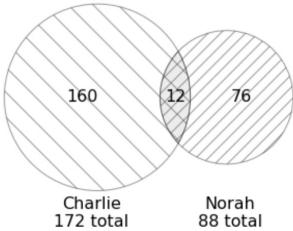


As illustrated in the diagram below, Charlie has 172 contacts in total, whereas Norah has 88 contacts. 12 of these contacts are shared, meaning they appear in both `charlie` and `norah`.

Disc 4, 8.2

Venn Diagram of Charlie and Norah's Contacts



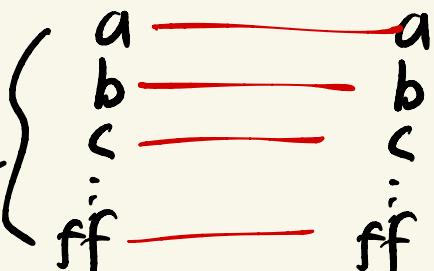
Problem 8.2

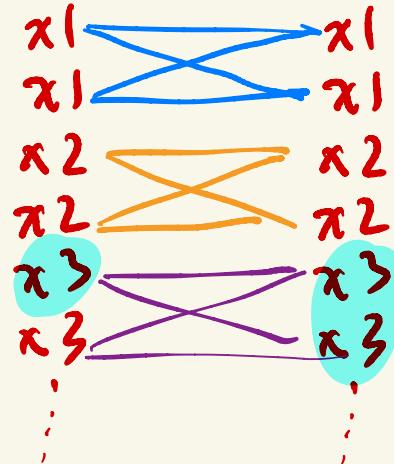
One day, when updating her phone's operating system, Norah accidentally duplicates the 12 contacts she has in common with Charlie. Now, the `norah` DataFrame has 100 rows.

What does the following expression evaluate to?

```
norah.merge(norah, left_index=True, right_index=True).shape[0]
```

76 unique {
 d
 c
 :
 z
 y

76 unique {  } $(1 \cdot 1) 76 = 76$



$4 \ x 1s$
 $4 \ x 2s$
 $4 \ x 3s$
 \vdots

$$\begin{aligned}
 &= 4 \cdot 12 \\
 &= 2 \cdot 24 \\
 &= 48
 \end{aligned}$$

$4 \ x^{12s}$

Discussion 5, 4-2

⇒ What does .strip() do?

↑
removes from beginning and end of string

" hi ".strip() → "hi"
'''

hi .strip() → "+74++".strip["+"]
→ "74"

""

Discussion 5, 3.5

`r = requests.get(url)`

$\Rightarrow r$ is a "Response" object

$\Rightarrow r.text$ is the content

\Rightarrow always a string $\xrightarrow{\text{BeautifulSoup}(r.text)}$

\Rightarrow could look like HTML \rightarrow use BeautifulSoup

\Rightarrow could look like JSON

$\Rightarrow r.json()$

\nearrow
Python dictionary

Monday Review, Problem 2

Answer: Option 1, Option 3

In this problem, we will be using the following DataFrame `students`, which contains various information about high school student university/universities they applied to.

| | Name | High School | Email | GPA | APs | University | Admit |
|---|-------------|------------------|---------------------|------|-----|------------------|-------|
| 0 | Billy King | La Jolla Private | billy@lprivate.hig | 3.92 | 8 | UC San Diego | Y |
| 1 | Billy King | La Jolla Private | billy@lprivate.hig | 3.92 | 8 | Stanford | Y |
| 2 | Sally Singh | Warren High | sally@warren.hs.edu | 4.05 | 14 | UC San Diego | Y |
| 3 | Sally Singh | Warren High | sally@warren.hs.edu | 4.05 | 14 | Columbia | N |
| 4 | Johnny Vu | La Jolla Private | johnny@lprivate.hig | 3.45 | 6 | UC San Diego | W |
| 5 | Johnny Vu | La Jolla Private | johnny@lprivate.hig | 3.45 | 6 | UC Santa Barbara | Y |
| 6 | Johnny Vu | La Jolla Private | johnny@lprivate.hig | 3.45 | 6 | UC Irvine | Y |

Which of the following blocks of code correctly assign `max_AP` to the maximum number of APs taken by a student who was rejected by UC San Diego?

Option 1:

```
cond1 = students["Admit"] == "N"
cond2 = students["University"] == "UC San Diego"
max_AP = students.loc[cond1 & cond2, "APs"].sort_values().iloc[-1]
```

ascending default

{ biggest value }

d3:

Option 2:

rejected by UCSD

```
cond1 = students["Admit"] == "N"
cond2 = students["University"] == "UC San Diego"
d3 = students.groupby(["University", "Admit"]).max().reset_index()
max_AP = d3.loc[cond1 & cond2, "APs"].iloc[0]
```

{ biggest value }

Option 3:

revert: same length as students: invalid indexer

```
p = students.pivot_table(index="Admit",
                          columns="University",
                          values="APs",
                          aggfunc="max")
max_AP = p.loc["N", "UC San Diego"]
```

| | Michigan | UCSD | Harvard | --- |
|---|----------|------|---------|-----|
| Y | | | | |
| N | | ✓ | | |

Option 4:

```
# .last() returns the element at the end. A Series it is called on
groups = students.sort_values(["APs", "Admit"]).groupby("University")
max_AP = groups["APs"].last()["UC San Diego"]
```

sort by APs, break ties by Admit
"N" before "Y"

groups: UCSD GPA AP Admit ---
Michigan
Harvard

| University | Admit | GPA | APs | Name | High School |
|------------|-------|-----|-----|------|-------------|
| Michigan | Y | | | | |
| Michigan | N | | | | |
| UCSD | Y | | | | |
| UCSD | N | | | | |
| UCSD | | | | | |



```
(  
    students  
    - sort_values(...)  
    - groupby("University")  
        ["APs"]  
        .last()  
    - loc["UCSD"]  
)
```

] DataFrameGroupBy object

] Series
 DataFrame] .iloc[-1]

df.sort_values("AP").groupby("University").last() ←
df.sort_values("AP").groupby("University").apply(lambda f:
 f.iloc[-1])

equivalent

df

| color | height | weight | |
|-------|--------|--------|--|
| red | 2 | 4 | |
| blue | 1 | 9 | |
| blue | 4 | 2 | |
| red | 7 | 3 | |
| green | 3 | 1 | |
| blue | 9 | -1 | |
| green | 1 | 0 | |

`df.groupby("color").last()`

usually, sort first

Monday Review, 2.3

```
students.groupby("Email").aggregate({"Name": "max",  
                                     "High School": "mean",  
                                     "GPA": "mean",  
                                     "APs": "max"})
```

can't
average strings!!!

$f: S \rightarrow "VMSS"$

$$S = \begin{bmatrix} VMSS \\ VMSS \\ \vdots \\ VMSS \end{bmatrix}$$

possible answers:

"min"

"max"

"first"

"last"

lambda hi shi. iloc[0]

Problem 2

In this problem, we will be using the following DataFrame `students`, which contains various information about high school students and the university/universities they applied to.

| | Name | High School | Email | GPA | APs | University | Admit |
|---|----------------|--------------------|------------------------|------|-----|------------------|-------|
| 0 | Billy King | La Jolla Private | billy@ljjprivate.high | 3.92 | 8 | UC San Diego | Y |
| 1 | Billy King | La Jolla Private | billy@ljjprivate.high | 3.92 | 8 | Stanford | Y |
| 2 | Sally Singh | Warren High | sally@warren.hs.edu | 4.05 | 14 | UC San Diego | Y |
| 3 | Sally Singh | Warren High | sally@warren.hs.edu | 4.05 | 14 | Columbia | N |
| 4 | Johnny Vu | La Jolla Private | johnny@ljjprivate.high | 3.45 | 6 | UC San Diego | W |
| 5 | Johnny Vu | La Jolla Private | johnny@ljjprivate.high | 3.45 | 6 | UC Santa Barbara | Y |
| 6 | Johnny Vu | La Jolla Private | johnny@ljjprivate.high | 3.45 | 6 | UC Irvine | Y |
| 7 | Cassie Charles | Triton Magnet High | cassie@triton.high | 3.84 | 9 | UC San Diego | N |

“17” → “17” < “9”

“9”

“17” vs. “EECS” : idk, but check
ASCII (o)

agg vs filter vs transform vs apply

see posted notebook and video
under week 7!