treatment = pd.read_csv("/Users/dunyenghuh/Desktop/Grad School/Quarter1/TreatmentSample01. csv") **Exploratory Data Analysis (Understanding both tables) Treatment Sample CSV** In [4]: print(treatment.shape) #double check over missing data print(treatment.isnull().sum()) treatment.head() (714, 3)PatientID 0 TreatmentDate 0 DrugCode 0 dtype: int64 Out[4]: **PatientID** TreatmentDate DrugCode 0 2038 2010-01-24 1 2038 Α 2010-01-27 **2** 2038 Α 2010-01-30 **3** 2038 2010-02-02 Α **4** 2038 Α 2010-02-06 In [5]: # Patient IDs print("Unique Patient IDs in Treatment CSV: ", treatment.PatientID.nunique()) Unique Patient IDs in Treatment CSV: 27 In [6]: # Treatment Dates print("Unique Dates in Treatment CSV: ", treatment.TreatmentDate.nunique()) treatment.agg({'TreatmentDate':[min, max]}) Unique Dates in Treatment CSV: 469 Out[6]: **TreatmentDate** min | 2010-01-24 max | 2014-10-25 In [7]: # Drug Codes print("Unique DrugCodes in Treatment CSV: ", treatment.DrugCode.nunique()) print(treatment.DrugCode.value_counts()) Unique DrugCodes in Treatment CSV: 3 231 Α 202 Name: DrugCode, dtype: int64 **Diagnosis Sample CSV** In [8]: print(diagnosis.shape) #double check over missing data print(diagnosis.isnull().sum()) diagnosis.head() (44, 5)PatientID DiagnosisDate 0 DiagnosisCode 0 Diagnosis 0 IsCancerDiagnosis dtype: int64 Out[8]: PatientID DiagnosisDate DiagnosisCode Diagnosis IsCancerDiagnosis 0 2634 2011-02-19 285.8 Anemia **1** 5657 2012-06-07 285.8 Anemia False **2** 7937 2013-01-06 285.8 Anemia False **3** 8615 2013-07-18 284.9 False Anemia **4** 4354 2012-02-04 284.9 False Anemia In [9]: #Patient ID print("Unique Patient IDs in Diagnosis CSV: ", diagnosis.PatientID.nunique()) print("Note that the number of Patient IDs in two files differs") #some patients have multiple diagnosis diagnosis.groupby(['PatientID']).Diagnosis.count().reset_index().sort_values(by='Diagnosis', Unique Patient IDs in Diagnosis CSV: 32 Note that the number of Patient IDs in two files differs Out[9]: PatientID Diagnosis **23** | 6922 **14** | 4374 3 3449 **10** 3757 2 **21** | 6877 **22** | 6889 2 **13** | 4354 2 **12** | 4256 3095 **24** 7230 **19** | 6321 **25** 7242 **26** 7796 1 **27** 7937 **28** 7976 **29** 8615 **30** 8827 **20** 6837 **16** 5259 **18** | 6281 **17** | 5657 1 **1** 2120 **15** | 4692 **11** 3948 **7** 2770 **6** 2763 **5** 2634 2462 **3** 2425 1 2 2407 **31** 9331 In [10]: # Diagnosis Date print("Unique Dates in Diagnosis CSV: ", diagnosis.DiagnosisDate.nunique()) diagnosis.agg({'DiagnosisDate':[min, max]}) Unique Dates in Diagnosis CSV: 38 Out[10]: DiagnosisDate 2010-01-09 max | 2013-08-23 In [11]: #Diagnosis Code print("Unique DiagnosisCode in Diagnosis CSV: ", diagnosis.DiagnosisCode.nunique()) diagnosis.DiagnosisCode.value_counts() Unique DiagnosisCode in Diagnosis CSV: 20 Out[11]: 174.9 285.8 4 174.1 174.8 153.5 401.9 153.3 153.9 284.9 174.5 174.7 153.4 174.3 153.6 285.9 1 153.8 174.6 174.4 401.1 174.2 Name: DiagnosisCode, dtype: int64 In [12]: #Diagnosis print("Unique Diagnosis in Diagnosis CSV: ", diagnosis.Diagnosis.nunique()) diagnosis.Diagnosis.value_counts() Unique Diagnosis in Diagnosis CSV: 5 Out[12]: Breast Cancer Colon Cancer 11 7 Anemia Hypertension 3 Hypertension 1 Name: Diagnosis, dtype: int64 In [13]: #Is Cancer Diagnosis diagnosis.IsCancerDiagnosis.value_counts() Out[13]: True 11 False Name: IsCancerDiagnosis, dtype: int64 Merging Two CSV Files to make a comprehensive table In [14]: diag0 = diagnosis.copy() tret1 = treatment.copy() print("Diagnosis csv shape: ", diag0.shape) print("Treatment csv shape: ", tret1.shape) # Patients who are diagnosed of cancer are going to be the ones who are treated by medicine. # Separate Cancer Patients and Non-Cancer Patients diag_true = diag0[diag0.IsCancerDiagnosis == True] diag_false = diag0[diag0.IsCancerDiagnosis == False] # make column 'alldate' for the ease of merging diag_true['alldate'] = diag_true['DiagnosisDate'] tret1['alldate'] = tret1['TreatmentDate'] # sort values before merging diag_true = diag_true.sort_values(by=['PatientID', 'alldate']) tret1 = tret1.sort_values(by=['PatientID', 'alldate']) # outer join on 'PatientID' and 'alldate' merged = pd.merge(diag_true, tret1, how='outer', on=['PatientID', 'alldate']) merged = merged.sort_values(by=['PatientID', 'alldate']).reset_index(drop=True) # The two tables are joined on the 'DiagnosedDate', which is earlier than 'TreatmentDate' # Patients can get multiple treatments for each diagnosis. Thus, forward fill the below 4 co lumns for same diagnosis merged['DiagnosisCode'] = merged.groupby('PatientID')['DiagnosisCode'].ffill() merged['Diagnosis'] = merged.groupby('PatientID')['Diagnosis'].ffill() merged['IsCancerDiagnosis'] = merged.groupby('PatientID')['IsCancerDiagnosis'].ffill() merged['DiagnosisDate'] = merged.groupby('PatientID')['DiagnosisDate'].ffill() # The DiagnosedDate, in which two tables are joined on, has NaN values because treatment did not occur yet for this date # Rather than having a NaN value for this rows, backward fill the below two columns. merged['TreatmentDate'] = merged.groupby('PatientID')['TreatmentDate'].bfill() merged['DrugCode'] = merged.groupby('PatientID')['DrugCode'].bfill() # select the necessary columns and drop duplicate rows made from forwardfill and backwardfil df1 = merged[['PatientID', 'DiagnosisDate', 'DiagnosisCode', 'Diagnosis', 'IsCancerDiagnosi s', 'TreatmentDate','DrugCode']] df1 = df1.drop_duplicates() # to concat the diag_false dataset, make 2 new columns. # Since these patients are not diagnosed of cancers, they do not receive treatments and the columns below remain as NaN diag_false['TreatmentDate'] = np.NaN diag_false['DrugCode'] = np.NaN # Append the non-cancer dataset and sort the data again to finalize the comprehensive datase df1 = df1.append(diag_false) Diagnosis csv shape: (44, 5) Treatment csv shape: (714, 3) final data shape: (730, 7) /anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:12: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guid e/indexing.html#returning-a-view-versus-a-copy if sys.path[0] == '': /anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:41: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guid e/indexing.html#returning-a-view-versus-a-copy /anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:42: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guid e/indexing.html#returning-a-view-versus-a-copy Out[14]: DiagnosisDate DiagnosisCode Diagnosis | IsCancerDiagnosis | TreatmentDate | DrugCode PatientID **0** 2038 2010-01-24 2010-01-21 174.9 Breast Cancer | True **1** 2038 174.9 2010-01-21 Breast Cancer | True 2010-01-27 Α **2** 2038 2010-01-21 174.9 Breast Cancer | True 2010-01-30 **3** 2038 2010-01-21 174.9 Breast Cancer True 2010-02-02 Α Breast Cancer | True 2038 2010-01-21 174.9 2010-02-06 **Questions** Questions 1 1a) Which types of cancer does the clinic see patients for? 1b) How many patients does the clinic see for each cancer type? In [15]: # Select cancer patients from the diagnosis data and group them by Diagnosis type # Then count unique PatientIDs, as a Patient can have multiple diagnosis. cancer_count = diagnosis[diagnosis['IsCancerDiagnosis'] == True].groupby(['Diagnosis'], as_in dex=False).PatientID.nunique() cancer_count = cancer_count.rename(columns={"Diagnosis": "Daignosis Type", "PatientID": "Uniq ue Patient Count"}) print("There are 2 types of cancer: Breast and Colon.") print("There are 20 unique patients for Breast Cancer and 9 for Colon Cancer") cancer_count #alternatively, the above result can be found thru the code below - used for sanity check There are 2 types of cancer: Breast and Colon. There are 20 unique patients for Breast Cancer and 9 for Colon Cancer Out[15]: Daignosis Type | Unique Patient Count **0** Breast Cancer Colon Cancer Question 2: The clinic wants to know how long it takes for patients to start therapy after being diagnosed, which they consider to be helpful in understanding the quality of care for the patients. 2a) How long after being diagnosed do cancer patients start treatment for each cancer type? In [16]: # select patients with breast cancer and colon cancer from the comprehensive dataset # group by PatientID and DiagnosisDate and follow the earliest Treatment Date breast_cancer = df1[df1.Diagnosis == 'Breast Cancer'].groupby(['PatientID', 'DiagnosisDate'], as_index=False).agg({'TreatmentDate':min}) colon_cancer = df1[df1.Diagnosis == 'Colon Cancer'].groupby(['PatientID', 'DiagnosisDate'], as _index=**False**).agg({'TreatmentDate':min}) # change both dates in the 2 tables to datetime format to calculate the difference in days breast_cancer['DiagnosisDate'] = pd.to_datetime(breast_cancer['DiagnosisDate']) breast_cancer['TreatmentDate'] = pd.to_datetime(breast_cancer['TreatmentDate']) colon_cancer['DiagnosisDate'] = pd.to_datetime(colon_cancer['DiagnosisDate']) colon_cancer['TreatmentDate'] = pd.to_datetime(colon_cancer['TreatmentDate']) # calculate the difference in days breast_cancer['nday'] = (breast_cancer['TreatmentDate'] - breast_cancer['DiagnosisDate']).d colon_cancer['nday'] = (colon_cancer['TreatmentDate'] - colon_cancer['DiagnosisDate']).dt.d breast_cancer_nday = list(breast_cancer.nday) colon_cancer_nday = list(colon_cancer.nday) print("Mean time difference from diagnosis to treatment for breast cancer is :", st.mean(breast_cancer_nday), "days, for colon cancer:", st.mean(colon_cancer_nday), "day print("Median time difference from diagnosis to treatment for breast cancer is :", st.median(breast_cancer_nday), "days, for colon cancer:", st.median(colon_cancer_nday), Mean time difference from diagnosis to treatment for breast cancer is : 5.1 days, for colon c ancer: 13.1 days Median time difference from diagnosis to treatment for breast cancer is : 4.0 days, for colon cancer: 14.5 days 2b) Are there any patients which are diagnosed but not treated at the practice? In [18]: # select cancer patients whose drugcode is null print(df1[(df1.IsCancerDiagnosis == True)&(df1.DrugCode.isnull())]) # alternatively, check if all patients with cancer and are treated patientid_with_cancer = diagnosis[diagnosis.IsCancerDiagnosis == **True**].PatientID.unique().to list() patientid_treated = treatment.PatientID.unique().tolist() print([i for i in patientid_with_cancer if i not in patientid_treated]) Empty DataFrame Columns: [PatientID, DiagnosisDate, DiagnosisCode, Diagnosis, IsCancerDiagnosis, TreatmentDat e, DrugCode] Index: [] [] Question 3: After being treated with a first line of treatment (a drug or combination of drugs), what proportion of all cancer patients go on to be treated with a second line of treatment? (For more information on the concept of "first-line therapy", please reference https://www.cancer.net/navigating-cancer-care/how-cancer-treated/when-first-treatment-doesnt-work) In [19]: # Calculate the number of all cancer patients :27 df1[df1.IsCancerDiagnosis == True].PatientID.nunique() # Some Patients are treated through monotherapy, while others are treated through combinatio n of drugs # Because treatments with combination of drugs occur on the same treatment date, group the d ata by PatientId and Treatment Date # Then, join the DrugCodes so that patients with DrugCode (A,B) will appear as 'AB' change_in_treat = df1[df1.IsCancerDiagnosis == True].groupby(['PatientID', 'TreatmentDate'])['DrugCode'].apply(''.join).reset_index() # Sort the data to use the shift function change_in_treat.sort_values(by=['PatientID', 'TreatmentDate'], inplace=True) # There are cases where patients are diagnosed with more than 1 cancer and treated with 1 dr # Because I am going to be using the shift function, I want to make sure that above cases ar e captured normally # If any values show same DrugCode appearing more than once, just keep the last value change_in_treat["DrugCode"] = change_in_treat["DrugCode"].str.replace(r"(.)(?=.*\1)", "").st # to see the change in treatment we want to compare the treatment from current date and the treatment from the previous hospital visit # create a column 'lag' which shows the DrugCode from the previous treatment (thus need to g roupby patient ID) change_in_treat['Lag'] = change_in_treat.groupby('PatientID')['DrugCode'].shift() # fill the NaN values occuring at the first treatment with backfill method change_in_treat['Lag'] = change_in_treat.groupby('PatientID')['Lag'].bfill() # if the DrugCode column and Lag column differs, put 1, otherwise put 0 change_in_treat['difference'] = np.where(change_in_treat['DrugCode'] != change_in_treat['La g'], 1, 0) change_in_treat[(change_in_treat.difference != 0)].difference.sum() There are in total 27 cancer patients 7 Cancer patients experienced second line of treatment The proportion of cancer patients treated with second line of treatment is: 0.26 Question 4: How does each drug used at the clinic compare in terms of its duration of therapy? In [20]: # change the treatment date to datetime format df1['TreatmentDate'] = pd.to_datetime(df1['TreatmentDate']) # group the data by patientID and drugcode to calcualte the earliest and latest treatment da te for each (patient, drugcode) date_per_drug = df1[df1.IsCancerDiagnosis == True].groupby(['PatientID', 'DrugCode'], as_index =False).agg({'TreatmentDate':[min,max]}) # assume that duration of therapy can be represented by (latest treatment date - earliest tr eatment date) date_per_drug['nday'] = date_per_drug['TreatmentDate']['max'] - date_per_drug['TreatmentDat e']['min'] date_per_drug # make individual list for each drugcode A = list(date_per_drug[date_per_drug.DrugCode == 'A'].nday.dt.days) B = list(date_per_drug[date_per_drug.DrugCode == 'B'].nday.dt.days) C = list(date_per_drug[date_per_drug.DrugCode == 'C'].nday.dt.days) #plot the data fig, ax = plt.subplots(3, 2, figsize = (15, 10))plt.subplots_adjust(hspace = 0.5) sns.boxplot(A, color= "blue", ax=ax[0,0]) ax[0,0].set_title("Duration of Therapy for DrugCode A") ax[0,0].set_xlabel("Number of days") sns.boxplot(B, color= "orange", ax=ax[1,0]) ax[1,0].set_title("Duration of Therapy for DrugCode B") ax[1,0].set_xlabel("Number of days") $B_{new} = B.copy()$ B_new.remove(114) sns.boxplot(B_new, color= "orange", ax=ax[1,1]) ax[1,1].set_title("Duration of Therapy for DrugCode B without outlier") ax[1,1].set_xlabel("Number of days") sns.boxplot(C, color= "green", ax=ax[2,0]) ax[2,0].set_title("Duration of Therapy for DrugCode C") ax[2,0].set_xlabel("Number of days") $C_new = C.copy()$ C_new.remove(520) sns.boxplot(C_new, color= "green", ax=ax[2,1]) ax[2,1].set_title("Duration of Therapy for DrugCode C without outlier") ax[2,1].set_xlabel("Number of days") print("The distribution of therapy duration for drug A is a bit skewed to the left. It's ske wness is", skew(A)) print("\nOne datapoint stands out from the distribution of therapy duration for drug B: Pati entID 7242 has been treated for more than 114 days.") print("We can think of the above datapoints as an outlier and plot the distribution disregar ding the datapoint.") print("The new distribution of therapy duration for drug B is approximately symmetric and ju st slightly skewed to the right. It's skewness is", skew(B_new)) print("\nOne datapoint stands out from the distribution of therapy duration for drug C: Pati The distribution of therapy duration for drug A is a bit skewed to the left. It's skewness is -0.5669376185649837 One datapoint stands out from the distribution of therapy duration for drug B: PatientID 7242 has been treated for more than 114 days. We can think of the above datapoints as an outlier and plot the distribution disregarding the The new distribution of therapy duration for drug B is approximately symmetric and just sligh tly skewed to the right. It's skewness is 0.18742239669160704 One datapoint stands out from the distribution of therapy duration for drug C: PatientID 7242 has been treated for more than 500 days. We can think of the above datapoints as an outlier and plot the distribution disregarding the Duration of Therapy for DrugCode A 1.0 0.8 0.6 0.4 0.2 0.0 0.2 Number of days Duration of Therapy for DrugCode B Duration of Therapy for DrugCode B without outlier Number of days Number of days Duration of Therapy for DrugCode C without outlier Duration of Therapy for DrugCode C 200 300 85 105 Number of days Number of days In [21]: # make 2 new dataframes to compare data with outliers and without outliers a= pd.DataFrame(A, columns = ['days']) a['DrugCode'] = 'A' b= pd.DataFrame(B, columns = ['days']) b['DrugCode'] = 'B' c = pd.DataFrame(C, columns = ['days']) c['DrugCode'] = 'C' distribution= a.append(b) distribution= distribution.append(c) # data without outliers b_new= pd.DataFrame(B_new, columns = ['days']) b_new['DrugCode'] = 'B' c_new = pd.DataFrame(C_new, columns = ['days']) c_new['DrugCode'] = 'C' distribution_new= a.append(b_new) distribution_new= distribution_new.append(c_new) #plotting fig, (ax1, ax2) = plt.subplots(nrows=1, ncols=2, figsize=(15, 9), sharey=True) sns.boxplot(x="days", y="DrugCode", data=distribution, ax= ax1) sns.boxplot(x="days", y="DrugCode", data=distribution_new, ax = ax2) ax1.set_title('Duration of Therapy') ax2.set_title('Duration of Therapy (without outliers)') # print the summary statistics print("Now let's take a look at the summary statistics while ignoring the outliers (outlier point comes from the same patient)") print("\nMean therapy duration for drug A:", st.mean(A),", B:",st.mean(B_new),", C:",st.mean print("Median therapy duration for drug A:", st.median(A),", B:",st.median(B_new),", C:",st. median(C_new)) print("The Spread of therapy duration for drug A:", max(A)-min(A),", B:", max(B_new)-min(B_ne w), ", C:", max(C_new)-min(C_new)) print("Maximum therapy duration for drug A:", max(A),", B:",max(B_new),", C:",max(C_new)) print("Minimum therapy duration for drug A:", min(A),", B:",min(B_new),", C:",min(C_new)) print("\nDrugCode A has the lowest mean therapy duration, followed by B and C") print("DrugCode B has the lowest median therapy duration, followed by A and C") Now let's take a look at the summary statistics while ignoring the outliers (outlier point co mes from the same patient) Mean therapy duration for drug A: 56.13333333333333333333333333333 , B: 59.25 , C: 84.866666666666666 Median therapy duration for drug A: 60 , B: 59.0 , C: 84 The Spread of therapy duration for drug A: 50 , B: 49 , C: 40 Maximum therapy duration for drug A: 77 , B: 85 , C: 106 Minimum therapy duration for drug A: 27 , B: 36 , C: 66 DrugCode A has the lowest mean therapy duration, followed by B and C DrugCode B has the lowest median therapy duration, followed by A and C The Mean, Median and the spread of therapy duration for A and B are very similar when compare Duration of Therapy Duration of Therapy (without outliers)

100

Done

200

300

400

500

90

100

Dunyeng Huh Flatiron Interview

treatment = pd.read_csv("data/TreatmentSample01.csv")
diagnosis = pd.read_csv("data/DiagnosisSample01.csv")

In [3]: import seaborn as sns

import pandas as pd
import datetime as dt
import numpy as np

import statistics as st

import matplotlib.pyplot as plt

from scipy.stats import skew, kurtosis

from tabulate import tabulate

Change this line to get data