_v, xtest_a, ytest_v):
    clf.fit(xtrain_a, ytrain_v)
    ytrainpred_v = clf.predict(xtrain_a)
    ypred_v = clf.predict(xtest_a)
    ypred_proba_a = clf.predict_proba(xtest_a)
    accuracy_train = ((ytrain_v==ytrainpred_v).sum())/ytrainpred_v.shape[0]
    accuracy_test = ((ytest_v==ypred_v).sum())/ytest_v.shape[0]



    # m_v = ypred_v == ytest_v
    # print('match samples:', ypred_v[m_v])
    # print('mismatch samples:', list(zip(ypred_v[logical_not(m_v)], ytest_v[logical_not(m_v)])))

    print(f'The training accuracy is: {accuracy_train} ')
    print(f'The test accuracy is {accuracy_test}')
    return ypred_proba_a

def z_score_conver(data_a):
    train_data_a = data_a['xvalues']
    #print(train_data_a)
    mean_a = train_data_a.mean(axis=0)
    standev_a = train_data_a.std(axis=0)
    zscore = (train_data_a - mean_a)/standev_a
    #print(zscore)
    return zscore

def z_score_convert2(data_a, input test_data_a):
    train_data_a = data_a['xvalues']
    test_data_a = input_test_data_a['xvalues']
    #print(train_data_a)
    train_mean_v = train_data_a.mean(axis=0)
    train_standev_v = train_data_a.std(axis=0)
    train_zscore_a = (train_data_a - train_mean_v)/train_standev_v
    test_zscore_a = (test_data_a - train_mean_v)/train_standev_v
    #print(zscore)
    return train_zscore_a, test_zscore_a

def time_shifter(patient_d):
    #adjusted_d = {}
    for patient_id, visitday in patient d.items():
        time_zero = visitday[0]['visitday']
        visitday[0] = 0
        for other_days in visitday[1:]:
            other_days['visitday']-=time_zero



def desired_data(patient_d):
    predicted_data_l = []
    for patient_id, array_values_a in patient_d.items():
        last_array = array_values a[-1]
        output_data = last_array['panss']
        predicted_data_l.append((patient_id, output_data))
    submitted_data_a = array(predicted_data_l, output_data)
    return submitted_data_a

def class_data_merge(test_data_a, maxproba_v):
    #print("see assesment shapes", test_data_a['assesmentid'].shape)
    class_proba_l = list(zip(test_data_a['assesmentid'],maxproba_v))
    class_proba_a = array (class_proba_l)

    return class_proba_a

def get_first_day(patient_d):
    return array([patient_a[0]['xvalues'] for patient_a in patient_d.values()])


def k_means_clustering(x, n_clusters):
    #best_k = max(range(2, 11), key=lambda k: np.mean(cross_val_score(KMeans(n_clusters=k), cv=10)))
    kf = KFold(n_splits=10, shuffle=True, random_state=42)
    kmeans = KMeans(n_clusters=n_clusters)
    kmeans.fit(x)
    cluster_labels = kmeans.labels_
    cluster_centers = kmeans.cluster_centers_
    plt.scatter(x[:, 0], x[:, 7],  marker='o')
    print("x", x)
    #In this case x:,0 and X:,7 are the p1 and n1 values
    plt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:, 1], marker='X')
    plt.show()
    return cluster_labels, cluster_centers
def dendo_construct(x_a):
    linkage_matrix = hierarchy.linkage(x_a, method='ward')

    plt.figure(figsize=(10, 6))
    dendrogram = hierarchy.dendrogram(linkage_matrix)
    plt.show()
def save_file(patient_data_a, fname):
    savetxt(fname, patient_data_a, delimiter=',')

def main():
    data_a = hstack([data_loader(fname) for fname in sys.argv[1:]])
    train_data_a = hstack([data_loader(fname) for fname in sys.argv[1:-1]])
    test_data_a = data_loader(sys.argv[-1])

    m_v = logical_not(train_data_a['leadstatus'] == '"Passed"')
    train_data_a = train_data_a[m_v]
    m_v = logical_not(test_data_a['leadstatus'] == '"Passed"')
    test_data_a = test_data_a[m_v]


    # print("seeing train after load", train_data_a.shape)
    #print("seeing test after load", test_data_a.shape)

    patientid_v = data_a['patientid']
    upatientid_v = unique(patientid_v)
    #print('Patient id:', upatientid_v.shape[0])

    #print("just viewing", patientid_v)

    #print(da)
    patient_d = find_patients(data_a)

    #print('------------')

```

```
    control_d, treatment_d = seperate_control_treatment(patient_d)
    #print("view control people", control_d)
    #print(type(control_d))
    #print("view treatment people", treatment_d)
    #print(type(control_d))
    #print('------------')

    control_patientdiff = difference_in_fields(control_d, 'panss')
    control_patientdiffstats = difference_in_scores_stats(control_patientdiff)
    #print(type(control_patientdiff))

    treatment_patientdiff = difference_in_fields(treatment_d, 'panss')
    treatment_patientdiffstats = difference_in_scores_stats(treatment_patientdiff)

    #print('------------')


    print(f'Control Patients: {control_patientdiffstats}')
    print(f'Treatment Patients: {treatment_patientdiffstats}')

    print('------------')
    #print(patient_d)
    fname = "upload1.csv"
    upload_data_a = desired_data(patient_d)
    print(upload_data_a)
    save_file(upload_data_a, fname)

    print('------------')
    zscore_train_a, zscore_test_a = z_score_convert2(train_data_a, test_data_a)
    first_x_a = get_first_day(patient_d)
    cluster labels, cluster_centers = k_means_clustering(first_x_a,3)
    print("cluster labels:", cluster_labels)
    print("cluster centers:", cluster_centers)

    #ypredproba_a = knn_pred(zscore_train_a, train_data_a['leadstatus'], zscore_test_a, test_data_a['leadstatus'])
    #ypredproba_a = knn_pred(train_data_a['xvalues'], train_data_a['leadstatus'], test_data_a['xvalues'], test_data_a['leadstatus'])

    #clf = KNeighborsClassifier(n_neighbors = 5)
    #clf = CategoricalNB(force_alpha=True)
    #clf = GaussianNB()
    #clf = tree.DecisionTreeClassifier(max_depth=8, min_samples_leaf=10, ccp_alpha=0.005, criterion='entropy')
    #clf = RandomForestClassifier(max_depth=8, min_samples_leaf=10, ccp_alpha=0.005, random_state=0)
    clf = svm.SVC(probability=True)
    ypredproba_a = class_pred(clf, train_data_a['xvalues'], train_data_a['leadstatus'], test_data_a['xvalues'], test_data_a['leadstatus'])


    print(ypredproba a)
    print('------------')

    max_proba_v = ypredproba_a.max(axis=1)
    class_data_a = class_data_merge(test_data_a, max_proba_v)

    fname = "classupload1.csv"
    save_file(class_data_a, fname)

    dendo_construct(get_first_day(patient_d))

    raise SystemExit

    # day_v = data_a['visitday']
    # # uday_v = unique(day_v)
    # week_v = uday_v/7
    # print(uday_v)
    # print('------------')
    # print('------------')
    # print(week_v)
    # print('------------')

    # print(f'data_a:10 {data_a[10:]}')
    m_v = data_a['txgroup'] ==   '"Treatment'
    day_v = data a[m v]['visitday']
    print('treatment day', sorted(day_v))
    uday_v = unique(day_v)
    # week_v = uday_v//7
    # print(f'uday_v treatments: {uday_v}')
    # print(f'week_v treatments: {week_v}')
    # #week_v = int32(week_v)
    # print('week count:', list(zip(arange(week_v.shape[0]), bincount(week_v))))
    patient d = find patients(data_a)
    print('Before delete:', len(patient_d))
    filter_patients(patient_d, day_limit=105)
    print('After delete:', len(patient_d))

    #print(len(a))
    print('------------')
    #print(d)
    print(patient_d)

    control_d, treatment_d = seperate_control_treatment(patient_d)
    #print("control", control_l)
    print('------------')
    print('------------')
    print('------------')
    print('------------')
    print('------------')
    #print("treatment", treatment_l)
    #plot_patient_data(control_l)
    #plot_patient data(treatment_l)
if __name__ == '__main__':
    main()
```

realp3.py

```
from cgi import test
from numpy import *
import sys
import matplotlib.pyplot as plt
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive_bayes import CategoricalNB
from sklearn.naive_bayes import GaussianNB
from sklearn import tree
from sklearn.ensemble import RandomForestClassifier

patient_visit_dt = dtype([('study','U10'),('country','U10'),('txgroup','U10'),('assesmentid', float64),('patientid', float64),('visitday', int32),('xvalues',float64,(30)),('panss',float64),

def data_loader(fname):
    #Study      Country PatientID       SiteID  RaterID AssessmentID    TxGroup VisitDay        P1      P2      P3      P4      P5      P6      P7      N1      N2      N3      N4      N5
    # 0     1       2           3       4       5               6       7           8   9   10  11  12  13  14  15  16  17  18  19  20  21  22  23  24  25  26  27
    # G7        G8      G9      G10     G11     G12     G13     G14     G15     G16     PANSS_Total
    # 28    29  30  31  32  33  34  35  36  37  38

    if 'Study_E.csv' in fname:
        cata_data_a = loadtxt(fname,skiprows=1, usecols=(0,1,6), delimiter=',', dtype=str )
    else:
        cata_data_a = loadtxt(fname,skiprows=1, usecols=(0,1,6,39), delimiter=',', dtype=str )
    quant_data_a = loadtxt(fname,skiprows=1, usecols=(2,3,4,5,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37,38), delimiter=',', dtype=float64)



    data_a = empty(quant_data_a.shape[0], dtype=patient_visit_dt)
    #print(data_a.shape)


    data_a['study']  = cata_data_a[:, 0]
    data_a['country']  = cata_data_a[:, 1]
    data_a['txgroup']  = cata_data_a[:, 2]
    data_a['assesmentid'] = quant_data_a[:,3]
    data_a['patientid'] = quant_data_a[:,0]
    data_a['visitday'] = quant_data_a[:,4]


    data_a['xvalues'] = quant_data_a[:,5:35]
    data_a['panss'] = quant_data_a[:,35]
    if 'Study_E.csv' in fname:
        data_a['leadstatus'] =  0
    else:
        data_a['leadstatus'] = cata_data_a[:,3]
    return data_a

def trial():
    num_l = []
    for i in range(100):
        num_l.append(i)
    return num_l

def find_patients(data_a):
    d = {}
    for i in range(data_a.shape[0]):
        key = data_a[i]['patientid']
        if key in d:
            d[key].append(i)
        else:
            d[key] = [i]

    patient_d = {}
    for patientid, index_l in d.items():
        patient_a = data_a[index_l]
        index_v = patient_a['visitday'].argsort()
        patient_d[patientid] = patient_a[index_v]

    print(type(patient_d))
    return patient_d

def seperate_control_treatment(patient_d):
    control_d = {}
    treatment_d = {}

    for patient_a in patient_d.values():

        if patient_a[0]['txgroup'] == '"Control"':
            control_d[patient_a[0]['patientid']] = patient_a
        else:
            #print('test')
            treatment_d[patient_a[0]['patientid']] = patient_a

    return control_d, treatment_d


def plot_patient_data(data_l):
    print(type(data_l))
    print(len(data_l))

    fig, ax = plt.subplots()

    for patient_a in data_l:

        x_l = patient_a['visitday']
        y_l = patient_a['panss']
        ax.plot(x_l, y_l)
    plt.show()

def filter_patients(patient_d, day_limit=126):
    print(type(patient_d))
    print(patient_d)
    #delete_patient_l = [patientid for patientid, patient_a in patient_d.items() if patient_a[-1]['visitday'] < day_limit]
    delete_patient_l = [patientid for patientid, patient_a in patient_d.items() if patient_a[-1]['visitday'] - patient_a[0]['visitday'] < day_limit]
    for patientid in delete_patient_l:
        del patient_d[patientid]

def difference_in_fields(patient_d, field_name):
    difference_values_l = []
    for patient_a in patient_d.values():
        dif1 = patient_a[-1][field_name]
        dif2 = patient_a[0][field_name]
        dif = dif1-dif2
        #dif = patient_a[-1]['panss'] - patient_a[1]['panss']
        difference_values_l.append(dif)
    difference_values_a = array(difference_values_l)
    return difference_values_a

def difference_in_scores(patient_d):
    assert 0
    difference_values_l = []
    for patient_a in patient_d.values():
        dif1 = patient_a[-1]['panss']
        dif2 = patient_a[0]['panss']
        dif = dif1-dif2
        difference_values_l.append(dif)
    return difference_values_l

def difference_in_scores_stats(difference_values_l):
    mean_difference = mean(difference_values_l)
```

```
    standev difference = std(difference_values_l)
    print(f'The mean difference is {mean_difference} and has a standard deviation of {standev_difference}')
    return mean_difference, standev_difference

def knn_pred(xtrain_a, ytrain_v, xtest_a, ytest_v):
    knn_model = KNeighborsClassifier(n_neighbors = 5)
    knn_model.fit(xtrain_a, ytrain_v)
    ytrainpred_v = knn_model.predict(xtrain_a)
    ypred_v = knn_model.predict(xtest_a)
    ypred_proba_a = knn_model.predict_proba(xtest_a)
    accuracy_train = ((ytrain_v==ytrainpred_v).sum())/ytrainpred_v.shape[0]
    accuracy_test = ((ytest_v==ypred_v).sum())/ytest_v.shape[0]

def class_pred(clf, xtrain_a, ytrain_v, xtest_a, ytest_v):
    clf.fit(xtrain_a, ytrain_v)
    ytrainpred_v = clf.predict(xtrain_a)
    ypred_v = clf.predict(xtest_a)
    ypred_proba_a = clf.predict_proba(xtest_a)
    accuracy_train = ((ytrain_v==ytrainpred_v).sum())/ytrainpred_v.shape[0]
    accuracy_test = ((ytest_v==ypred_v).sum())/ytest_v.shape[0]



    # m_v = ypred_v == ytest_v
    # print('match samples:', ypred_v[m_v])
    # print('mismatch samples:', list(zip(ypred_v[logical_not(m_v)], ytest_v[logical_not(m_v)])))

    print(f'The training accuracy is: {accuracy_train} ')
    print(f'The test accuracy is {accuracy_test}')
    return ypred_proba_a

def z_score_conver(data_a):
    train_data_a = data_a['xvalues']
    #print(train_data_a)
    mean_a = train_data_a.mean(axis=0)
    standev_a = train_data_a.std(axis=0)
    zscore = (train_data_a - mean_a)/standev_a
    #print(zscore)
    return zscore

def z_score_convert2(data_a, input_test_data_a):
    train_data_a = data_a['xvalues']
    test_data_a = input_test_data_a['xvalues']
    #print(train_data_a)
    train_mean_v = train_data_a.mean(axis=0)
    train_standev_v = train_data_a.std(axis=0)
    train_zscore_a = (train_data_a - train_mean_v)/train_standev_v
    test_zscore_a = (test_data_a - train_mean_v)/train_standev_v
    #print(zscore)
    return train_zscore_a, test_zscore_a

def time_shifter(patient_d):
    #adjusted_d = {}
    for patient_id, visitday in patient_d.items():
        time_zero = visitday[0]['visitday']
        visitday[0] = 0
        for other_days in visitday[1:]:
            other_days['visitday']-=time_zero



def desired_data(patient_d):
    predicted_data_l = []
    for patient_id, array_values_a in patient_d.items():
        last_array = array_values_a[-1]
        output_data = last_array['panss']
        predicted_data_l.append((patient_id, output_data))
    submitted_data_a = array(predicted_data_l, output_data)
    return submitted_data_a

def class_data_merge(test_data_a, maxproba_v):
    #print("see assesment shapes", test_data_a['assesmentid'].shape)
    class_proba_l = list(zip(test_data_a['assesmentid'],maxproba_v))
    class_proba_a = array (class_proba_l)

    return class_proba_a

def save_file(patient_data_a, fname):
    savetxt(fname, patient_data_a, delimiter=',')

def main():
    data_a = hstack([data_loader(fname) for fname in sys.argv[1:]])
    train_data_a = hstack([data_loader(fname) for fname in sys.argv[1:-1]])
    test_data_a = data_loader(sys.argv[-1])

    m_v = logical_not(train_data_a['leadstatus'] == '"Passed"')
    train_data_a = train_data_a[m_v]
    m_v = logical_not(test_data_a['leadstatus'] == '"Passed"')
    test_data_a = test_data_a[m_v]


    # print("seeing train after load", train_data_a.shape)
    #print("seeing test after load", test_data_a.shape)

    patientid_v = data_a['patientid']
    upatientid_v = unique(patientid_v)
    #print('Patient id:', upatientid_v.shape[0])

    #print("just viewing", patientid_v)

    #print(da)
    patient_d = find_patients(data_a)

    #print('------------')

    control_d, treatment_d = seperate_control_treatment(patient_d)
    #print("view control people", control_d)
    #print(type(control_d))
    #print("view treatment people", treatment_d)
    #print(type(control_d))
    #print('------------')

    control_patientdiff = difference_in_fields(control_d, 'panss')
    plt.hist(control_patientdiff, bins = 10)
    plt.xlabel('Newest Measurement - Oldest Measurement')
    plt.ylabel('Frequency')
    plt.title('Histogram of Differences in Control Group')
    plt.show()
    control_patientdiffstats = difference_in_scores_stats(control_patientdiff)
    #print(type(control_patientdiff))

    treatment_patientdiff = difference_in_fields(treatment_d, 'panss')
    plt.hist(treatment_patientdiff, bins = 10)
    plt.xlabel('Newest Measurement - Oldest Measurement')
    plt.ylabel('Frequency')
    plt.title('Histogram of Differences in Treatment Group')
    plt.show()
    treatment_patientdiffstats = difference_in_scores_stats(treatment_patientdiff)

    #print('------------')


    print(f'Control Patients: {control_patientdiffstats}')
    print(len(control_d))
```

```
    print(f'Treatment Patients: {treatment_patientdiffstats}')
    print(len(treatment_d))

    print('------------')
    #print(patient_d)
    fname = "upload1.csv"
    upload_data_a = desired_data(patient_d)
    print(upload_data_a)
    save_file(upload data_a, fname)
    print('------------')
    zscore_train_a, zscore_test_a = z_score_convert2(train_data_a, test_data_a)




    #ypredproba_a = knn_pred(zscore_train_a, train_data_a['leadstatus'], zscore_test_a, test_data_a['leadstatus'])
    #ypredproba_a = knn_pred(train_data_a['xvalues'], train_data_a['leadstatus'], test_data_a['xvalues'], test_data_a['leadstatus'])

    #clf = KNeighborsClassifier(n_neighbors = 5)
    #clf = CategoricalNB(force_alpha=True)
    #clf = GaussianNB()
    #clf = tree.DecisionTreeClassifier(max_depth=8, min_samples_leaf=10, ccp_alpha=0.005, criterion='entropy')
    clf = RandomForestClassifier(max_depth=8, min_samples_leaf=10, ccp_alpha=0.005, random_state=0)
    ypredproba_a = class_pred(clf, train_data_a['xvalues'], train_data_a['leadstatus'], test_data_a['xvalues'], test_data_a['leadstatus'])


    print(ypredproba_a)
    print('------------')

    max_proba_v = ypredproba_a.max(axis=1)
    class_data_a = class_data_merge(test_data_a, max_proba_v)

    fname = "classupload1.csv"
    save_file(class_data_a, fname)
    raise SystemExit
    plot_patient_data(control_d)
    plot_patient_data(treatment_d)
    raise SystemExit

    # day_v = data_a['visitday']
    # # uday_v = unique(day_v)
    # week_v = uday_v/7
    # print(uday_v)
    # print('------------')
    # print('------------')
    # print(week_v)
    # print('------------')

    # print(f'data_a:10 {data_a[10:]}')
    m_v = data_a['txgroup'] ==   '"Treatment'
    day_v = data_a[m_v]['visitday']
    print('treatment day', sorted(day_v))
    uday_v = unique(day_v)
    # week_v = uday_v//7
    # print(f'uday_v treatments: {uday_v}')
    # print(f'week_v treatments: {week_v}')
    # #week_v = int32(week_v)
    # print('week count:', list(zip(arange(week_v.shape[0]), bincount(week_v))))
    patient_d = find_patients(data_a)
    print('Before delete:', len(patient_d))
    filter_patients(patient_d, day_limit=105)
    print('After delete:', len(patient_d))

    #print(len(a))
    print('------------')
    #print(d)
    print(patient_d)

    control_d, treatment_d = seperate_control_treatment(patient_d)
    #print("control", control_l)
    print('------------')
    print('------------')
    print('------------')
    print('------------')
    print('------------')
    #print("treatment", treatment_l)
    plot_patient_data(control_l)
    plot_patient_data(treatment_l)
if __name__ == '__main__':
    main()
```