#### Mudcard

- loc and iloc :)
  - we will go through this again today
- should we just use pandas in this class or would using both/polars benefit us more in and out of the classroom?
  - I'd say learn both, that's the most benefitial for you out of classroom.
  - As far as problem sets and quizzes are concerned, it's fine to use one of the packages.
  - I'll mostly use pandas going forward.
- What are some ways to get more familiar with working with python?
  - easy leetcode problems are a good start to learn standard python (no packages)
  - you will practice how to work with popular data science packages well enough during class and the problem sets I think
- need to review basic python functions, like skiprows
  - skiprows is an argument of the pd.read\_csv function :)

### Lecture 3

#### Working with data (step 0) continued

#### Learning objectives

By the end of the lecture, you will be able to

- filter columns
- merge and append data frames
- modify a dataframe

#### **Pandas and Polars**

- data are often distributed over multiple files/databases (e.g., csv and excel files, sql databases)
- each file/database is read into a pandas dataframe last lecture
- you often need to filter dataframes by selecting specific rows last lecture
- you often need to filter dataframes by selecting specific columns today
- multiple dataframes need to be merged and appended today
- dataframes sometimes need to be modified today

#### Some notes and advice

- ALWAYS READ THE HELP OF THE METHODS/FUNCTIONS YOU USE!
- stackoverflow is your friend, use it! https://stackoverflow.com/
- you can also use generative AI to help you fix bugs
  - Gemini is supported by Brown OIT (see here https://go.brown.edu/gemini)
- here is an excellet review of the syntax differences between pandas and polars

#### Learning objectives

By the end of the lecture, you will be able to

- filter columns
- merge and append data frames
- modify a dataframe

```
In [1]: import pandas as pd
        df_pd = pd.read_csv('../data/adult_data.csv')
        columns = df pd.columns
        print(columns)
        # select columns by column name
        #print(df_pd[['age','hours-per-week']])
        #print(columns[[1,5,7]])
        #print(df_pd[columns[[1,5,7]]])
        # select columns by index using iloc
        #print(df_pd.iloc[:,3])
        # select columns by index - not standard python indexing
        #print(df_pd.iloc[:,[3,5,6]])
        # select columns by index - standard python indexing
        #print(df_pd.iloc[:,::2])
       Index(['age', 'workclass', 'fnlwgt', 'education', 'education-num',
              'marital-status', 'occupation', 'relationship', 'race', 'sex',
              'capital-gain', 'capital-loss', 'hours-per-week', 'native-country',
              'gross-income'],
             dtype='object')
In [2]: import polars as pl
        df_pl = pl.read_csv('../data/adult_data.csv')
        columns = df pl.columns
        print(columns)
        # select columns by column name
        #print(df_pl['age','hours-per-week'])
```

```
#print(df_pl.select(['age','hours-per-week'])) # use .select if you know the
#print(columns[1:4]) # indices must be integers or slices
#print(df_pl[columns[1:4]])

# select columns by index, polars has no .iloc
#print(df_pl[:,3])

# select columns by index - not standard python indexing but it works
#print(df_pl[:,[3,5,6]])

# select columns by index - standard python indexing
#print(df_pl[:,::2])
```

['age', 'workclass', 'fnlwgt', 'education', 'education-num', 'marital-statu s', 'occupation', 'relationship', 'race', 'sex', 'capital-gain', 'capital-lo ss', 'hours-per-week', 'native-country', 'gross-income']

#### Learning objectives

By the end of the lecture, you will be able to

- filter columns
- merge and append data frames
- modify a dataframe

#### How to merge dataframes in Pandas?

Merge - info on data points are distributed in multiple files

```
In [3]: # We have two datasets from two hospitals
        hospital1 = {'ID':['ID1','ID2','ID3','ID4','ID5','ID6','ID7'],'col1':[5,8,2,
        df1 = pd.DataFrame(data=hospital1)
        print(df1)
        hospital2 = {'ID':['ID2','ID5','ID6','ID10','ID11'],'col3':[12,76,34,98,65],
        df2 = pd.DataFrame(data=hospital2)
        print(df2)
           ID col1 col2
       0 ID1
                  5
                       У
       1 TD2
                  8
                       j
       2 ID3
                  2
       3 ID4
                  6
                       b
       4 ID5
                  0
                       а
       5 ID6
                  2
                       h
       6 ID7
                  5
           ID col3 col2
       0
          ID2
                  12
       1
          ID5
                  76
       2
          ID6
                  34
                        е
       3 TD10
                  98
                        l
```

р

4 ID11

65

```
In [4]: # we are interested in only patients from hospital1
    df_left = df1.merge(df2,how='left',on='ID') # IDs from the left dataframe (c
    #print(df_left)

# we are interested in only patients from hospital2
# #df_right = df1.merge(df2,how='right',on='ID') # IDs from the right datafram
# #df_right = df2.merge(df1,how='left',on='ID')
# print(df_right)

# we are interested in patiens who were in both hospitals
# #df_inner = df1.merge(df2,how='inner',on='ID') # merging on IDs present in the
# #print(df_inner)

# we are interested in all patients who visited at least one of the hospital
# #df_outer = df1.merge(df2,how='outer',on='ID') # merging on IDs present in
# #print(df_outer)
```

#### How to append dataframes in pandas?

Append - new data comes in over a period of time. E.g., one file per month/quarter/fiscal year etc.

You want to combine these files into one data frame.

```
In [5]: df_append = pd.concat([df1,df2]) # note that rows with ID2, ID5, and ID6 ar
    print(df_append)

#df_append = pd.concat([df1,df2],ignore_index=True) # note that rows with ID
    #print(df_append)

# d3 = {'ID':['ID23','ID94','ID56','ID17'],'col1':['rt','h','st','ne'],'col2
    # df3 = pd.DataFrame(data=d3)
    # print(df3)

# df_append = pd.concat([df1,df2,df3],ignore_index=True) # multiple datafram
    # print(df_append)

ID col1 col2 col3
```

```
0
   TD1
         5.0
                   NaN
               У
1
   ID2
         8.0
                   NaN
               i
2
   ID3
         2.0
                   NaN
                   NaN
3
   ID4
         6.0
               b
   ID5
         0.0
4
                   NaN
5
   ID6
         2.0
                   NaN
6
   ID7
         5.0
                   NaN
              q 12.0
   ID2
         NaN
1
         NaN
             u 76.0
   ID5
2
   ID6
         NaN
               e 34.0
3 ID10
         NaN
               1 98.0
4 TD11
         NaN
               p 65.0
```

#### How to merge/join dataframes in Polars?

```
In [6]: hospital1 = {'ID':['ID1','ID2','ID3','ID4','ID5','ID6','ID7'],'col1':[5,8,2,
        df1 = pl.DataFrame(data=hospital1)
        #print(df1)
        hospital2 = {'ID':['ID2','ID5','ID6','ID10','ID11'],'col3':[12,76,34,98,65],
        df2 = pl.DataFrame(data=hospital2)
        #print(df2)
        # left join
        #df_left = df1.join(df2,how='left',on='ID') # IDs from the left dataframe (c
        #print(df left)
        # riaht ioin
        #df_right = dfl.join(df2,how='right',on='ID') # IDs from the right dataframe
        #df_right = df2.join(df1,how='left',on='ID')
        #print(df right)
        # inner join
        #df inner = dfl.join(df2,how='inner',on='ID') # merging on IDs present in bo
        #print(df inner)
        # outer join is called a full join
        #df_outer = df1.join(df2,how='full',on='ID')  # merging on IDs present in ar
        #print(df_outer)
```

#### Quiz

```
In [7]: raw_data_1 = {
                'subject_id': ['1', '2', '3', '4', '5'],
                'first_name': ['Alex', 'Amy', 'Allen', 'Alice', 'Ayoung'],
                'last_name': ['Anderson', 'Ackerman', 'Ali', 'Aoni', 'Atiches']}
        raw_data_2 = {
                'subject_id': ['6', '7', '8', '9', '10'],
                'first_name': ['Billy', 'Brian', 'Bran', 'Bryce', 'Betty'],
                'last name': ['Bonder', 'Black', 'Balwner', 'Brice', 'Btisan']}
        raw data 3 = {
                'subject_id': ['1', '2', '3', '4', '5', '7', '8', '9', '10', '11'],
                'test_id': [51, 15, 15, 61, 16, 14, 15, 1, 61, 16]}
        # Create three data frames from raw data 1, 2, and 3.
        # Append the first two data frames and assign it to df append.
        # Merge the third data frame with df_append such that only subject_ids from
        # Assign the new data frame to df merge.
        # How many rows and columns do we have in df_merge?
```

# Always check that the resulting dataframe is what you wanted to end up with!

small toy datasets are ideal to test your code.

## If you need to do a more complicated dataframe operations, check out pd.concat() and pl.concat()!

#### Learning objectives

By the end of the lecture, you will be able to

- filter columns
- merge and append data frames
- · modify a dataframe

### Do not EVER overwrite the original data files!

#### Always save the modified dataframe to a new file!

#### Why do we need to modify the dataset?

- feature engineering
  - generating new features (adding new informative columns) can improve the performance of the ML model
- fix dataset issues
  - there might be data quality issues that need to be manually fixed
  - typos, missing values, etc

```
import pandas as pd
df_pd = pd.read_csv('../data/adult_data.csv')
print(df_pd.head())

# let's generate a new feature
# is immigrant? False (0) if the person's home country is the USA, True (1)
# such a feature has only two categories but it might be pretty informative
df_pd['is immigrant'] = df_pd['native-country'] != ' United-States'

#print(df_pd.head())
```

	age		workclass	†n Lw	gt	educati	.on	education-nur	n \	
0	39	State-gov		77516 Bac		Bachelo	rs	13	3	
1	50	Self-emp-not-inc		83311		Bachelors		13	3	
2	38	Private		215646 HS		HS-gr	ad	Ć	9	
3	53	Private		234721 1		11	th	<del>-</del>	7	
4	28	8 Private		33840	09 Bachelo		rs	13	3	
`	marital-status			occupation				relationship	race	sex
١		Novor m	arriad	Λ.	dm c	lorical		lot in family	White	Mala
0	Never-married			Adm-clerical			11	Not-in-family	White	Male
1	Married-civ-spouse			Exec-managerial				Husband	White	Male
2	Divorced H			Handlers-cleaners			Ν	Not-in-family	White	Male
3			Handlers-cleaners				Husband	Black	Male	
4	Married-civ-spouse		Prof-specialty				Wife	Black	Female	
4	riai i	ieu-civ-	spouse	710	ı – 5 µ	естасту		WITE	Diack	i ellia ce
	capit	tal-gain	capital-l	oss l	hour	s-per-we	ek	native-count	ry gross	-income
0		2174		0			40	United-State	es	<=50K
1		0		0		13	United-States		<=50K	
2	0		0			40	United-States		<=50K	
3	0		0			40	United-State	es	<=50K	
4	0		0			40	Cul	oa	<=50K	

## Mud card

In [ ]: