

PPMI Merge

2022-05-04

Here, we utilize PPMI subject clinical data files (PPMI Study Data) to create a single .csv file that contains longitudinal clinical information of PPMI subjects (included in Analytic Cohort).

Getting the data

- 1.) Get access to PPMI database and login.
- 2.) Click download from the navigation bar and select study data.
- 3.) Select ALL documents and zip files and click download.
- 4.) Unzip and save the folder on your computer.

Import required modules

```
import pandas as pd
import numpy as np
import boto3
from datetime import datetime
from dateutil import relativedelta

userdir = '/Users/areardon/Desktop/ppmi_merge/'
ppmi_download_path = userdir + 'PPMI_Study_Data_Download/'
invicro_data_path = userdir
genetics_path = userdir + 'genetic_data/'
version = '0.0.3' # File version to save as i.e. ppmi_merge_v0.1.csv

def create_cohort_df(xlsx, sheet) :
    cohort_df = pd.read_excel(xlsx, sheet)
    cols_delete = ['Unnamed', 'CONDATE'] # Columns to remove from sheet/df
    cohort_df = cohort_df.loc[:, ~cohort_df.columns.str.startswith(tuple(cols_delete))] # Remove column

    # Create new columns for Enrollment.Subtype and Consensus.Subtype in pd_df
    cohort_df["Enrollment.Subtype"] = '' # Create new column called 'Enrollment.Subtype' in cohort_df
    cohort_df["Consensus.Subtype"] = '' # Create new column called 'Consensus.Subtype' in cohort_df

    if sheet == 'HC' :
        cohort_df['Subgroup'] = 'Healthy Control'
    if sheet == 'SWEDD' :
        cohort_df['Comments'] = '' # Replace comments with empty string because we don't want to merge

    return cohort_df

def merge_columns(df : pd.DataFrame , old_df_columns : list, new_df_column_name : str, separator = str)
    """
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Takes entries in each of old_df_columns and joins them together with a separator of choice. Removes empty/nan column entries.
"""
df = df.replace(r'^\s*$', np.NaN, regex=True) # Fill in empty cells with nan
df[new_df_column_name] = df[old_df_columns].agg(lambda x: x.dropna().str.cat(sep= separator), axis=
df.drop(old_df_columns, axis = 1, inplace = True)
return df

def getNamesFromDataframe(df, str) :
    col_names = [col for col in df.columns if str in col]
    return col_names

def merge_new_csv(df, csv_filename, list_cols, merge_on, merge_how) :
    demo_df = pd.read_csv(ppmi_download_path + csv_filename, skipinitialspace = True)
    demo_df = demo_df[list_cols]
    ppmi_merge = pd.merge(df, demo_df, on = merge_on, how = merge_how) # Merge
    return ppmi_merge

def isNaN(num):
    return num != num

def decode_0_1_no_yes(df, list) :
    for i in list :
        df[i] = df[i].astype(int, errors = "ignore")
        df[i].replace({0 : 'No' , 1 : 'Yes', 2 : 'Uncertain'}, inplace = True) # FIXME replace Uncertain
    return df

def decode_none2severe(df, list):
    for i in list :
        df[i] = df[i].astype(int, errors = "ignore")
        df[i].replace({0 : 'None', 1 : 'Slight', 2: 'Mild', 3 : 'Moderate', 4 : 'Severe'}, inplace = True)
    return df

def condensed_df(df, keep_col_list, rename_col_dict, drop_col_list) :
    new_df = df[keep_col_list]
    new_df.rename(columns = rename_col_dict, inplace = True)
    new_df.dropna(subset = drop_col_list, inplace = True)
    return new_df

#### CLINICAL INFO ####
# Create cohort df
xlsx = pd.ExcelFile(ppmi_download_path + 'Consensus_Committee_Analytic_Datasets_28OCT21.xlsx') # Read i
pd_df = create_cohort_df(xlsx, 'PD') # Create Parkinson's Disease data frame from 'PD' sheet in 'Consen
prodromal_df = create_cohort_df(xlsx, 'Prodromal') # Create Prodromal data frame from 'PD' sheet in 'Con
hc_df = create_cohort_df(xlsx, 'HC') # Create HC data frame from 'PD' sheet in 'Consensus_Committee_Anal
swedd_df = create_cohort_df(xlsx, 'SWEDD') # Create SWEDD data frame from 'PD' sheet in 'Consensus_Comm

# Decode Enrollment.Subtype ('Summary' sheet) and Consensus.Subtype ('Summary Analytic' sheet) of 'Cons
pd_df.loc[(pd_df['ENRLPD'] == 1) & (pd_df['ENRLRRK2'] == 0) & (pd_df['ENRLGBA'] == 0) & (pd_df['ENRLSN
pd_df.loc[(pd_df['ENRLPD'] == 1) & (pd_df['ENRLRRK2'] == 1), 'Enrollment.Subtype'] = ' : LRRK2'
pd_df.loc[(pd_df['ENRLPD'] == 1) & (pd_df['ENRLGBA'] == 1), 'Enrollment.Subtype'] = ' : GBA'
pd_df.loc[(pd_df['ENRLPD'] == 1) & (pd_df['ENRLSNCA'] == 1), 'Enrollment.Subtype'] = ' : SNCA'
pd_df.loc[(pd_df['CONPD'] == 1) & (pd_df['CONLRRK2'] == 0) | (pd_df['CONLRRK2'] == '.') & (pd_df['CONGB

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pd_df.loc[(pd_df['CONPD'] == 1) & (pd_df['CONLRRK2'] == 1) & (pd_df['CONGBA'] == 0) & (pd_df['CONSNCA'] == 0)] = 'non-PD' # FIXME
pd_df.loc[(pd_df['CONPD'] == 1) & (pd_df['CONLRRK2'] == 0) & (pd_df['CONGBA'] == 1) & (pd_df['CONSNCA'] == 0)] = 'non-PD' # FIXME
pd_df.loc[(pd_df['CONPD'] == 1) & (pd_df['CONLRRK2'] == 1) & (pd_df['CONGBA'] == 1) & (pd_df['CONSNCA'] == 0)] = 'non-PD' # FIXME
pd_df.loc[(pd_df['CONPD'] == 1) & (pd_df['CONLRRK2'] == 0) & (pd_df['CONGBA'] == 0) & (pd_df['CONSNCA'] == 1)] = 'non-PD' # FIXME
pd_df.loc[(pd_df['CONPD'] == 0) & (pd_df['CONPROD'] == 0) & (pd_df['CONLRRK2'] == 1) & (pd_df['CONGBA'] == 1) & (pd_df['CONSNCA'] == 1)] = 'non-PD' # FIXME
pd_df.loc[(pd_df['CONPD'] == 0) & (pd_df['CONPROD'] == 1) & (pd_df['CONLRRK2'] == 1) & (pd_df['CONGBA'] == 1) & (pd_df['CONSNCA'] == 1)] = 'non-PD' # FIXME
pd_df.loc[(pd_df['CONPD'] == 0) & (pd_df['CONPROD'] == 1) & (pd_df['CONLRRK2'] == 0) & (pd_df['CONGBA'] == 1) & (pd_df['CONSNCA'] == 1)] = 'non-PD' # FIXME

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# Decode Enrollment.Subtype and Consensus.Subtype to be categorical variables for prodromal df
prodromal_df.loc[(prodromal_df['ENRLPROD'] == 1) & (prodromal_df['ENRLRRK2'] == 1), 'Enrollment.Subtype'] = 'ENRLPROD'
prodromal_df.loc[(prodromal_df['ENRLPROD'] == 1) & (prodromal_df['ENRLGBA'] == 1), 'Enrollment.Subtype'] = 'ENRLGBA'
prodromal_df.loc[(prodromal_df['ENRLPROD'] == 1) & (prodromal_df['ENRLSNCA'] == 1), 'Enrollment.Subtype'] = 'ENRLSNCA'
prodromal_df.loc[(prodromal_df['ENRLPROD'] == 1) & (prodromal_df['ENRLHPSM'] == 1), 'Enrollment.Subtype'] = 'ENRLHPSM'
prodromal_df.loc[(prodromal_df['ENRLPROD'] == 1) & (prodromal_df['ENRLRBD'] == 1), 'Enrollment.Subtype'] = 'ENRLRBD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['PHENOCNV'] == 0) & (prodromal_df['CONLRRK2'] == 1) & (prodromal_df['CONGBA'] == 0) & (prodromal_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['PHENOCNV'] == 1) & (prodromal_df['CONLRRK2'] == 1) & (prodromal_df['CONGBA'] == 0) & (prodromal_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 0) & (prodromal_df['CONLRRK2'] == 1) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['PHENOCNV'] == 0) & (prodromal_df['CONLRRK2'] == 1) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['PHENOCNV'] == 1) & (prodromal_df['CONLRRK2'] == 1) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 0) & (prodromal_df['CONLRRK2'] == 0) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['PHENOCNV'] == 0) & (prodromal_df['CONLRRK2'] == 0) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['PHENOCNV'] == 1) & (prodromal_df['CONLRRK2'] == 0) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 0) & (prodromal_df['CONLRRK2'] == 0) & (prodromal_df['CONGBA'] == 0) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['PHENOCNV'] == 0) & (prodromal_df['CONLRRK2'] == 0) & (prodromal_df['CONGBA'] == 0) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['PHENOCNV'] == 1) & (prodromal_df['CONLRRK2'] == 0) & (prodromal_df['CONGBA'] == 0) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 0) & (prodromal_df['CONLRRK2'] == 0) & (prodromal_df['CONGBA'] == 0) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['CONRBD'] == 1) & (prodromal_df['PHENOCNV'] == 0) & (prodromal_df['CONLRRK2'] == 1) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['CONRBD'] == 1) & (prodromal_df['PHENOCNV'] == 1) & (prodromal_df['CONLRRK2'] == 1) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'
prodromal_df.loc[(prodromal_df['CONPROD'] == 1) & (prodromal_df['CONRBD'] == 1) & (prodromal_df['PHENOCNV'] == 0) & (prodromal_df['CONLRRK2'] == 0) & (prodromal_df['CONGBA'] == 1) & (prodromal_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONPROD'

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# Decode Enrollment.Subtype and Consensus.Subtype to be categorical variables for Healthy Control df
hc_df.loc[(hc_df['ENRLHC'] == 1), 'Enrollment.Subtype'] = 'ENRLHC' # Healthy Control already covered in Subgr
hc_df.loc[(hc_df['CONHC'] == 1) & (hc_df['CONLRRK2'] == 0) & (hc_df['CONGBA'] == 0) & (hc_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONHC'
hc_df.loc[(hc_df['CONHC'] == 1) & (hc_df['CONLRRK2'] == 1) & (hc_df['CONGBA'] == 0) & (hc_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONHC'
hc_df.loc[(hc_df['CONHC'] == 1) & (hc_df['CONLRRK2'] == 0) & (hc_df['CONGBA'] == 1) & (hc_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONHC'
hc_df.loc[(hc_df['CONHC'] == 1) & (hc_df['CONLRRK2'] == 1) & (hc_df['CONGBA'] == 1) & (hc_df['CONSNCA'] == 0)], 'Consensus.Subtype'] = 'CONHC'
hc_df.loc[(hc_df['CONHC'] == 1) & (hc_df['CONLRRK2'] == 0) & (hc_df['CONGBA'] == 0) & (hc_df['CONSNCA'] == 1)], 'Consensus.Subtype'] = 'CONHC'
hc_df.loc[(hc_df['CONHC'] == 0), 'Consensus.Subtype'] = 'non-HC' # FIXME

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# Decode Enrollment.Subtype and Consensus.Subtype to be categorical variables for swedd df
swedd_df.loc[swedd_df['ENRLSWEDD'] == 1, 'Enrollment.Subtype'] = 'SWEDD Legacy'
swedd_df.loc[swedd_df['CONSWEDD'] == 0, 'Consensus.Subtype'] = 'SWEDD/PD Active'
swedd_df.loc[swedd_df['CONSWEDD'] == 1, 'Consensus.Subtype'] = 'SWEDD/non-PD Active'

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# Merge the four cohort dataframes (HC, Prodromal, PD, SWEDD)
full_df = pd_df.append([prodromal_df, hc_df, swedd_df]) # Concat all 4 cohort dfs

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# Decode and re-organize full_df
full_df['CONPD'].replace({1 : 'Parkinson\'s Disease', 0 : ''}, inplace = True)

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full_df['CONPROD'].replace({1 : 'Prodromal', 0 : ''}, inplace = True)
full_df['CONHC'].replace({1 : 'Healthy Control', 0 : ''}, inplace = True)
full_df['CONSWEDD'].replace({1 : 'SWEDD', 0 : 'SWEDD/PD'}, inplace = True)
full_df['Comments'].replace({'MSA' : 'Multiple System Atrophy'}, inplace = True)
full_df = decode_0_1_no_yes(full_df, ['PHENOCNV'])
full_df = merge_columns(full_df, ['CONPD', 'CONPROD', 'CONHC', 'CONSWEDD', 'Comments'], 'Consensus.Diagnosis')
full_df = merge_columns(full_df, ['Subgroup', 'Enrollment.Subtype'], 'Enroll.Subtype', '') # Get one column
full_df = full_df.loc[:, ~full_df.columns.str.startswith(tuple(['CON', 'ENRL']))] # Remove columns that start with CON or ENRL
full_df.rename(columns = {'Cohort' : 'Enroll.Diagnosis', 'PHENOCNV' : 'Subject.Phenoconverted'}, inplace = True)
full_df = full_df[['PATNO', 'Enroll.Diagnosis', 'Enroll.Subtype', 'Consensus.Diagnosis', 'Consensus.Subtype']]
analytic_cohort_subids = full_df['PATNO'].unique() # subids for analytic cohort

## Add in age info at each visit
ppmi_merge = merge_new_csv(full_df, 'Age_at_visit.csv', ['PATNO', 'EVENT_ID', 'AGE_AT_VISIT'], merge_on = ['PATNO', 'EVENT_ID'])
ppmi_merge['EVENT_ID'].fillna('SC', inplace = True) # One subject (41358) has event_id as NaN - replace with SC

## Add demographics, vital signs, and pd diagnosis history info
ppmi_merge = merge_new_csv(ppmi_merge, 'Demographics.csv', ['PATNO', 'EVENT_ID', 'SEX', 'HANDED', 'BIRTHDT'])
ppmi_merge = merge_new_csv(ppmi_merge, 'Vital_Signs.csv', ['PATNO', 'EVENT_ID', 'INFODT', 'WGTKG', 'HTCM'])
ppmi_merge = merge_new_csv(ppmi_merge, 'PD_Diagnosis_History.csv', ['PATNO', 'EVENT_ID', 'SXDT', 'PDDXD'])
ppmi_merge['SEX'].replace({0 : 'Female', 1 : 'Male'}, inplace = True) # Decode sex
ppmi_merge['HANDED'].replace({1 : 'Right', 2 : 'Left', 3 : 'Mixed'}, inplace = True) # Decode handedness
ppmi_merge.rename(columns = {'AGE_AT_VISIT' : 'Age', 'SEX' : 'Sex', 'HANDED' : 'Handed', 'BIRTHDT' : 'BirthDate'})

## Add a PD.Disease.Duration variable (in years)
ppmi_merge['PD.Disease.Duration'] = '' # Initialize PD.Disease.Duration variable
for row_num in range(len(ppmi_merge['PD.Disease.Date'])):
    if isinstance(ppmi_merge['PD.Disease.Date'].loc[row_num], str) and isinstance(ppmi_merge['INFODT'].loc[row_num], str):
        diag_year = int(ppmi_merge['PD.Disease.Date'].loc[row_num].split('/')[1]) # Diagnosis year
        diag_month = int(ppmi_merge['PD.Disease.Date'].loc[row_num].split('/')[0]) # Diagnosis month
        event_year = int(ppmi_merge['INFODT'].loc[row_num].split('/')[1]) # Visit date year
        event_month = int(ppmi_merge['INFODT'].loc[row_num].split('/')[0]) # Visit date month
        diff = relativedelta.datetime(event_year, event_month, 1), datetime(diag_year, diag_month, 1)
        ppmi_merge['PD.Disease.Duration'].iloc[row_num] = ((diff.years)*12 + diff.months)/12 # PD.Disease.Duration in years

# Add Final (FNL) Event ID and info
ppmi_merge = merge_new_csv(ppmi_merge, 'Conclusion_of_Study_Participation.csv', ['PATNO', 'EVENT_ID', 'COMPLT'])
ppmi_merge = decode_0_1_no_yes(ppmi_merge, ['COMPLT'])
ppmi_merge['WDRSN'].replace({1 : 'Adverse Event', 2 : 'Completed study per protocol', 3 : 'Death', 4 : 'Withdrawal'}, inplace = True)
ppmi_merge.rename(columns = {'COMPLT' : 'Completed.Study', 'WDRSN' : 'Reason.for.Withdrawal', 'WDDT' : 'Withdrawal.Date'})

## Add diagnosis change info on event id
diag_vis1 = condensed_df(full_df, ['PATNO', 'DIAG1', 'DIAG1VIS'], {'DIAG1VIS' : 'EVENT_ID'}, ['EVENT_ID'])
diag_vis2 = condensed_df(full_df, ['PATNO', 'DIAG2', 'DIAG2VIS'], {'DIAG2VIS' : 'EVENT_ID'}, ['EVENT_ID'])
ppmi_merge.drop(['DIAG1', 'DIAG1VIS', 'DIAG2', 'DIAG2VIS'], axis = 1, inplace = True) # Drop these from ppmi_merge
ppmi_merge = pd.merge(diag_vis1, ppmi_merge, on = ['EVENT_ID', 'PATNO'], how = "outer") # Merge in first diagnosis
ppmi_merge = pd.merge(diag_vis2, ppmi_merge, on = ['EVENT_ID', 'PATNO'], how = "outer") # Merge in second diagnosis
ppmi_merge['DIAG1'].replace({'PD' : 'Parkinson's Disease', 'DLB' : 'Dementia with Lewy Bodies'}, inplace = True)
ppmi_merge['DIAG2'].replace({'MSA' : 'Multiple System Atrophy', 'DLB' : 'Dementia with Lewy Bodies'}, inplace = True)
ppmi_merge.rename(columns = {'DIAG1' : 'First.Diagnosis.Change', 'DIAG2' : 'Second.Diagnosis.Change'}, inplace = True)

## Dominant side of disease
ppmi_merge = merge_new_csv(ppmi_merge, 'PPMI_Original_Cohort_BL_to_Year_5_Dataset_Apr2020.csv', ['PATNO', 'EVENT_ID', 'DOMINANT_SIDE'])

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ppmi_merge['DOMSIDE'].replace({1 : 'Left', 2 : 'Right', 3 : 'Symmetric'}, inplace = True) # Decode

## Participant Motor Function Questionnaire
ppmi_merge = merge_new_csv(ppmi_merge, 'Participant_Motor_Function_Questionnaire.csv', ['PATNO', 'EVENT_ID', 'PAG_NAME', 'CMPLBY2'])
ppmi_merge['CMPLBY2'].replace({1: 'Participant', 2 : 'Caregiver', 3 : 'Participant and Caregiver'}, inplace = True)
ppmi_merge['PAG_NAME'].replace({'PQUEST' : 'Participant Motor Function Questionnaire'}, inplace = True)
ppmi_merge = decode_0_1_no_yes(ppmi_merge, ['TRBUPCHR', 'WRTSMLR', 'VOICSFTR', 'POORBAL', 'FTSTUCK', 'LSTUCK'])
ppmi_merge = ppmi_merge.rename(columns = {'PAG_NAME' : 'Motor.Function.Page.Name', 'CMPLBY2' : 'Motor.Function.Completeness'})

## Cognitive symptoms - Cognitive Categorization
ppmi_merge = merge_new_csv(ppmi_merge, 'Cognitive_Categorization.csv', ['PATNO', 'EVENT_ID', 'PAG_NAME', 'COGDECLN', 'FNCDCOG'])
ppmi_merge = decode_0_1_no_yes(ppmi_merge, ['COGDECLN', 'FNCDCOG'])
ppmi_merge['COGDXCL'].replace({1 : '90 - 100%', 2 : '50 - 89%', 3 : '10 - 49%', 4 : '0 - 9%'}, inplace = True)
ppmi_merge['PTCGBOTH'].replace({1 : 'Participant', 2 : 'Caregiver', 3 : 'Participant and Caregiver'}, inplace = True)
ppmi_merge['COGSTATE'].replace({1 : 'Normal Condition', 2 : 'Mild Cognitive Impairment', 3 : 'Dementia'}, inplace = True)
ppmi_merge = ppmi_merge.rename(columns = {'PAG_NAME' : 'Cognitive.Page.Name', 'COGDECLN' : 'Cognitive.Dementia'})
ppmi_merge['Cognitive.Page.Name'].replace({'COGCATG' : 'Cognitive Categorization'}, inplace = True) # Rename

## Cognitive symptoms - MOCA
ppmi_merge = merge_new_csv(ppmi_merge, 'Montreal_Cognitive_Assessment_MoCA.csv', ['PATNO', 'EVENT_ID', 'MCATOT'])
ppmi_merge.rename(columns = {'MCATOT' : 'MOCA.Total'}, inplace = True) # Rename

## Medication status - OLD WAY START
## Medication Status - Concomitant Med Log # FIXME not a useful way to show medication status
med_df = pd.read_csv(ppmi_download_path + 'Concomitant_Medication_Log.csv', skipinitialspace=True) # Merge
med_df.replace({';' : ','}, regex = True, inplace = True) # Replace ';' with ',' in med_df
# FIXME - do we want to get rid of certain supplements we aren't interested in? Or leave all?
med_df['CMTRT'] = med_df['CMTRT'].str.title() # Capitalize all medication names
med_df['CMTRT'] = med_df['CMTRT'].str.strip()
n = 20 # FIXME how many of these to keep
selection = med_df['CMTRT'].value_counts()[:n].index.tolist()
med_df = med_df.loc[med_df['CMTRT'].isin(selection)]
med_df['CMTRT'] = med_df['CMTRT'].str.title() # Capitalize all medication names
med_df['STARTDT'].fillna('NA', inplace = True) # Fillna
med_df['STOPDT'].fillna('NA', inplace = True) # Fillna
med_df = med_df.astype({'STARTDT' : 'str', 'STOPDT' : 'str'}) # Change start and stop date to strings
med_df = merge_columns(med_df, ['STARTDT', 'STOPDT', 'Start_Stop', '-']) # Merge columns Start and stop
med_df['Start_Stop'] = '(' + med_df['Start_Stop'].astype(str) + ') # Put parenthesis around dates so wh
med_df = merge_columns(med_df, ['CMTRT', 'Start_Stop', 'Medication', ''])
med_df = med_df.groupby(['PATNO', 'EVENT_ID'])['Medication'].apply('; '.join)
ppmi_merge = pd.merge(ppmi_merge, med_df, on = ['PATNO', 'EVENT_ID'], how = "outer") # Merge med_df in
ppmi_merge = ppmi_merge.sort_values(by = ['PATNO', 'Age']).reset_index(drop = True) # Sort values by sub

## Medication status - OLD WAY END

# ## Medication status- NEW WAY START
# ## Medication Status - Concomitant Med Log # FIXME not a useful way to show medication status
# med_df = pd.read_csv(ppmi_download_path + 'Concomitant_Medication_Log.csv', skipinitialspace=True) # Merge
# med_df.replace({';' : ','}, regex = True, inplace = True) # Replace ';' with ',' in med_df

# # FIXME - do we want to get rid of certain supplements we aren't interested in? Or leave all?

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# med_df = pd.read_csv('/Users/areardon/Desktop/ppmi_merge/PPMI_Study_Data_Download/Concomitant_Medicat
# med_df['CMTRT'] = med_df['CMTRT'].str.title() # Capitalize all medication names
# med_df['CMTRT'] = med_df['CMTRT'].str.strip()
# med_df['CMTRT'] = med_df.CMTRT.str.replace(r'(. *Vitamin D.*$)', 'Vitamin D')
# n = 20 # FIXME how many of these to keep
# selection = med_df['CMTRT'].value_counts()[:n].index.tolist() # Get n most frequently reported medica
# med_df = med_df.loc[med_df['CMTRT'].isin(selection)] # Keep only entries that are in selection list
# med_df['CMTRT'] = med_df['CMTRT'].str.title() # Capitalize all medication names
# med_df = med_df[['PATNO', 'CMTRT', 'STARTDT', 'STOPDT', 'ONGOING']] # Keep only # FIXME do we want to
# med_df['STARTDT'] = pd.to_datetime(med_df['STARTDT'], format = '%m/%Y', errors = 'ignore') # change t
# med_df['STOPDT'] = pd.to_datetime(med_df['STOPDT'], format = '%m/%Y') # change to STOPDT to type date
# med_df['STOPDT2'] = med_df['STOPDT'] # Initialize second stop date variable

# # Fill in Stopdt2 before exploding # FIXME made assumptions here need to check
# med_df['ONGOING'].fillna('NA', inplace = True) # Fill in NaN stop dates with start dates # ASSUMPTION
# med_df['STOPDT2'].fillna('NA', inplace = True) # Fill in NaN stop dates with start dates # ASSUMPTIO
# med_df.loc[(med_df['STOPDT2'].isna()) & (med_df['ONGOING'] == 0), 'STOPDT2'] = med_df['STARTDT']
# med_df.loc[(med_df['STOPDT2'].isna()) & (med_df['ONGOING'] == 1.0), 'STOPDT2'] = datetime.strptime('0
# med_df.loc[(med_df['STOPDT2'].isna()) & (med_df['ONGOING'] == 'NA'), 'STOPDT2'] = med_df['STARTDT'] #
# med_df['STOPDT2'] = pd.to_datetime(med_df['STOPDT2'], format = '%m/%Y') # change to STOPDT to type da

# # If Startdt < screening or Baseline visit date (first date), Startdt = first visit date
# # IF Stopdt > last visit date, Stopdt = last visit date
# med_df['STARTDT'] = pd.to_datetime(med_df['STARTDT'], format = '%m/%Y') # change to STOPDT to type da
# med_df['STOPDT2'] = pd.to_datetime(med_df['STOPDT2'], format = '%m/%Y') # change to STOPDT to type da
# med_df.loc[(med_df['STARTDT'] != np.nan) & (med_df['STARTDT'] < datetime.strptime('01/2011', '%m/%Y')
# med_df['STARTDT'].fillna(datetime.strptime('06/2022', '%m/%Y'), inplace = True)
# med_df['STOPDT2'].fillna(datetime.strptime('06/2022', '%m/%Y'), inplace = True)
# med_df = med_df.merge(med_df.apply(lambda s: pd.date_range(s.STARTDT, s.STOPDT2, freq='MS', inclusive
# med_df = med_df.reset_index(drop = True)
# med_df['STARTDT'] = med_df['STARTDT'].dt.strftime('%m/%Y')
# med_df.loc[med_df['INFODT'].isna(), 'INFODT'] = med_df['STARTDT']
# med_df.drop(['STOPDT2', 'ONGOING'], axis = 1, inplace = True) # Drop
# med_df.rename(columns = {'STARTDT' : 'Med.STARTDT', 'STOPDT': 'Med.STOPDT'}, inplace = True) # Rename
# ppmi_merge_temp = ppmi_merge[['PATNO', 'EVENT_ID', 'INFODT']]
# ppmi_merge_temp['INFODT'] = ppmi_merge_temp['INFODT'].astype(str)
# med_df['INFODT'] = med_df['INFODT'].astype(str)
# ppmi_merge_temp = pd.merge(ppmi_merge_temp, med_df, on = ['PATNO', 'INFODT'], how = "left")
# ppmi_merge_temp = ppmi_merge_temp[['PATNO', 'EVENT_ID', 'CMTRT', 'Med.STARTDT', 'Med.STOPDT']] # keep
# ppmi_merge = pd.merge(ppmi_merge, ppmi_merge_temp, on = ['PATNO', 'EVENT_ID'], how = 'left')
# ## Medication status - NEW WAY END

## LEDD Medication Status - FIXME Assumption : If stop date is NA we assume LEDD only occurred in that m
LEDD_med_df = pd.read_csv(ppmi_download_path + 'LEDD_Concomitant_Medication_Log.csv', skipinitialspace =
LEDD_med_df = LEDD_med_df[['PATNO', 'LEDD', 'STARTDT', 'STOPDT']] # Keep only
ppmi_merge['INFODT'] = pd.to_datetime(ppmi_merge['INFODT'], format = '%m/%Y', errors = 'ignore') # chang
LEDD_med_df['STARTDT'] = pd.to_datetime(LEDD_med_df['STARTDT'], format = '%m/%Y', errors = 'ignore') # c

LEDD_med_df['STOPDT'] = pd.to_datetime(LEDD_med_df['STOPDT'], format = '%m/%Y', errors = 'ignore') # cha
LEDD_med_df['STOPDT2'] = LEDD_med_df['STOPDT'] # Initialize second stop date variable
LEDD_med_df['STOPDT2'].fillna(LEDD_med_df['STARTDT'], inplace = True) # Fill in NaN stop dates with sta
LEDD_med_df = LEDD_med_df.merge(LEDD_med_df.apply(lambda s: pd.date_range(s.STARTDT, s.STOPDT2, freq='MS

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LEDD_med_df.drop(['STOPDT2'], axis = 1, inplace = True) # Drop
LEDD_med_df.rename(columns = {'STARTDT' : 'LEDD.STARTDT', 'STOPDT': 'LEDD.STOPDT'}, inplace = True) # R

# Get a new df with LEDD dates merged onto corect event ids through infodts
ppmi_merge_temp = ppmi_merge[['PATNO', 'EVENT_ID', 'INFODT']]
ppmi_merge_temp = pd.merge(ppmi_merge_temp, LEDD_med_df, on = ['PATNO', 'INFODT'], how = "left")
ppmi_merge_temp = ppmi_merge_temp[['PATNO', 'EVENT_ID', 'LEDD', 'LEDD.STARTDT', 'LEDD.STOPDT']] # keep

# Df with LD formulas as variables
LD_only = ppmi_merge_temp[ppmi_merge_temp["LEDD"].str.contains("LD") == True] # Get df with only column
LD_only['STOPDT2'] = LD_only['LEDD.STOPDT'] # Initialize second stop date variable
LD_only['STOPDT2'].fillna(LD_only['LEDD.STARTDT'], inplace = True) # Fill in NaN stop dates with start
LD_only = LD_only.merge(LD_only.apply(lambda s: pd.date_range(s['LEDD.STARTDT'], s.STOPDT2, freq='MS',
LD_only.drop(['STOPDT2', 'EVENT_ID', 'LEDD.STARTDT', 'LEDD.STOPDT'], axis = 1, inplace = True) # Drop
ppmi_merge_LD_only = ppmi_merge[['PATNO', 'EVENT_ID', 'INFODT']]
ppmi_merge_LD_only = pd.merge(ppmi_merge_LD_only, LD_only, on = ['PATNO', 'INFODT'], how = "left")
ppmi_merge_LD_only = ppmi_merge_LD_only[['PATNO', 'EVENT_ID', 'LEDD']] # keep only
ppmi_merge_LD_only.rename(columns = {'LEDD' : 'LD'}, inplace = True)
ppmi_merge_LD_only = ppmi_merge_LD_only.drop_duplicates() # with ongoing stop dates some overlap when m

# Df without LD formula var - only numeric LEDD vars
ppmi_merge_temp = ppmi_merge_temp[ppmi_merge_temp["LEDD"].str.contains("LD") == False] # Remove LD x .33
ppmi_merge_temp = ppmi_merge_temp.astype({'LEDD' : 'float'})
LEDD_sum = ppmi_merge_temp.groupby(['PATNO', 'EVENT_ID'])['LEDD'].sum().reset_index() # Get sum
LEDD_sum.rename(columns = {'LEDD' : 'LEDD.sum'}, inplace = True) # Rename
ppmi_merge = pd.merge(ppmi_merge, LEDD_sum, on = ['PATNO', 'EVENT_ID'], how = "left") # Merge into ppmi_
ppmi_merge = pd.merge(ppmi_merge, ppmi_merge_LD_only, on = ['PATNO', 'EVENT_ID'], how = "left") # Merge

## If in LD column there is a (i.e. 150 + LD) * 0.33) - create a new duplicate column with just 150 var
import re
def func(row):
    if '(' in str(row['LEDD']):
        row2 = row.copy()
        # make edits to row2
        temp = row2['LEDD']
        row2['LEDD'] = re.search('\(([0-9]+)', temp).group(1)
        return pd.concat([row, row2], axis=1)
    return row

## GET LEVODOPA ONLY FOR BELOW CALCULATIONS - # FIXME - need more clarification before adding this
LEDD_med_df = pd.read_csv(ppmi_download_path + 'LEDD_Concomitant_Medication_Log.csv', skipinitialspace =
LEDD_med_df = LEDD_med_df[['PATNO', 'LEDTRT', 'LEDD', 'STARTDT', 'STOPDT']] # Keep only
search_for = ['levodopa', 'sinemet', 'rytary', 'stalevo'] # FIXME per Pavan comments but there might be
levodopa_only = LEDD_med_df[LEDD_med_df["LEDTRT"].str.contains('|'.join(search_for), case = False) == T
levodopa_only = pd.concat([func(row) for _, row in levodopa_only.iterrows()], ignore_index=True, axis=1)
levodopa_only['STOPDT2'] = levodopa_only['STOPDT'] # Initialize second stop date variable
levodopa_only['STOPDT2'].fillna(levodopa_only['STARTDT'], inplace = True) # Fill in NaN stop dates with
levodopa_only = levodopa_only.merge(levodopa_only.apply(lambda s: pd.date_range(s['STARTDT'], s.STOPDT2
levodopa_only.drop(['STOPDT2', 'STARTDT', 'STOPDT'], axis = 1, inplace = True)
ppmi_merge_levodopa_only = ppmi_merge[['PATNO', 'EVENT_ID', 'INFODT']]
ppmi_merge_levodopa_only = pd.merge(ppmi_merge_levodopa_only, levodopa_only, on = ['PATNO', 'INFODT'],

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ppmi_merge_levodopa_only = ppmi_merge_levodopa_only[['PATNO', 'EVENT_ID', 'LEDD']] # keep only
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only.drop_duplicates() # with ongoing stop dates some overlap
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only[ppmi_merge_levodopa_only['LEDD'].str.contains("LD")]
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only.astype({'LEDD' : 'float'})
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only.groupby(['PATNO', 'EVENT_ID'])['LEDD'].sum().reset_index()
ppmi_merge_levodopa_only.rename(columns = {'LEDD' : 'Levodopa.only'}, inplace = True)
ppmi_merge = pd.merge(ppmi_merge, ppmi_merge_levodopa_only, on = ['PATNO', 'EVENT_ID'], how = "left") # Merge

# Calculate LD formulas and add to LEDD column numbers # FIXME need to switch calculation from LEDD to LD
ppmi_merge['LEDD.sum'].fillna(0, inplace = True)
for row_num in range(len(ppmi_merge)):
    if ppmi_merge['LD'].loc[row_num] == 'LD x 0.33':
        ppmi_merge['LEDD.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + ppmi_merge['LD'].loc[row_num]
    elif ppmi_merge['LD'].loc[row_num] == '(150.0 + LD) x 0.33':
        ppmi_merge['LEDD.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 150.0
    elif ppmi_merge['LD'].loc[row_num] == '(600.0 + LD) x 0.33':
        ppmi_merge['LEDD.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 600.0
    elif ppmi_merge['LD'].loc[row_num] == 'LD x 0.5':
        ppmi_merge['LEDD.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.5 + ppmi_merge['LD'].loc[row_num]
    elif ppmi_merge['LD'].loc[row_num] == '(800.0 + LD) x 0.33':
        ppmi_merge['LEDD.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 800.0
    elif ppmi_merge['LD'].loc[row_num] == '(300.0 + LD) x 0.33':
        ppmi_merge['LEDD.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 300.0
    elif ppmi_merge['LD'].loc[row_num] == '(225.0 + LD) x 0.33':
        ppmi_merge['LEDD.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 225.0
    elif ppmi_merge['LD'].loc[row_num] == '(450.0 + LD) x 0.33':
        ppmi_merge['LEDD.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 450.0
ppmi_merge.drop(['LD'], axis = 1, inplace = True)
ppmi_merge = ppmi_merge.drop_duplicates() # with ongoing stop dates some overlap when merging on infodts
ppmi_merge['LEDD.sum'].replace({0 : np.NaN}, inplace = True)
ppmi_merge.drop(['Levodopa.only'], axis = 1, inplace = True) # Drop

## LEDD Medication Status - FIXME Assumption : If stop date is NA we assume therapy is ongoing
LEDD_med_df = pd.read_csv(ppmi_download_path + 'LEDD_Concomitant_Medication_Log.csv', skipinitialspace = True)
LEDD_med_df = LEDD_med_df[['PATNO', 'LEDD', 'STARTDT', 'STOPDT']] # keep only
ppmi_merge['INFODT'] = pd.to_datetime(ppmi_merge['INFODT'], format = '%m%Y', errors = 'ignore') # Change format
LEDD_med_df['STARTDT'] = pd.to_datetime(LEDD_med_df['STARTDT'], format = '%m%Y', errors = 'ignore') # Change format
LEDD_med_df['STOPDT'] = pd.to_datetime(LEDD_med_df['STOPDT'], format = '%m%Y', errors = 'ignore') # Change format
LEDD_med_df['STOPDT3'] = LEDD_med_df['STOPDT']
LEDD_med_df['STOPDT3'].fillna('01/2022', inplace = True) # ASSUMPTION fill in ongoing stopdate (12/2021)
LEDD_med_df['STOPDT3'] = pd.to_datetime(LEDD_med_df['STOPDT3'], format = '%m%Y', errors = 'ignore') # Change format
LEDD_med_df = LEDD_med_df.merge(LEDD_med_df.apply(lambda s: pd.date_range(s.STARTDT, s.STOPDT3, freq='M'), axis=1), on=['PATNO', 'LEDD'], how='left')
LEDD_med_df.rename(columns = {'LEDD' : 'LEDD.ongoing', 'STARTDT' : 'LEDD.STARTDT.ongoing', 'STOPDT3' : 'LEDD.STOPDT3.ongoing'}, inplace = True)

# Get a new df with LEDD dates merged onto corect event ids through infodts
ppmi_merge_temp = ppmi_merge[['PATNO', 'EVENT_ID', 'INFODT']]
ppmi_merge_temp = pd.merge(ppmi_merge_temp, LEDD_med_df, on = ['PATNO', 'INFODT'], how = "left")
ppmi_merge_temp = ppmi_merge_temp[['PATNO', 'EVENT_ID', 'LEDD.ongoing', 'LEDD.STARTDT.ongoing', 'LEDD.STOPDT3.ongoing']]

# Df with LD formulas as variables
LD_only = ppmi_merge_temp[ppmi_merge_temp['LEDD.ongoing'].str.contains("LD")==True] # Get df with only LD
LD_only['STOPDT2'] = LD_only['LEDD.STOPDT.ongoing'] # Initialize second stop date variable
LD_only['STOPDT2'].fillna('01/2022', inplace = True) # Fill in NaN stop dates with start dates # ASSUMPTION

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LD_only['STOPDT2'] = pd.to_datetime(LD_only['STOPDT2'], format = '%m%Y', errors = 'ignore') # change to
LD_only = LD_only.merge(LD_only.apply(lambda s: pd.date_range(s['LEDD.STARTDT.ongoing'], s.STOPDT2, freq='D'), axis=1, inplace=True))
LD_only.drop(['STOPDT2', 'EVENT_ID', 'LEDD.STARTDT.ongoing', 'LEDD.STOPDT.ongoing'], axis = 1, inplace = True)
ppmi_merge_LD_only = ppmi_merge[['PATNO', 'EVENT_ID', 'INFODT']]
ppmi_merge_LD_only = pd.merge(ppmi_merge_LD_only, LD_only, on = ['PATNO', 'INFODT'], how = "left")
ppmi_merge_LD_only = ppmi_merge_LD_only[['PATNO', 'EVENT_ID', 'LEDD.ongoing']] # keep only
ppmi_merge_LD_only.rename(columns = {'LEDD.ongoing' : 'LD'}, inplace = True)
ppmi_merge_LD_only = ppmi_merge_LD_only.drop_duplicates() # with ongoing stop dates some overlap when m

# Df without LD formula var - only numeric LEDD vars
ppmi_merge_temp = ppmi_merge_temp[ppmi_merge_temp['LEDD.ongoing'].str.contains("LD") == False] # Remove
ppmi_merge_temp = ppmi_merge_temp.astype({'LEDD.ongoing' : 'float'})
LEDD_sum = ppmi_merge_temp.groupby(['PATNO', 'EVENT_ID'])['LEDD.ongoing'].sum().reset_index() # Sum
LEDD_sum.rename(columns = {'LEDD.ongoing' : 'LEDD.ongoing.sum'}, inplace = True)
ppmi_merge = pd.merge(ppmi_merge, LEDD_sum, on = ['PATNO', 'EVENT_ID'], how = "left") # Merge into ppmi_merge
ppmi_merge = pd.merge(ppmi_merge, ppmi_merge_LD_only, on = ['PATNO', 'EVENT_ID'], how = "left") # Merge

## If in LD column there is a (i.e. 150 + LD) * 0.33 - create a new duplicate column with just 150 var
## GET LEVODOPA ONLY FOR BELOW CALCULATIONS - # FIXME - need more clarification before adding this
LEDD_med_df = pd.read_csv(ppmi_download_path + 'LEDD_Concomitant_Medication_Log.csv', skipinitialspace=True)
LEDD_med_df = LEDD_med_df[['PATNO', 'LEDTRT', 'LEDD', 'STARTDT', 'STOPDT']] # Keep only
search_for = ['levodopa', 'sinemet', 'rytary', 'stalevo'] # FIXME per Pavan comments but there might be
levodopa_only = LEDD_med_df[LEDD_med_df['LEDTRT'].str.contains('|'.join(search_for), case = False) == True]
levodopa_only = pd.concat([func(row) for _, row in levodopa_only.iterrows()], ignore_index=True, axis=1)
levodopa_only['STOPDT2'] = levodopa_only['STOPDT'] # Initialize second stop date variable
levodopa_only['STOPDT2'].fillna('01/2022', inplace = True) # Fill in NaN stop dates with start dates
levodopa_only = levodopa_only.merge(levodopa_only.apply(lambda s: pd.date_range(s['STARTDT'], s.STOPDT2, freq='D'), axis=1, inplace=True))
levodopa_only.drop(['STOPDT2', 'STARTDT', 'STOPDT'], axis = 1, inplace = True)
ppmi_merge_levodopa_only = ppmi_merge[['PATNO', 'EVENT_ID', 'INFODT']]
ppmi_merge_levodopa_only = pd.merge(ppmi_merge_levodopa_only, levodopa_only, on = ['PATNO', 'INFODT'], how = "left")
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only[['PATNO', 'EVENT_ID', 'LEDD']] # keep only
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only.drop_duplicates() # with ongoing stop dates some overlap when m
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only[ppmi_merge_levodopa_only['LEDD'].str.contains("LD")]
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only.astype({'LEDD' : 'float'})
ppmi_merge_levodopa_only = ppmi_merge_levodopa_only.groupby(['PATNO', 'EVENT_ID'])['LEDD'].sum().reset_index()
ppmi_merge_levodopa_only.rename(columns = {'LEDD' : 'Levodopa.only'}, inplace = True)
ppmi_merge = pd.merge(ppmi_merge, ppmi_merge_levodopa_only, on = ['PATNO', 'EVENT_ID'], how = "left") # Merge

# Calculate LD formulas and add to LEDD column numbers # FIXME need to switch calculation from LEDD to Levodopa
for row_num in range(len(ppmi_merge)):
    if ppmi_merge['LD'].loc[row_num] == 'LD x 0.33':
        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33
    elif ppmi_merge['LD'].loc[row_num] == '(150.0 + LD) x 0.33':
        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 150.0
    elif ppmi_merge['LD'].loc[row_num] == '(600.0 + LD) x 0.33':
        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 600.0
    elif ppmi_merge['LD'].loc[row_num] == 'LD x 0.5':
        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.5
    elif ppmi_merge['LD'].loc[row_num] == '(800.0 + LD) x 0.33':
        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 800.0
    elif ppmi_merge['LD'].loc[row_num] == '(300.0 + LD) x 0.33':
        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 300.0
    elif ppmi_merge['LD'].loc[row_num] == '(225.0 + LD) x 0.33':
        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) * 0.33 + 225.0

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        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) *
    elif ppmi_merge['LD'].loc[row_num] == '(450.0 + LD) x 0.33' :
        ppmi_merge['LEDD.ongoing.sum'].loc[row_num] = float(ppmi_merge['Levodopa.only'].loc[row_num]) *
#ppmi_merge['LEDD.STOPDT.ongoing'].replace({'01/2022' : np.NaN}, inplace = True) # Replace filler stop
ppmi_merge.drop(['LD'], axis = 1, inplace = True)

## Comorbidities # FIXME not useful column
comorbid_df = pd.read_csv(ppmi_download_path + 'Medical_Conditions_Log.csv', skipinitialspace=True) # M
comorbid_df.replace({';' : ','}, regex = True, inplace = True) # Replace ';' with ','
comorbid_df = comorbid_df[['PATNO', 'EVENT_ID', 'MHDIAGDT', 'MHTERM']] # keep only
comorbid_df['MHTERM'] = comorbid_df['MHTERM'].str.capitalize() # Capitalize all MHTERM names
comorbid_df['MHDIAGDT'].fillna('NA', inplace = True) # If no diagnosis date - fill in with NA
comorbid_df['MHDIAGDT'] = '(' + comorbid_df['MHDIAGDT'].astype(str) + ')' # Put parentheses around diagn
comorbid_df = comorbid_df.astype({'MHDIAGDT' : 'str'}) # Change date to string
comorbid_df = merge_columns(comorbid_df, ['MHTERM', 'MHDIAGDT'], 'Medical.History.Description(Diagnosis
comorbid_df = comorbid_df.groupby(['PATNO', 'EVENT_ID'])['Medical.History.Description(Diagnosis.Date)'].
ppmi_merge = pd.merge(ppmi_merge, comorbid_df, on = ['PATNO', 'EVENT_ID'], how = "outer")

# ## NEW Comorbidities START
# comorbid_df = pd.read_csv(ppmi_download_path + 'Medical_Conditions_Log.csv', skipinitialspace=True) #
# comorbid_df.replace({';' : ','}, regex = True, inplace = True) # Replace ';' with ','

# ## Replace comorbid MHTERM columns with common names
# # FIXME still need to take care of other medical conditions such as - tinnitus, vertigo, cholesterol,
# comorbid_df = pd.read_csv('/Users/areardon/Desktop/ppmi_merge/PPMI_Study_Data_Download/Medical_Condit
# comorbid_df['MHTERM'] = comorbid_df['MHTERM'].replace({'s': "s"})
# comorbid_df['MHTERM'] = comorbid_df['MHTERM'].str.lower() # Capitalize all medication names
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*parkin.*$)', 'parkinson')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*diabetes.*$)', 'diabetes') # FIXME
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*sulfa.*$)', 'sulfa allergy')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*nickel allergy.*$)', 'nickel allergy')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*herpes.*$)', 'herpes') # FIXME
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*lactose.*$)', 'lactose intolerance')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*hypothyroid.*$)', 'hypothyroidism')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*back pain.*$)', 'back pain') # FIXME can
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*rem .*$)', 'rem sleep behavior disorder')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*kidney stone.*$)', 'kidney stones')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*migraine.*$)', 'migraines')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*near.*$)', 'myopia') # FIXME
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*headache.*$)', 'migraines') # FIXME chro
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*covid.*$)', 'covid-19')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*cramps.*$)', 'muscle cramps')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*attention deficit.*$)', 'adhd')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*high blood pressure (hypertension).*$)',
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*trigger finger.*$)', 'trigger finger')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*squamous.*$)', 'squamous cell carcinoma')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*osteoarthritis.*$)', 'osteoarthritis')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*fatigue.*$)', 'fatigue') # FIXME
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*sjögren.*$)', 'sjogren\'s syndrome')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*erythromycin.*$)', 'erythromycin allergy')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*osteo arthritis.*$)', 'osteoarthritis')

```

```

# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*bursitis.*$)', 'bursitis')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*shoulder injury.*$)', 'shoulder injury')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*gaucher.*$)', 'gauchers disease')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*knee replacement.*$)', 'knee replacement')
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*meniscus.*$)', 'torn meniscus') # FIXME
# comorbid_df['MHTERM'] = comorbid_df.MHTERM.str.replace(r'(^.*cataract.*$)', 'cataracts') # FIXME can
# comorbid_df['MHTERM'] = comorbid_df['MHTERM'].str.strip()
# comorbid_df['MHTERM'].replace({'high cholestrol': 'high cholesterol', 'low red blood cell count': 'low red blood cell count'})

## Get list of top n comorbidities/ medical history descriptions
# n = 20 # FIXME how many of these to keep?
# selection = comorbid_df['MHTERM'].value_counts()[:n].index.tolist()
# comorbid_df = comorbid_df.loc[comorbid_df['MHTERM'].isin(selection)]

# comorbid_df = comorbid_df[['PATNO', 'EVENT_ID', 'MHTERM']] # keep only
# comorbid_df['MHTERM'] = comorbid_df['MHTERM'].str.capitalize() # Capitalize all MHTERM names
# comorbid_df['MHTERM'].fillna('NA', inplace = True) # If no diagnosis date - fill in with NA
# comorbid_df['MHTERM'] = '(' + comorbid_df['MHTERM'].astype(str) + ')' # Put parentheses around diagnosis
# comorbid_df = comorbid_df.astype({'MHTERM' : 'str'}) # Change date to string
# comorbid_df = merge_columns(comorbid_df, ['MHTERM', 'MHTERM'], 'Medical.History.Description(Diagnosis.Date)')
# comorbid_df = comorbid_df.groupby(['PATNO', 'EVENT_ID'])['Medical.History.Description(Diagnosis.Date)'].first().reset_index()
# ppmi_merge = pd.merge(ppmi_merge, comorbid_df, on = ['PATNO', 'EVENT_ID'], how = "outer")
## NEW COMORBIDITIES END

## Education (in years)
education_df = pd.read_csv(ppmi_download_path + 'Socio-Economics.csv', skipinitialspace = True) # Education
education_df = education_df[['PATNO', 'EVENT_ID', 'EDUCYRS']] # Keep info
education_df.rename(columns = {'EDUCYRS' : 'Education.Years'}, inplace = True) # Rename
education_df = education_df.groupby('PATNO').mean().reset_index() # Take the mean of education years if multiple
ppmi_merge = pd.merge(ppmi_merge, education_df, on = ['PATNO'], how = "outer") # Merge

## Add in 'Analytic.Cohort' column
analytic_cohort = ppmi_merge['PATNO'].isin(analytic_cohort_subids)
ppmi_merge['Analytic.Cohort'] = '' # Initialize Analytic.Cohort column
ppmi_merge['Analytic.Cohort'].loc[analytic_cohort] = 'Analytic Cohort'
ppmi_merge['Analytic.Cohort'].fillna('Not Analytic Cohort', inplace = True)

## Reindex
ppmi_merge = ppmi_merge.reindex(columns = ['PATNO', 'EVENT_ID', 'INFODT', 'Enroll.Diagnosis', 'Enroll.Date'])

## Add in other csvs
ppmi_merge = merge_new_csv(ppmi_merge, 'Modified_Boston_Naming_Test.csv', ['PATNO', 'EVENT_ID', 'MBSTNS'])
ppmi_merge = merge_new_csv(ppmi_merge, 'Clock_Drawing.csv', ['PATNO', 'EVENT_ID', 'CLCKTOT'], merge_on = 'PATNO')
ppmi_merge = merge_new_csv(ppmi_merge, 'Benton_Judgement_of_Line_Orientation.csv', ['PATNO', 'EVENT_ID', 'BJLO'])
ppmi_merge = merge_new_csv(ppmi_merge, 'Letter_-_Number_Sequencing.csv', ['PATNO', 'EVENT_ID', 'LNS_TOT'])
ppmi_merge = merge_new_csv(ppmi_merge, 'Modified_Semantic_Fluency.csv', ['PATNO', 'EVENT_ID', 'DVS_SFTAI'])
ppmi_merge = merge_new_csv(ppmi_merge, 'Hopkins_Verbal_Learning_Test_-_Revised.csv', ['PATNO', 'EVENT_ID', 'HVLTR'])
ppmi_merge = merge_new_csv(ppmi_merge, 'Symbol_Digit_Modality_Test.csv', ['PATNO', 'EVENT_ID', 'SDMTOT'])
ppmi_merge.rename(columns = {'CLCKTOT' : 'Clock.Drawing.Total', 'JLOTOTCALC' : 'JOLO.Total', 'LNS_TOTRAI' : 'LNS.Total'})

## REM Sleep behavior disorder questionnaire
ppmi_merge = merge_new_csv(ppmi_merge, 'REM_Sleep_Behavior_Disorder_Questionnaire.csv', ['PATNO', 'EVENT_ID', 'REMSQ'])

```

```

ppmi_merge.rename(columns = {'PAG_NAME' : 'REM.Sleep.Behavior.Disorder.Page.Name'}, inplace = True)
ppmi_merge['REM.Sleep.Behavior.Disorder.Page.Name'].replace({'REMSLEEP' : 'REM Sleep Behavior Disorder (
ppmi_merge['PTCGBOTH'].replace({1 : 'Participant', 2 : 'Caregiver', 3: 'Participant and Caregiver'}, in
rem_list = ['DRMVIVID', 'DRMAGRAC', 'DRMNOCTB', 'SLPLMBMV', 'SLPINJUR', 'DRMVERBL', 'DRMFIGHT', 'DRMUMV
ppmi_merge['RBDTotal.REM'] = ppmi_merge[rem_list].sum(axis = 1) # Add an RBDTotal.REM column
ppmi_merge = decode_0_1_no_yes(ppmi_merge, rem_list)
ppmi_merge.rename(columns = {'PTCGBOTH' : 'Sleep.Behavior.Source.REM', 'DRMVIVID' : 'Vivid.Dreams.REM' ,

#### IMAGING INFO ####
# FIXME - laterality issue? SUV ?
ppmi_merge = merge_new_csv(ppmi_merge, 'DaTScan_Analysis.csv', ['PATNO', 'EVENT_ID', 'DATSCAN_DATE', 'DATSCAN
ppmi_merge = merge_new_csv(ppmi_merge, 'DaTScan_Imaging.csv', ['PATNO', 'EVENT_ID', 'PAG_NAME', 'INFODT', 'D
ppmi_merge['DATSCAN'].replace({0 : 'Not Completed', 1: 'Completed', 2 : 'Completed using a previously ac
ppmi_merge['SCNLOC'].replace({ 1 : 'Site', 2 : 'IND'}, inplace = True )
ppmi_merge['SCNINJCT'].replace({ 1 : 'DaTSCAN', 2 : 'Beta-CIT'}, inplace = True)
ppmi_merge['VSINTRPT'].replace({ '1' : 'Consistent with evidence' , '2' : 'Not consistent with evidence
ppmi_merge['VSRPTTELG'].replace({1 : 'Eligible', 2 : 'Not eligible'}, inplace = True)
ppmi_merge.rename(columns = {'INFODT_x' : 'INFODT', 'INFODT_y' : 'DaTScan.INFODT', 'SCNLOC' : 'Location
ppmi_merge['INFODT'].fillna(ppmi_merge['DaTScan.INFODT'], inplace = True) # Fill in NaN infodts with da

# FIXME - After merging in datscan files, duplicate event ids being created one with DATSCAN as "Comple
duplicate_datscan = ppmi_merge[['PATNO', 'EVENT_ID']].duplicated(keep = False) # Find locations of True
duplicate_datscan_index = ppmi_merge[duplicate_datscan == True].index.tolist() # Get index of duplicate
dup_subid_list = [] # Initialize duplicate subid list variable
[dup_subid_list.append(index) for index in duplicate_datscan_index if ppmi_merge['DATSCAN'][index] == '

ppmi_merge = ppmi_merge.reset_index(drop = True)
[ppmi_merge.drop(index = i, axis = 1, inplace = True) for i in reversed(dup_subid_list) if ppmi_merge['I

## MRI.csv

mri_df = pd.read_csv(ppmi_download_path + 'Magnetic_Resonance_Imaging_MRI.csv', skipinitialspace=True)
mri_df = mri_df[['PATNO', 'EVENT_ID', 'INFODT', 'MRICMPLT', 'MRIWDTI', 'MRIWRSS', 'MRIRSLT', 'MRIRSSDF
mri_df['MRICMPLT'].replace({0 : 'Not Completed', 1 : 'Completed'}, inplace = True) # Decode
mri_df = decode_0_1_no_yes(mri_df, ['MRIWDTI', 'MRIWRSS', 'MRIRSSDF'])
mri_df['MRIRSLT'].replace({1 : 'Normal', 2 : 'Abnormal, not clinically significant', 3 : 'Abnormal, cli
mri_df.rename(columns = { 'INFODT' : 'Image.Acquisition.Date', 'MRICMPLT' : 'MRI.Completed', 'MRIWDTI'

# FIXME Some subjects had two baseline rows - 1 with incomplete MRI.Completed and 1 with complete as MR
duplicate_mri = mri_df[['PATNO', 'EVENT_ID']].duplicated(keep = False) # Find locations of True for dupl
duplicate_mri_index = mri_df[duplicate_mri == True].index.tolist() # Get index of duplicates
dup_subid_list = [] # Initialize duplicate subid list variable
[dup_subid_list.append(index) for index in duplicate_mri_index if mri_df['MRI.Completed'][index] == 'No

mri_df = mri_df.reset_index(drop = True)
[mri_df.drop(index = i, axis = 1, inplace = True) for i in reversed(dup_subid_list) if mri_df['MRI.Comp

ppmi_merge = pd.merge(ppmi_merge, mri_df, on = ['PATNO', 'EVENT_ID'], how = "outer")

## For all subjects - see if we have T1 images in ppmi-image-data bucket for given dates

```



```

def search_s3(bucket, prefix, search_string):
    client = boto3.client('s3', region_name="us-east-1")
    paginator = client.get_paginator('list_objects')
    pages = paginator.paginate(Bucket=bucket, Prefix=prefix)
    keys = []
    for page in pages:
        contents = page['Contents']
        for c in contents:
            keys.append(c['Key'])
    if search_string:
        keys = [key for key in keys if search_string in key]
    return keys
keys = search_s3('invicro-ia-object-repository', 'refined/ppmi/data/PPMI/', 'T1w/') # PPMI1.0 and PPMI2.0

# Set a variable in ppmi_merge 'Subid.Date.TEMP' that is the subid and image acquisition date to match
ppmi_merge['Subid.Date.TEMP'] = ''
for row_num in range(len(ppmi_merge['Image.Acquisition.Date'])) :
    if isinstance(ppmi_merge['Image.Acquisition.Date'].loc[row_num], str) :
        ppmi_merge['Subid.Date.TEMP'].iloc[row_num] = ppmi_merge["PATNO"].iloc[row_num].astype(str) + ' ' + ppmi_merge['Image.Acquisition.Date'].loc[row_num].astype(str)
subid_date_ordered = ppmi_merge['Subid.Date.TEMP'].dropna().tolist() # Make column into list

# Get keys of subjects in ppmi-image-data bucket with T1w/ image folder
woutppmi = [key.split('PPMI/')[1] for key in keys] # Remove PPMI/ from key
woutt1w = [key.split('/T1w/')[0] for key in woutppmi] # Remove 'T1w' from key
s3woutdate = [current_subid_date[:-2] for current_subid_date in woutt1w] # Remove the day from date - w
matches = [current_subid_date for current_subid_date in subid_date_ordered if current_subid_date in s3woutdate]

# Add in T1 s3 Info
s3_df = pd.DataFrame(columns = ['PATNO', 'EVENT_ID', 'T1.s3.Image.Name']) # create s3_df dataframe
for current_subid_date_temp in matches :
    for image_id in woutppmi :
        if current_subid_date_temp in image_id and image_id.endswith('.nii.gz'):
            image_id_split = image_id.split('/')
            s3_df = s3_df.append({'PATNO' : image_id_split[0], 'EVENT_ID' : ppmi_merge.loc[ppmi_merge['Subid.Date.TEMP'] == image_id_split[0], 'EVENT_ID'].iloc[0]})

# Create a column in s3_df for just object name so later we can merge T1 file with object name
s3_df['PATNO'] = s3_df['PATNO'].astype(int) # Needed for merge on PATNO
s3_df['Image_ID_merge'] = ''
for row_num in range(len(s3_df['T1.s3.Image.Name'])) :
    image0 = s3_df['T1.s3.Image.Name'].iloc[row_num].split('.')[0] # Get name of image before .nii.gz
    s3_df['Image_ID_merge'].iloc[row_num] = image0.split('-')[4] # Get ImageID from s3 filename and put
ppmi_merge = pd.merge(ppmi_merge, s3_df, on = ['PATNO', 'EVENT_ID'], how = "outer")

#### MDS-UPDRS Scores UPDRS 1-4 as Numeric Variables ####
def setup_updrs_df(updrs_filename, drop_cols, decode_dict, rename_col_dict) :
    updrs_df = pd.read_csv(ppmi_download_path + updrs_filename, skipinitialspace = True)
    updrs_df.drop(drop_cols, axis = 1, inplace = True)
    updrs_df['PAG_NAME'].replace(decode_dict, inplace = True)
    updrs_df.rename(columns = rename_col_dict, inplace = True)
    return updrs_df

updrs_part1_df = setup_updrs_df('MDS-UPDRS_Part_I.csv', ['ORIG_ENTRY', 'LAST_UPDATE', 'REC_ID'], {'NUPDRS1': 'UPDRS1', 'NUPDRS2': 'UPDRS2', 'NUPDRS3': 'UPDRS3', 'NUPDRS4': 'UPDRS4'})

```



```

updrs_part1_pq_df = setup_updrs_df('MDS-UPDRS_Part_I_Patient_Questionnaire.csv', ['ORIG_ENTRY', 'LAST_UPDATE'])
updrs_part2_pq_df = setup_updrs_df('MDS-UPDRS_Part_II_Patient_Questionnaire.csv', ['ORIG_ENTRY', 'LAST_UPDATE'])
updrs_part3_dos_df = setup_updrs_df('MDS-UPDRS_Part_III_ON_OFF_Determination__Dosing.csv', ['ORIG_ENTRY', 'LAST_UPDATE'])
updrs_part4_motor_df = setup_updrs_df('MDS-UPDRS_Part_IV_Motor_Complications.csv', ['ORIG_ENTRY', 'LAST_UPDATE'])

# UPDRS Part 3
updrs_part3_df = pd.read_csv(ppmi_download_path + 'MDS-UPDRS_Part_III.csv', skipinitialspace = True)
updrs_part3_df.drop(['ORIG_ENTRY', 'LAST_UPDATE', 'REC_ID', 'INFO_DT'], axis = 1, inplace = True)
updrs_part3_df.rename(columns = {'PAG_NAME' : 'UPDRS.Part3.Page.Name'}, inplace = True)

# Change all UPDRS dataframe cols begin with 'N' to floats
updrs_list = [updrs_part1_df, updrs_part1_pq_df, updrs_part2_pq_df, updrs_part3_df, updrs_part3_dos_df, updrs_part4_motor_df]
for df in updrs_list :
    for col_name in df :
        if col_name.startswith('N') :
            df[col_name] = pd.to_numeric(df[col_name], errors = 'coerce', downcast = 'float')

# Create a copy of each dataframe to use later to create categorical versions of variables
updrs_part1_df_copy = updrs_part1_df.copy()
updrs_part1_pq_df_copy = updrs_part1_pq_df.copy()
updrs_part2_pq_df_copy = updrs_part2_pq_df.copy()
updrs_part3_df_copy = updrs_part3_df.copy()
updrs_part3_dos_df_copy = updrs_part3_dos_df.copy()
updrs_part4_motor_df_copy = updrs_part4_motor_df.copy()

# For all UPDRS df columns - add the respective extension for which UPDRS assessment it is
def add_extension_to_column_names(df, skip_col_list, ext):
    for col_name in df :
        if col_name not in skip_col_list :
            df.rename(columns = {str(col_name) : str(col_name) + ext }, inplace = True)
    return df

# Add extensions to updrs dfs
updrs_list_str = ['.UPDRS1', '.UPDRS1', '.UPDRS2', '.UPDRS3', '.UPDRDOSE', '.UPDRS4'] # extensions
for i in range(len(updrs_list)) :
    updrs_list[i] = add_extension_to_column_names(updrs_list[i], ['PATNO', 'EVENT_ID', 'INFO_DT'], updrs_list_str[i])

# Create one df for updrs_numeric
updrs_numeric = pd.merge(updrs_part1_df, updrs_part1_pq_df , on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_numeric = pd.merge(updrs_numeric, updrs_part2_pq_df , on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_numeric = pd.merge(updrs_numeric, updrs_part3_df, on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_numeric = pd.merge(updrs_numeric, updrs_part3_dos_df , on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_numeric = pd.merge(updrs_numeric, updrs_part4_motor_df, on = ['PATNO', 'EVENT_ID'], how = "outer")

## Keep only variables in .Num that are numeric variables
numeric_vars = []
for col_name in updrs_numeric :
    if col_name.startswith('NP') or col_name.startswith('PATNO') or col_name.startswith('EVENT') or col_name.startswith('INFO_DT') :
        numeric_vars.append(col_name)
updrs_numeric = updrs_numeric[numeric_vars]

# Rename columns in updrs_numeric
updrs_numeric.rename(columns = {'NHY.UPDRS3' : 'Hoehn.and.Yahr.Stage.UPDRS3', 'NP3BRADY.UPDRS3' : 'Global.Cognition.UPDRS3'})

```

```

updrs_numeric.to_csv('/Users/areardon/Desktop/updrs_numeric.csv')
## Subscore info from :
# https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7242837/
# https://www.movementdisorders.org/MDS-Files1/PDFs/Rating-Scales/MDS-UPDRS_English_FINAL.pdf

## Add Brady-Rigidity, Tremor and PIGD Subscores
def calculate_subscores(df, new_col_name, subscore_list) :
    subscore_only = df[subscore_list] # extract df of only cols we are interested in for current subscore
    df[new_col_name] = 0 # initialize new col for subscore
    idx = subscore_only.loc[pd.isnull(subscore_only).any(1), :].index
    df[new_col_name] = subscore_only.sum(axis = 1) # Get sum of subscore
    df[new_col_name].iloc[idx] = np.nan # Replace rows that should be nan with nan
    return df

updrs_numeric = calculate_subscores(updrs_numeric, 'Brady.Rigidity.Subscore.UPDRS3', ['Rigidity.Neck', 'Rigidity.Arms', 'Rigidity.Legs'])
updrs_numeric = calculate_subscores(updrs_numeric, 'Tremor.Subscore.UPDRS3', ['Tremor.UPDRS2', 'Postural.Stability'])
updrs_numeric = calculate_subscores(updrs_numeric, 'PIGD.Subscore.UPDRS3', ['Walking.Difficulty.UPDRS2', 'Walking.Speed'])
updrs_numeric = add_extension_to_column_names(updrs_numeric, ['PATNO', 'EVENT_ID', 'PIGD.Subscore.UPDRS3'])

### UPDRS CATEGORICAL ###
# UPDRS3 (four dataframes) decode and rename in loop
updrs_part3_df_copy = decode_none2severe(updrs_part3_df_copy, ['NP3SPCH', 'NP3FACXP', 'NP3RIGN', 'NP3RIG'])
updrs_part3_df_copy['DBS_STATUS'].replace({0 : 'OFF', 1 : 'ON'}, inplace = True)
updrs_part3_df_copy = decode_0_1_no_yes(updrs_part3_df_copy, ['DYSKPRE', 'DYSKIRAT'])
updrs_part3_df_copy['NHY'].replace({0 : 'Asymptomatic', 1 : 'Unilateral Movement Only', 2 : 'Bilateral incoordination'})
updrs_part3_df_copy['PDTRTMNT'].replace({0 : 'OFF', 1 : 'ON'}, inplace = True)
updrs_part3_df_copy.rename(columns = {'DBS_STATUS' : 'Deep.Brain.Stimulation.Treatment', 'NP3SPCH' : 'NP3SPCH'}, inplace = True)
updrs_part3_df_copy = add_extension_to_column_names(updrs_part3_df_copy, ['PATNO', 'EVENT_ID', 'INFODT'])

# Combine pd med dose date and time into one column
updrs3_merge = merge_columns(updrs_part3_df_copy, ['PDMEDDT.UPDRS3', 'PDMEDTM.UPDRS3'], 'Most.Recent.PD.Med.Dose.Date')
updrs3_merge.to_csv("/Users/areardon/Desktop/updrs3_merge.csv")
# Rename page name variables
# FIXME upDRS
#updrs3_merge['PAG_NAME'].replace({'NUPDRS3' : 'MDS-UPDRS Part III (No Treatment)', 'NUPDRS3A' : 'MDS-UPDRS Part III (No Treatment)'})

# Merge all UPDRS dfs together
updrs_cat = pd.merge(updrs_part1_df_copy, updrs_part1_pq_df_copy, on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_cat = pd.merge(updrs_cat, updrs_part2_pq_df_copy, on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_cat = pd.merge(updrs_cat, updrs3_merge, on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_cat = pd.merge(updrs_cat, updrs_part3_dos_df_copy, on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_cat = pd.merge(updrs_cat, updrs_part4_motor_df_copy, on = ['PATNO', 'EVENT_ID'], how = "outer")

## Decode UPDRS df variables
# MNDS_UPDRS Part 1
updrs_cat['UPDRS.Part1.Source'].replace({1 : 'Patient', 2 : 'Caregiver', 3 : 'Patient and Caregiver'})
updrs_cat = decode_none2severe(updrs_cat, ['NP1COG', 'NP1HALL', 'NP1DPRS', 'NP1ANXS', 'NP1APAT', 'NP1DD'])
updrs_cat.rename(columns = {'UPDRS.Part1.Source' : 'UPDRS.Part1.Source.UPDRS1', 'NP1COG' : 'Cognitive.Instrumentation'})

# MDS_UPDRS_Part 1 Patient Questionnaire
updrs_cat['UPDRS.Part1.PQ.Source'].replace({1 : 'Patient', 2 : 'Caregiver', 3 : 'Patient and Caregiver'})
updrs_cat = decode_none2severe(updrs_cat, ['NP1SLPN', 'NP1SLPD', 'NP1PAIN', 'NP1URIN', 'NP1CNST', 'NP1L'])

```

```

updrs_cat.rename(columns = {'UPDRS.Part1.PQ.Source' : 'UPDRS.Part1.Patient.Questionnaire.Source.UPDRS1'})

# MDS_UPDRS Part 2
updrs_cat['UPDRS.Part2.Source'].replace({1 : 'Patient', 2 : 'Caregiver', 3 : 'Patient and Caregiver'}, 1)
updrs_cat = decode_none2severe(updrs_cat, ['NP2SPCH', 'NP2SALV', 'NP2SWAL', 'NP2EAT', 'NP2DRES', 'NP2HYGN', 'NP2SPCH'])
updrs_cat.rename(columns = {'UPDRS.Part2.Source' : 'UPDRS.Part2.Source.UPDRS2', 'NP2SPCH' : 'UPDRS2.Spe

# MDS_UPDRS Part 3 On OFF determination Dosing
updrs_cat = decode_0_1_no_yes(updrs_cat, ['RMEXAM', 'DBSYN', 'OFFEXAM', 'OFFEXAM', 'ONEXAM'])
updrs_cat['RMTOFFRSN'].replace({1 : 'ON state not reached', 2 : 'Scheduling issues', 3 : 'Other reason'})
updrs_cat['ONOFFORDER'].replace({1 : 'OFF', 2 : 'ON'}, inplace = True)
updrs_cat['RMONOFF'].replace({1 : 'OFF', 2 : 'ON'}, inplace = True)
updrs_cat['ONNORSN'].replace({1 : 'ON state not reached', 2 : 'Scheduling issues', 3 : 'Other reason'})
updrs_cat['PDMEDYN'].replace({0 : False, 1 : True}, inplace = True)
updrs_cat = merge_columns(updrs_cat, ['ONPDMEDDT', 'ONPDMEDTM'], 'Most.Recent.PD.Med.Dose.Date.Time.Before')
updrs_cat['OFFNORSN'].replace({1 : 'Disease severity preventing turning off of DBS', 2 : 'Did not bring
updrs_cat = merge_columns(updrs_cat, ['OFFPDMEDDT', 'OFFPDMEDTM'], 'Most.Recent.PD.Med.Dose.Date.Time.Before')
updrs_cat.rename(columns = {'RMONOFF' : 'MDS-UPDRS.Part3.Remote.Visit.ON.or.OFF.UPDRDOSE', 'Most.Recent

# MDS_UPDRS Part 4
updrs_cat['NP4WDYSK'].replace({0 : 'No dyskinesias', 1 : 'Slight: <= 25% of waking day', 2 : 'Mild : 26-50
updrs_cat['NP4DYSKI'].replace({0 : 'No dyskinesias', 1 : 'Slight', 2 : 'Mild', 3 : 'Moderate', 4 : 'Severe
updrs_cat['NP4OFF'].replace({0 : 'Normal: No OFF time', 1 : 'Slight: <= 25% of waking day', 2 : 'Mild :
updrs_cat['NP4FLCTI'].replace({0 : 'Normal', 1 : 'Slight', 2 : 'Mild', 3 : 'Moderate', 4 : 'Severe'}, inplace = True)
updrs_cat['NP4DYSTN'].replace({0 : 'No dystonia', 1 : 'Slight: <= 25% of time in OFF state', 2 : 'Mild :
updrs_cat['NP4FLCTX'].replace({0 : 'Normal', 1 : 'Slight', 2 : 'Mild', 3 : 'Moderate', 4 : 'Severe'}, inplace = True)
updrs_cat['RMNOPRT3'].replace({1 : 'Visit was not conducted with video', 2 : 'Scheduling issues', 3 : 'Other reason'})
updrs_cat.rename(columns = {'RMNOPRT3' : 'Reason.UPDRSPart3.Not.Administered.Remote.Visit.UPDRDOSE', 'NP4
updrs_cat = add_extension_to_column_names(updrs_cat, ['PATNO', 'EVENT_ID', 'INFODT'], '.Cat') # Add a .Cat
updrs_cat.to_csv("/Users/areardon/Desktop/updrs_cat.csv")
# Create one df with UPDRS scores .Num (numeric) and all UPDRS scores .Cat (categorical)
updrs_cat.drop(['INFODT', 'PATNO', 'EVENT_ID'], axis = 1, inplace = True)
updrs_merged = pd.concat([updrs_cat, updrs_numeric], axis = 1)
#updrs_merged = pd.merge(updrs_cat, updrs_numeric, on = ['PATNO', 'EVENT_ID'], how = "outer")
updrs_merged.replace({'UR' : np.nan}, inplace = True) # Unable to Rate --> nan
ppmi_merge = pd.merge(ppmi_merge, updrs_merged, on = ['PATNO', 'EVENT_ID'], how = "outer")
ppmi_merge.to_csv('/Users/areardon/Desktop/ppmi_mergex.csv')

#### CLEAN UP DF ####
ppmi_merge.drop(['Subid.Date.TEMP'], axis = 1, inplace = True)
ppmi_merge.rename(columns = {'EVENT_ID' : 'Event.ID', 'INFODT' : 'Event.ID.Date', 'Medication' : 'Medication'})

# Change names of event ids to be indicative of months
ppmi_merge['Event.ID'].replace({'FNL' : 'Final Visit', 'BL' : 'Baseline', 'SC' : 'Screening', 'LOG' : 'Log'}, inplace = True)
ppmi_merge.to_csv('/Users/areardon/Desktop/ppmi_mergex.csv')

race_list = ['African.Berber.Race', 'Ashkenazi.Jewish.Race', 'Basque.Race', 'Hispanic.Latino.Race', 'Asian']
for col_name in race_list :
    ppmi_merge[col_name].replace({0 : 'No', 1 : 'Yes', 2 : np.NaN}, inplace = True)

# Fill in cells that are NA with subject information that we know from other event ids/rows for fixed variables
fixed_var_list = ['Dominant.Side.Disease', 'Enroll.Diagnosis', 'Enroll.Subtype', 'Consensus.Diagnosis',

```

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for col_name in fixed_var_list :
    ppmi_merge[col_name].fillna('NA', inplace = True)
    for row_num in range(len(ppmi_merge[col_name])):
        if ppmi_merge[col_name].iloc[row_num] == 'NA' : # if any entry is NA
            current_sub = ppmi_merge['PATNO'].iloc[row_num] # get current subid
            fixed_var_value = ppmi_merge.loc[(ppmi_merge['PATNO'] == current_sub ) & (ppmi_merge[col_name] == 'NA')]
            if fixed_var_value.any() :
                ppmi_merge.loc[row_num, col_name] = fixed_var_value[0] # fill in baseline value at NA
ppmi_merge = ppmi_merge.rename(columns = {'PATNO' : 'Subject.ID'})
ppmi_merge.replace({'NA' : np.nan}, inplace = True)

race_list = ['African.Berber.Race', 'Ashkenazi.Jewish.Race', 'Basque.Race', 'Hispanic.Latino.Race', 'Asian']
for col_name in race_list :
    ppmi_merge[col_name].replace({'No' : False , 'Yes' : True, 2 : np.NaN}, inplace = True)

#### GENETICS INFO ####
lrrk2_genetics_df = pd.read_csv(genetics_path + 'lrrk2_geno_012_mac5_missing_geno.csv')
scna_genetics_df = pd.read_csv(genetics_path + 'scna_geno_012_mac5_missing_geno.csv')
apoe_genetics_df = pd.read_csv(genetics_path + 'apoe_geno_012_mac5_missing_geno.csv')
tmem_genetics_df = pd.read_csv(genetics_path + 'tmem175_geno_012_mac5_missing_geno.csv')
gba_genetics_df = pd.read_csv(genetics_path + 'gba_geno_012_mac5_missing_geno.csv')

def format_genetics_df(genetics_df : pd.DataFrame ) :
    """
    Format genetics_df to make merge-able with ppmi_merge
    """
    genetics_df.drop(['COUNTED', 'ALT', 'SNP', '(C)M'], axis = 1, inplace = True) # Remove unnecessary

    # Change column names to be just subid
    for col in genetics_df:
        if '_' in col :
            subid = int(col.split('_')[-1])
            genetics_df.rename(columns = {col : subid}, inplace = True)

    # Combine CHR and POS columns
    genetics_df['CHR'] = 'CHR' + genetics_df['CHR'].astype(str) # Need to be strings before you use merge
    genetics_df['POS'] = 'POS' + genetics_df['POS'].astype(str) # Need to be strings before you use merge
    genetics_df = merge_columns(genetics_df, ['CHR', 'POS'], 'Chromosome.Position', '.')

    # Pivot df so position is column name and subid is row
    genetics_df = genetics_df.T # Transpose df so that rows are subid
    genetics_df.rename(columns = genetics_df.iloc[-1], inplace = True) # Move Chr.Pos to column names
    genetics_df.index.names = ['Subject.ID'] # Rename index to 'Subject.ID'
    genetics_df = genetics_df.drop(['Chromosome.Position'], axis = 0) # Drop last row (repeat of col name)
    genetics_df = genetics_df.reset_index(drop = False)

    return genetics_df

lrrk2_genetics_df_formatted = format_genetics_df(lrrk2_genetics_df)
scna_genetics_df_formatted = format_genetics_df(scna_genetics_df)
apoe_genetics_df_formatted = format_genetics_df(apoe_genetics_df)
tmem_genetics_df_formatted = format_genetics_df(tmem_genetics_df)

```

```

gba_genetics_df_formatted = format_genetics_df(gba_genetics_df)

# Merge genetics dataframes together to create one genetics_df
genetics_df = pd.merge(lrrk2_genetics_df_formatted, scna_genetics_df_formatted, on = ['Subject.ID'], how = "outer")
genetics_df = pd.merge(genetics_df, apoe_genetics_df_formatted, on = ['Subject.ID'], how = "outer")
genetics_df = pd.merge(genetics_df, tmem_genetics_df_formatted, on = ['Subject.ID'], how = "outer")
genetics_df = pd.merge(genetics_df, gba_genetics_df_formatted, on = ['Subject.ID'], how = "outer")

# Change genetics col names int to float (remove .0 in all 'CHR.POS' columns)
for col_name in genetics_df :
    if col_name.startswith('CHR'):
        genetics_df[col_name] = genetics_df[col_name].fillna(-9999.0)
        genetics_df[col_name] = genetics_df[col_name].astype(int)
genetics_df.replace({-9999.0 : 'NA'}, inplace = True)

# Merge ppmi_merge with genetics df
ppmi_merge_genetics = pd.merge(ppmi_merge, genetics_df, on = 'Subject.ID', how = "outer")

#### T1 Info - Taylor's File ####
ppmi_t1_df = pd.read_csv(invicro_data_path + 'ppmi_mergewide_t1.csv') # Read in Taylor's T1 results file
ppmi_t1_df.rename(columns = {'u_hier_id_OR': 'Subject.ID'}, inplace = True) # Rename subject id column

# Create a column for object name to merge on with ppmi_merge
ppmi_t1_df['Image_ID_merge'] = ''
for row_num in range(len(ppmi_t1_df['ImageID'])) :
    image_id = ppmi_t1_df['ImageID'].iloc[row_num].split('-')[2]
    ppmi_t1_df['Image_ID_merge'].iloc[row_num] = image_id

# Merge ppmi_merge_genetics with t1 info
ppmi_merge = pd.merge(ppmi_merge_genetics, ppmi_t1_df, on = ['Subject.ID', 'Image_ID_merge'], how = "left")
ppmi_merge.drop(['Image_ID_merge'], axis = 1, inplace = True) # Drop

# Put full date in Image.Acquisition.Date column
for row_num in range(len(ppmi_merge['T1.s3.Image.Name'])) :
    if isinstance(ppmi_merge['T1.s3.Image.Name'].iloc[row_num], str) :
        date = ppmi_merge['T1.s3.Image.Name'].iloc[row_num].split('-')[2]
        ppmi_merge['Image.Acquisition.Date'].iloc[row_num] = date[4:6] + '/' + date[6:8] + '/' + date[0:3]

ppmi_merge.to_csv('/Users/areardon/Desktop/ppmi_merge916.csv')
## Add in 'Analytic.Cohort' column
analytic_cohort_subids = analytic_cohort_subids.tolist()
ppmi_merge['Analytic.Cohort'] = ppmi_merge['Subject.ID'].map(lambda x : 'Analytic Cohort' if x in analytic_cohort_subids else 'Not Analytic Cohort')
ppmi_merge.to_csv('/Users/areardon/Desktop/ppmi_merge10.csv')

## Get Enrollment Diagnosis for subjects in Not Analytic Cohort - do this using the participants_status
analytic = ppmi_merge[ppmi_merge['Analytic.Cohort'] == 'Analytic Cohort'] # Split up Analytic cohort df
not_analytic = ppmi_merge[ppmi_merge['Analytic.Cohort'] == 'Not Analytic Cohort'] # Split up Analytic cohort df
participant_status = pd.read_csv(ppmi_download_path + 'Participant_Status.csv') # Read in participant_status
participant_status = participant_status[['PATNO', 'COHORT_DEFINITION']] # Keep only
participant_status.rename(columns = {'PATNO' : 'Subject.ID', 'COHORT_DEFINITION' : 'Enroll.Diagnosis'}, inplace = True)
not_analytic_participant_status = pd.merge(not_analytic, participant_status, on = ['Subject.ID'], how = "left")
not_analytic_participant_status.drop(['Enroll.Diagnosis_x'], axis = 1, inplace = True) # Remove the extra column
not_analytic_participant_status.rename(columns = {'Enroll.Diagnosis_y' : 'Enroll.Diagnosis'}, inplace = True)

```



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not_analytic_participant_status.loc[not_analytic_participant_status['Enroll.Diagnosis'] == 'Healthy Control'] = 'Healthy Control'

# Merge df of Not Analytic and Analytic subjects and sort by SubID and Event.ID.Date
ppmi_merge = pd.concat([analytic, not_analytic_participant_status])

# Change Event.ID.Date to date time and corrected format
ppmi_merge['Event.ID.Date'] = ppmi_merge['Event.ID.Date'].astype(str)
ppmi_merge['Event.ID.Date'] = pd.to_datetime(ppmi_merge['Event.ID.Date'], errors = "ignore") # Change event ID date to datetime
ppmi_merge = ppmi_merge.sort_values(by = ['Subject.ID', 'Event.ID.Date']) # Sort values by subject and event ID date
ppmi_merge['Event.ID.Date'] = ppmi_merge['Event.ID.Date'].astype(str) # Change Event.ID.Date back to string

## HERE - lose 3101

# Reformat Event.ID.Date from pd.to_datetime to month/year
for row_num in range(len(ppmi_merge['Event.ID.Date'])):
    if ppmi_merge['Event.ID.Date'].iloc[row_num] != 'NaT':
        split = ppmi_merge['Event.ID.Date'].iloc[row_num].split('-')
        new_date = split[1] + '/' + split[0] # month/year format
        ppmi_merge['Event.ID.Date'].iloc[row_num] = new_date
ppmi_merge['Event.ID.Date'] = ppmi_merge['Event.ID.Date'].replace('NaT', 'NA')

#### Add in ppmi_qc_BA.csv - Brian sent on slack 1/31/22 ####
ppmi_qc_BA = pd.read_csv(invicro_data_path + 'ppmi_qc_BA.csv')
ppmi_qc_BA = ppmi_qc_BA.reset_index(drop = False) # Move index to first column so we can rename
ppmi_qc_BA.rename(columns = {'ID' : 'ImageID'}, inplace = True) # Rename ImageID

# Update ImageID column bc info from T1 file (where ImageID was created from) does not contain all the info
for row_num in range(len(ppmi_merge['ImageID'])) :
    if isinstance(ppmi_merge['T1.s3.Image.Name'].iloc[row_num], str):
        imageID = ppmi_merge['T1.s3.Image.Name'].iloc[row_num].split('.')[0] # take info before .nii.gz
        ppmi_merge['ImageID'].iloc[row_num] = imageID.split('-')[1] + '-' + imageID.split('-')[2] + '-' + imageID.split('-')[3]
ppmi_merge = pd.merge(ppmi_merge, ppmi_qc_BA, on = ['ImageID'], how = "left") # Merge - keep only from ppmi_merge

#### BILATERAL SUBTYPE SCORES (Tremor and Brady) ####
brady_right3 = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num']
brady_right3a = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num']
brady_right3ON = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num']
brady_right3OF = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num']

brady_left3 = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num']
brady_left3a = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num']
brady_left3ON = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num']
brady_left3OF = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num']

brady_sym3 = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num']
brady_sym3a = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num']
brady_sym3ON = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num']
brady_sym3OF = ['Dominant.Side.Disease', 'Rigidity.Neck.UPDR30F.Num', 'Rigidity.RUE.UPDR30F.Num', 'Rigidity.LUE.UPDR30F.Num', 'Rigidity.Neck.UPDRS3A.Num', 'Rigidity.RUE.UPDRS3A.Num', 'Rigidity.LUE.UPDRS3A.Num', 'Rigidity.Neck.UPDRS3.Num', 'Rigidity.RUE.UPDRS3.Num', 'Rigidity.LUE.UPDRS3.Num', 'Rigidity.Neck.UPDR30N.Num', 'Rigidity.RUE.UPDR30N.Num', 'Rigidity.LUE.UPDR30N.Num']

# # # Make any dominant side of disease that are NA into 'Symmetric'
# # domsideisna = ppmi_merge['Dominant.Side.Disease'].isna()
# # ppmi_merge['Dominant.Side.Disease'].loc[domsideisna] = 'Symmetric'

```

```

def add_lateralized_subscores(df : pd.DataFrame, subscore_side_list : list, side : str, new_col_name : str) :
    """
    Include lateralized subscores (i.e. Brady Rigidity and Tremor subscores) into dataframe.

    Arguments
    -----
    df : pd.DataFrame containing scores that make up subscore

    subscore_side_list : list containing column names of the scores that make up the subscore

    side : 'Left' or 'Right'

    new_col_name : name of new column name with lateralized subscore

    """
    subscore_side = df[subscore_side_list] # Get dataframe of only columns in brady_left
    df[new_col_name] = 0 # Initialize lateralized variable
    subscore_side.loc[subscore_side['Dominant.Side.Disease'] != side, :] = np.nan # Make all rows nan if dom side != side
    subscore_side_temp = subscore_side.drop('Dominant.Side.Disease',1) # Drop dominant side of disease
    idx = subscore_side_temp.loc[pd.isnull(subscore_side_temp).any(1), :].index
    df[new_col_name] = subscore_side_temp.sum(axis = 1) # Sum of all columns in each row where dom side == side
    df[new_col_name].iloc[idx] = np.nan # Fill in subscores that should be nans as nan

    return df

def add_symmetric_subscores(df : pd.DataFrame, subscore_side_list : list, side : str, new_col_name : str) :
    subscore_side = df[subscore_side_list] # Get dataframe of only columns in brady_left
    df[new_col_name] = 0 # Initialize lateralized variable
    subscore_side.loc[subscore_side['Dominant.Side.Disease'] != side, :] = np.nan # Make all rows nan if dom side != side
    subscore_side_temp = subscore_side.drop('Dominant.Side.Disease',1) # Drop dominant side of disease
    idx = subscore_side_temp.loc[pd.isnull(subscore_side_temp).any(1), :].index
    x = subscore_side_temp.fillna(0) # because adding 1 plus nan equals nan
    if "Brady" in new_col_name :
        df[new_col_name] = x['Rigidity.Neck' + side] + (x['Rigidity.RUE' + side] + x['Rigidity.LUE' + side])
        df[new_col_name].iloc[idx] = np.nan # Fill in subscores that should be nans as nan
    elif "Tremor" in new_col_name :
        df[new_col_name] = x['Tremor.UPDRS2.Num'] + (x['Postural.Tremor.Right.Hand' + side] + x['Postural.Tremor.Left.Hand' + side])
        df[new_col_name].iloc[idx] = np.nan
    return df

ppmi_merge = add_lateralized_subscores(ppmi_merge, brady_left3, 'Left', 'Brady.Rigidity.Subscore-left.UPDRS3')
ppmi_merge = add_lateralized_subscores(ppmi_merge, brady_right3, 'Right', 'Brady.Rigidity.Subscore-right.UPDRS3')
ppmi_merge = add_symmetric_subscores(ppmi_merge, brady_sym3, 'Symmetric', 'Brady.Rigidity.Subscore-sym.UPDRS3')

# Combine left and right and sym subscores into same column
ppmi_merge["Brady.Rigidity.Subscore.lateralized.UPDRS3"] = ppmi_merge.pop("Brady.Rigidity.Subscore-right.UPDRS3")
ppmi_merge["Brady.Rigidity.Subscore.lateralized.UPDRS3"] = ppmi_merge.pop("Brady.Rigidity.Subscore.left.UPDRS3")

#### TREMOR ####
tremor_right3 = ['Dominant.Side.Disease', 'Tremor.UPDRS2.Num', 'Postural.Tremor.Right.Hand.UPDRS3.Num', 'Postural.Tremor.Left.Hand.UPDRS3.Num']
tremor_left3 = ['Dominant.Side.Disease', 'Tremor.UPDRS2.Num', 'Postural.Tremor.Left.Hand.UPDRS3.Num', 'Postural.Tremor.Right.Hand.UPDRS3.Num']
tremor_sym3 = ['Dominant.Side.Disease', 'Tremor.UPDRS2.Num', 'Postural.Tremor.Right.Hand.UPDRS3.Num', 'Postural.Tremor.Left.Hand.UPDRS3.Num']

```

```

ppmi_merge = add_lateralized_subscores(ppmi_merge, tremor_left3, 'Left', 'Tremor.Subscore-left.UPDRS3')
ppmi_merge = add_lateralized_subscores(ppmi_merge, tremor_right3, 'Right', 'Tremor.Subscore-right.UPDRS3')
ppmi_merge = add_symmetric_subscores(ppmi_merge, tremor_sym3, 'Symmetric', 'Tremor.Subscore-sym.UPDRS3')

# Combine left and right and sym scores into same column (.lateralized)
ppmi_merge["Tremor.Subscore.lateralized.UPDRS3"] = ppmi_merge.pop("Tremor.Subscore-right.UPDRS3").fillna(0)
ppmi_merge["Tremor.Subscore.lateralized.UPDRS3"] = ppmi_merge.pop("Tremor.Subscore-left.UPDRS3").fillna(0)

## If lateralized subscore is nan - input the non-lateralized score for Tremor and Brady.Rigidity
def fill_non_lat_subscore(df, subscore_lateralized_name, subscore_name):
    latisna = df[subscore_lateralized_name].isna() & df[subscore_name].notna()
    df[subscore_lateralized_name].loc[latisna] = df[subscore_name].loc[latisna]
    return df

ppmi_merge = fill_non_lat_subscore(ppmi_merge, 'Tremor.Subscore.lateralized.UPDRS3', 'Tremor.Subscore.UPDRS3')
ppmi_merge = fill_non_lat_subscore(ppmi_merge, 'Brady.Rigidity.Subscore.lateralized.UPDRS3', 'Brady.Rigidity.Subscore.UPDRS3')

## Include columns for bestEventID (bestScreening, bestBaseline, etc) and denote the highest resnetGrade
myevs = ppmi_merge['Event.ID'].unique() # Unique event ids
uids = ppmi_merge['Subject.ID'].unique() # Unique subject ids
for myev in myevs:
    mybe = "best" + myev # Create best Visit column
    ppmi_merge[mybe] = False # Set all best visit to be False
    for u in uids:
        selu = ppmi_merge.loc[(ppmi_merge['Subject.ID'] == u) & (ppmi_merge['Event.ID'] == myev) & (ppmi_merge['resnetGrade'] > 0)]
        if len(selu) == 1: # If there is one event id for that subject
            idx = selu.index # Get the index
            ppmi_merge.loc[idx, mybe] = True
        if len(selu) > 1: # IF there is more than one event id for that subject and resnet grade is not 0
            maxidx = selu[['resnetGrade']].idxmax() # Get the higher resnetGrade for each visit if there is more than one
            ppmi_merge.loc[maxidx, mybe] = True

## Include a column for bestAtImage.Acquisition.Date - denote the one or highest resnetGrade with True
ppmi_merge['bestAtImage.Acquisition.Date'] = False # Initialize bestAtImage.Acquisition.Date col
for myev in myevs:
    for u in uids:
        selu = ppmi_merge.loc[(ppmi_merge['Subject.ID'] == u) & (ppmi_merge['Event.ID'] == myev) & (ppmi_merge['resnetGrade'] > 0)]
        if len(selu) == 1: # If there is one event id for that subject
            idx = selu.index # Get the index
            ppmi_merge.loc[idx, 'bestAtImage.Acquisition.Date'] = True
        if len(selu) > 1: # IF there is more than one event id for that subject and resnet grade is not 0
            maxidx = selu[['resnetGrade']].idxmax() # Get the higher resnetGrade for each visit if there is more than one
            ppmi_merge.loc[maxidx, 'bestAtImage.Acquisition.Date'] = True

# DX simplified column
ppmi_merge['DXsimplified'] = '' # Initialize DXsimplified
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == 'Healthy Control', 'DXsimplified'] = 'HC'
ppmi_merge.loc[ppmi_merge['Enroll.Diagnosis'] == 'Healthy Control', 'DXsimplified'] = 'HC'
ppmi_merge.loc[ppmi_merge['Enroll.Diagnosis'] == 'Parkinson\'s Disease', 'DXsimplified'] = 'Sporadic_PD'
ppmi_merge.loc[ppmi_merge['Enroll.Diagnosis'] == 'SWEDD', 'DXsimplified'] = "nonPDorMSA"
ppmi_merge.loc[ppmi_merge['Enroll.Diagnosis'] == "Prodromal", 'DXsimplified'] = "Sporadic_Pro"

```

```

ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "GBA", 'DXsimplified'] = "GBA_HC"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : GBA", 'DXsimplified'] = "GBA_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : GBA not Prodromal", 'DXsimplified'] = "GBA_Pro"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : GBA Prodromal", 'DXsimplified'] = "GBA_Pro"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : LRRK2", 'DXsimplified'] = "LRRK2_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : LRRK2 + GBA", 'DXsimplified'] = "LRRK2_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : LRRK2 + GBA not Prodromal", 'DXsimplified'] = "LRRK2_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : LRRK2 + GBA Prodromal", 'DXsimplified'] = "LRRK2_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : LRRK2 not Prodromal", 'DXsimplified'] = "LRRK2_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : LRRK2 Phenoconverted", 'DXsimplified'] = "LRRK2_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : LRRK2 Prodromal", 'DXsimplified'] = "LRRK2_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : SNCA", 'DXsimplified'] = "SNCA_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Genetic : SNCA Prodromal", 'DXsimplified'] = "SNCA_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Hyposmia", 'DXsimplified'] = "Sporadic_Pro"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Hyposmia : Phenoconverted", 'DXsimplified'] = "Sporadic_Pro"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "No Mutation not Prodromal", 'DXsimplified'] = np.NaN
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "non-HC", 'DXsimplified'] = np.NaN
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "non-PD", 'DXsimplified'] = "nonPDorMSA"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "RBD", 'DXsimplified'] = "Sporadic_Pro"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "RBD : Phenoconverted", 'DXsimplified'] = "Sporadic_Pro"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "RBD : Phenoconverted with GBA", 'DXsimplified'] = "Sporadic_Pro"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "Sporadic", 'DXsimplified'] = "Sporadic_PD"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "SWEDD/non-PD Active", 'DXsimplified'] = "nonPDorMSA"
ppmi_merge.loc[ppmi_merge['Consensus.Subtype'] == "SWEDD/PD Active", 'DXsimplified'] = np.NaN

```

Add in min PD Disease duration

Create a simplified diagnosis group of just HC, PD, Prodromal, or nonPDorMSA

```

ppmi_merge['DXs2'] = ''
ppmi_merge['DXsimplified'].fillna('NA', inplace = True)
ppmi_merge.loc[ppmi_merge['DXsimplified'].str.contains('HC'), 'DXs2'] = 'HC'
ppmi_merge.loc[ppmi_merge['DXsimplified'].str.contains('_PD'), 'DXs2'] = 'PD'
ppmi_merge.loc[ppmi_merge['DXsimplified'].str.contains('_Pro'), 'DXs2'] = 'Pro'
ppmi_merge.loc[ppmi_merge['DXsimplified'].str.contains('nonPDorMSA'), 'DXs2'] = 'nonPDorMSA'

```

If HC, Prodromal or nonPDorMSA - fill in with 0 for PD.Min.Disease.Duration

```

ppmi_merge.loc[ppmi_merge['DXs2'] == 'HC', 'PD.Min.Disease.Duration'] = 0
ppmi_merge.loc[ppmi_merge['DXs2'] == 'Pro', 'PD.Min.Disease.Duration'] = 0
ppmi_merge.loc[ppmi_merge['DXs2'] == 'nonPDorMSA', 'PD.Min.Disease.Duration'] = 0

```

For PD subjects, fill in PD.Min.Disease.Duration

```

ppmi_merge['PD.Diagnosis.Duration'].fillna('', inplace = True)

```

subids = ppmi_merge['Subject.ID'].unique() # Unique subject ids

for subid in subids :

```

    df = ppmi_merge[(ppmi_merge['Subject.ID'] == subid) & (ppmi_merge['DXs2'] == 'PD') & (ppmi_merge['PD.Min.Disease.Duration'] > 0)]

```

if len(df) > 0 :

```

    print(df['PD.Diagnosis.Duration'])

```

```

    mydd = min(df['PD.Diagnosis.Duration'])

```

```

    ppmi_merge.loc[ppmi_merge['Subject.ID'] == subid, 'PD.Min.Disease.Duration'] = mydd

```

Add in Visit column

```

ppmi_merge['Visit'] = np.NaN # Initialize Visit col
searchfor = ['Baseline', 'Visit Month '] # Strings to search for
temp = ppmi_merge['Event.ID'].str.contains('|'.join(searchfor)) # locations of where row contains str:
ppmi_merge['Visit'].loc[temp == True] = ppmi_merge['Event.ID'].loc[temp == True] # Fill in 'Visit' col
ppmi_merge['Visit'] = ppmi_merge['Visit'].str.replace("Remote Visit Month ", "") # Replace Remote visit
ppmi_merge['Visit'] = ppmi_merge['Visit'].str.replace("Visit Month ", "") # Replace Visit month with ''
ppmi_merge['Visit'] = ppmi_merge['Visit'].str.replace("Baseline", "0") # Replace baseline with 0
ppmi_merge['Visit'] = ppmi_merge['Visit'].fillna(9999) # Filling with 9999 so we can change this col to
ppmi_merge['Visit'] = ppmi_merge['Visit'].astype(int) # str to int
ppmi_merge['Visit'] = ppmi_merge['Visit'].replace(9999, np.nan) # Replace 9999 with nan

# Final re-organization of ppmi_merge and save
ppmi_merge.set_index('Subject.ID', inplace = True)
ppmi_merge.fillna('NA', inplace = True)
cols_delete = ['Unnamed'] # Columns to remove from sheet/df
ppmi_merge = ppmi_merge.loc[:, ~ppmi_merge.columns.str.startswith(tuple(cols_delete))] # Remove columns
ppmi_merge.to_csv(userdir + 'ppmi_merge_v' + version + '.csv')

```