

LECTURE 4

Pandas, Part III

Advanced Pandas (More on Grouping, Aggregation, Pivot Tables Merging)

Data 100/Data 200, Spring 2023 @ UC Berkeley

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Content credit: Lisa Yan, Josh Hug





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① Start presenting to display the joining instructions on this slide.



New Syntax / Concept Summary



Today we'll cover:

- Groupby: Output of .groupby("Name") is a DataFrameGroupBy object. Condense back into a DataFrame or Series with:
 - groupby.size
 - o groupby.filter
 - and more...
- Pivot tables: An alternate way to group by exactly two columns.
- Joining tables using pd.merge.
- Exploratory data analysis





Groupby Review

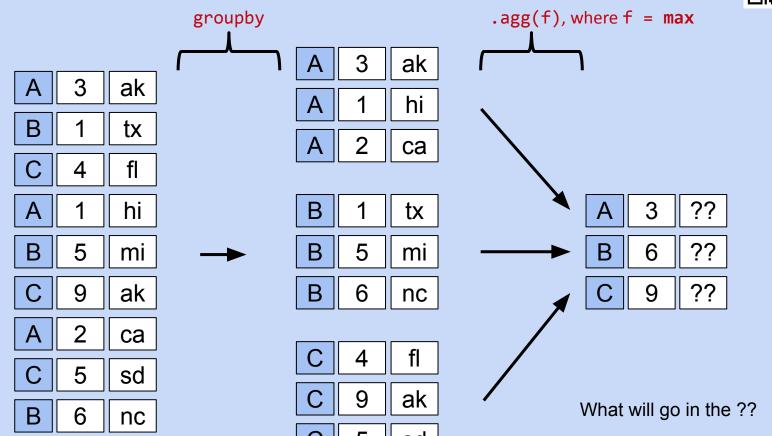
Lecture 04, Data 100 Spring 2023

- Groupby Review
- Other DataFrameGroupBy Features
- Groupby and PivotTables
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Groupby Review Question









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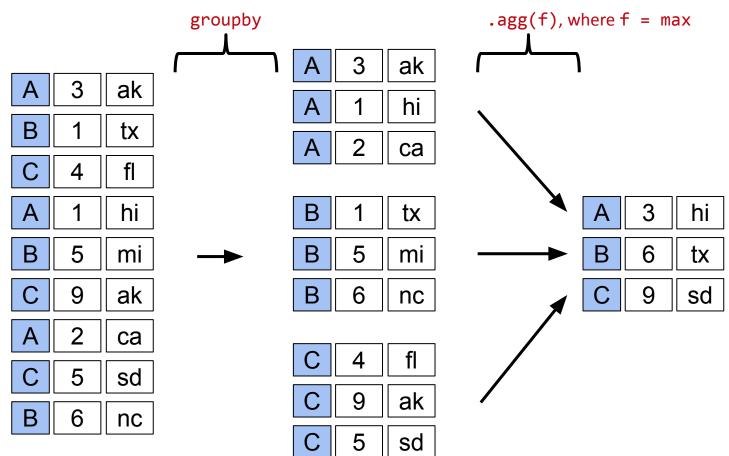


What goes in the ?? for row Δ ?

① Start presenting to display the poll results on this slide.











Why does the table seem to claim that Woodrow Wilson won the presidency in 2020?

elections.groupby("Party").agg(max).head(10)

	Year	Candidate	Popular vote	Result	%
Party					
American	1976	Thomas J. Anderson	873053	loss	21.554001
American Independent	1976	Lester Maddox	9901118	loss	13.571218
Anti-Masonic	1832	William Wirt	100715	loss	7.821583
Anti-Monopoly	1884	Benjamin Butler	134294	loss	1.335838
Citizens	1980	Barry Commoner	233052	loss	0.270182
Communist	1932	William Z. Foster	103307	loss	0.261069
Constitution	2016	Michael Peroutka	203091	loss	0.152398
Constitutional Union	1860	John Bell	590901	loss	12.639283
Democratic	2020	Woodrow Wilson	81268924	win	61.344703
Democratic-Republican	1824	John Quincy Adams	151271	win	57.210122





Why does the table seem to claim that Woodrow Wilson won the presidency in 2020?

elections.groupby("Party").agg(max).head(10)

Every column is calculated independently! Among Democrats:

- Last year they ran: 2020
- Alphabetically latest candidate name: Woodrow Wilson
- Highest % of vote: 61.34

%	Result	Popular vote	Candidate	Year	
					Party
21.554001	loss	873053	Thomas J. Anderson	1976	American
13.571218	loss	9901118	Lester Maddox	1976	American Independent
7.821583	loss	100715	William Wirt	1832	Anti-Masonic
1.335838	loss	134294	Benjamin Butler	1884	Anti-Monopoly
0.270182	loss	233052	Barry Commoner	1980	Citizens
0.261069	loss	103307	William Z. Foster	1932	Communist
0.152398	loss	203091	Michael Peroutka	2016	Constitution
12.639283	loss	590901	John Bell	1860	Constitutional Union
61.344703	win	81268924	Woodrow Wilson	2020	Democratic
57.210122	win	151271	John Quincy Adams	1824	Democratic-Republican





Very hard puzzle: Try to write code that returns the table below.

- Each row shows the best result (in %) by each party.
 - > For example: Best Democratic result ever was Johnson's 1964 win.

	Year	Candidate	Popular vote	Result	%
Party					
American	1856	Millard Fillmore	873053	loss	21.554001
American Independent	1968	George Wallace	9901118	loss	13.571218
Anti-Masonic	1832	William Wirt	100715	loss	7.821583
Anti-Monopoly	1884	Benjamin Butler	134294	loss	1.335838
Citizens	1980	Barry Commoner	233052	loss	0.270182
Communist	1932	William Z. Foster	103307	loss	0.261069
Constitution	2008	Chuck Baldwin	199750	loss	0.152398
Constitutional Union	1860	John Bell	590901	loss	12.639283
Democratic	1964	Lyndon Johnson	43127041	win	61.344703





Very hard puzzle: Try to write code that returns the table below.

- First sort the DataFrame so that rows are in descending order of %.
- Then group by Party and take the first item of each series.
- Note: Lab will give you a chance to try this out if you didn't quite follow during lecture.

elections_sorted_by_percent = elections.sort_values("%", ascending=False)
elections_sorted_by_percent.groupby("Party").agg(lambda x : x.iloc[0])

	Year	Candidate	Party	Popular vote	Result	%
114	1964	Lyndon Johnson	Democratic	43127041	win	61.344703
91	1936	Franklin Roosevelt	Democratic	27752648	win	60.978107
120	1972	Richard Nixon	Republican	47168710	win	60.907806
79	1920	Warren Harding	Republican	16144093	win	60.574501
133	1984	Ronald Reagan	Republican	54455472	win	59.023326

Candidate Popular vote Result Year **Party** Millard Fillmore 873053 American 1856 loss 21.554001 American Independent 1968 George Wallace 9901118 13.571218 100715 7.821583 Anti-Masonic 1832 William Wirt Anti-Monopoly Benjamin Butler 134294 1.335838 **Barry Commoner** 0.270182 Citizens 1980 233052 Communist 1932 William Z. Foster 103307 0.261069 Constitution 2008 Chuck Baldwin 199750 0.152398 Constitutional Union John Bell 590901 12.639283 Democratic 1964 Lyndon Johnson 43127041 win 61.344703

There's More Than One Way to Find the Best Result by Party



In Pandas, there's more than one way to get to the same answer.

- Each approach has different tradeoffs in terms of readability, performance, memory consumption, complexity, etc.
- Takes a very long time to understand these tradeoffs!
- If you find your current solution to be particularly convoluted or hard to read, maybe try finding another way!



Groupby Puzzle #1 - Alternate Approaches



Using a lambda function

```
elections_sorted_by_percent = elections.sort_values("%", ascending=False)
elections_sorted_by_percent.groupby("Party").agg(lambda x : x.iloc[0])
```

Using idxmax function

```
best_per_party = elections.loc[elections.groupby("Party")["%"].idxmax()]
```

Using drop_duplicates function

```
best_per_party2 = elections.sort_values("%").drop_duplicates(["Party"], keep="last")
```





Other DataFrameGroupBy Features

Lecture 04, Data 100 Spring 2023

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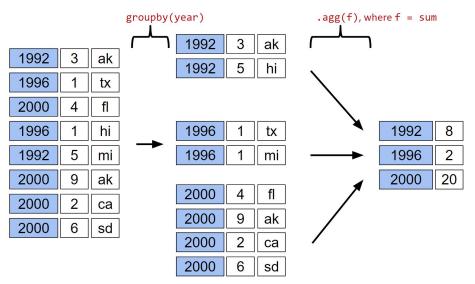


Revisiting groupby.agg



A groupby operation involves some combination of splitting the object, applying a function, and combining the results.

- So far, we've seen that df.groupby("year").agg(sum):
 - Organizes all rows with the same year into a subframe for that year.
 - Creates a new dataframe with one row representing each subframe year.
 - All integer rows in each subframe are combined using the sum function.





Raw groupby Objects



The result of a groupby operation applied to a DataFrame is a DataFrameGroupBy object.

• It is not a DataFrame!

```
grouped_by_year = elections.groupby("Year")
type(grouped_by_year)
```

pandas.core.groupby.generic.DataFrameGroupBy

Given a DataFrameGroupBy object, can use various functions to generate DataFrames (or Series). agg is only one choice:

- agg: Creates a new DataFrame with one aggregated row per subframe.
- max: Creates a new DataFrame aggregated using the max function.
- size: Creates a new Series with the size of each subframe.
- **filter**: Creates a copy of the original DataFrame, but keeping only rows from subframes that obey the provided condition.



More on DataFrameGroupby Object



We can look into DataFrameGroupby objects in following ways:

```
grouped_by_year = elections.groupby("Year")
grouped_by_year.groups.keys()
```

dict_keys([1824, 1828, 1832, 1836, 1840, 1844, 1848, 1852, 1856, 1860, 1864, 1868, 1872, 1876, 1880, 1884, 1888, 1892, 1896, 1900, 1904, 1908, 1912, 1916, 1920, 1924, 1928, 1932, 1936, 1940, 1944, 1948, 1952, 1956, 1960, 1964, 1968, 1972, 1976, 1980, 1984, 1988, 1992, 1996, 2000, 2004, 2008, 2012, 2016, 2020])

```
grouped_by_year.groups[2020]
```

Int64Index([178, 179, 180, 181], dtype='int64')

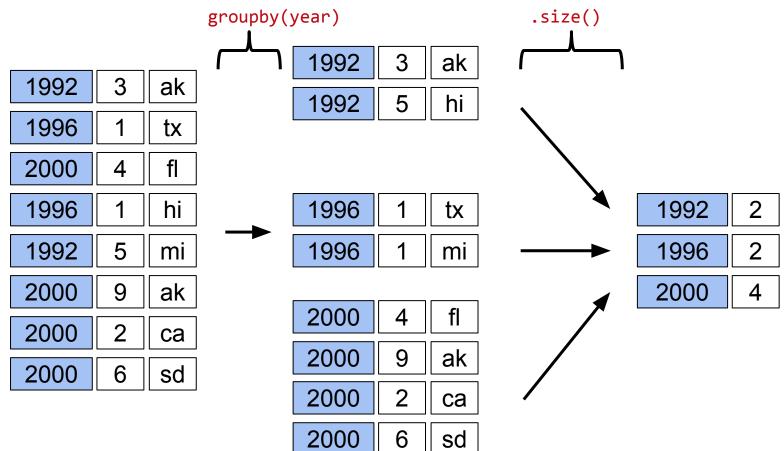
grouped_by_year.get_group[2020]

	Year	Candidate	Party	Popular vote	Result	%
178	2020	Joseph Biden	Democratic	81268924	win	51.311515
179	2020	Donald Trump	Republican	74216154	loss	46.858542
180	2020	Jo Jorgensen	Libertarian	1865724	loss	1.177979
181	2020	Howard Hawkins	Green	405035	loss	0.255731



groupby.size()







Filtering by Group

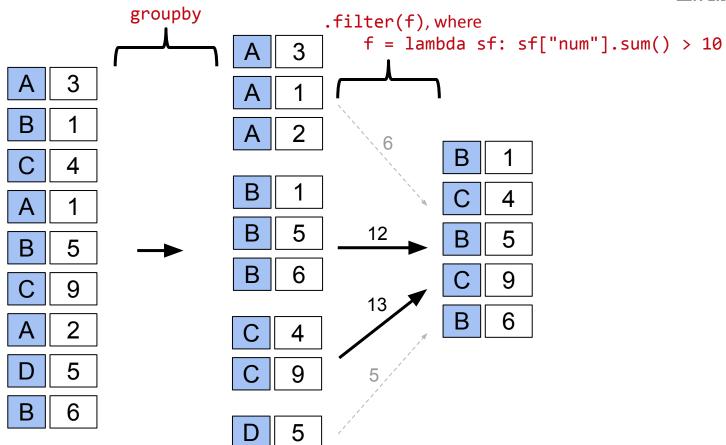


Another common use for groups is to filter data.

- groupby.filter takes an argument f.
- f is a function that:
 - Takes a DataFrame as input.
 - Returns either True or False.
- For each group g, f is applied to the subframe comprised of the rows from the original dataframe corresponding to that group.









Filtering Elections Dataset



Let's keep only election year results where the max '%' is less than 45%.

```
elections.groupby("Year")
    .filter(lambda sf: sf["%"].max() < 45)
    .set_index("Year")
    .sort_index()</pre>
```

	Candidate	Party	Popular vote	Result	%
Year					
1860	Abraham Lincoln	Republican	1855993	win	39.699408
1860	John Bell	Constitutional Union	590901	loss	12.639283
1860	John C. Breckinridge	Southern Democratic	848019	loss	18.138998
1860	Stephen A. Douglas	Northern Democratic	1380202	loss	29.522311
1912	Eugene V. Debs	Socialist	901551	loss	6.004354
1912	Eugene W. Chafin	Prohibition	208156	loss	1.386325
1912	Theodore Roosevelt	Progressive	4122721	loss	27.457433
1912	William Taft	Republican	3486242	loss	23.218466
1912	Woodrow Wilson	Democratic	6296284	win	41.933422
1968	George Wallace	American Independent	9901118	loss	13.571218
1968	Hubert Humphrey	Democratic	31271839	loss	42.863537
1968	Richard Nixon	Republican	31783783	win	43.565246
1992	Andre Marrou	Libertarian	290087	loss	0.278516
1992	Bill Clinton	Democratic	44909806	win	43.118485
1992	Bo Gritz	Populist	106152	loss	0.101918
1992	George H. W. Bush	Republican	39104550	loss	37.544784
1992	Ross Perot	Independent	19743821	loss	18.956298





Groupby and PivotTables

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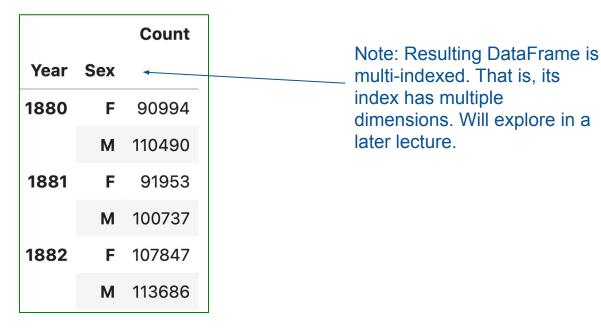
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Grouping by Multiple Columns

Suppose we want to build a table showing the total number of babies born of each sex in each year. One way is to **groupby** using both columns of interest:

Example: babynames.groupby(["Year", "Sex"]).agg(sum).head(6)





Pivot Tables



A more natural approach is to use our Data 8 brains and create a pivot table.

```
babynames_pivot = babynames.pivot_table(
    index="Year",  # rows (turned into index)
    columns="Sex",  # column values
    values=["Count"], # field(s) to process in each group
    aggfunc=np.sum,  # group operation
)
babynames_pivot.head(6)
```

		Count
Sex	F	М
Year		
1880	90994	110490
1881	91953	100737
1882	107847	113686
1883	112319	104625
1884	129019	114442
1885	133055	107799



groupby(["Year", "Sex"]) vs. pivot_table



The pivot table more naturally represents our data.

groupby output

		Count
Year	Sex	
1880	F	90994
	М	110490
1881	F	91953
	М	100737
1882	F	107847
	М	113686

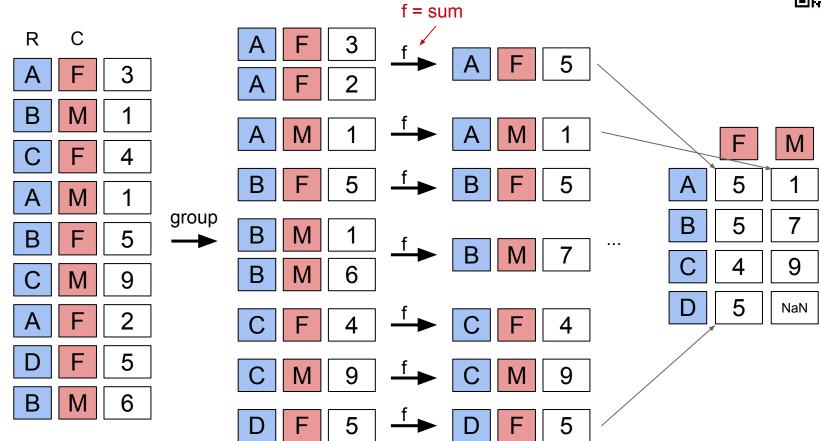
pivot_table output

		Count
Sex	F	М
Year		
1880	90994	110490
1881	91953	100737
1882	107847	113686
1883	112319	104625
1884	129019	114442
1885	133055	107799



Pivot Table Mechanics







Pivot Tables



We can include multiple values in our pivot tables.

```
babynames pivot = babynames.pivot table(
   index="Year",  # rows (turned into index)
   columns="Sex", # column values
   values=["Count", "Name"],
   aggfunc=np.max, # group operation
babynames pivot.head(6)
```

		Count		Name		
Sex	F	M	F	M		
Year						
1910	295	237	Yvonne	William		
1911	390	214	Zelma	Willis		
1912	534	501	Yvonne	Woodrow		
1913	584	614	Zelma	Yoshio		
1914	773	769	Zelma	Yoshio		
1915	998	1033	Zita	Yukio		





A Quick Look at Joining Tables

Lecture 04, Data 100 Spring 2023

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Joining Tables



Suppose want to know the 2020 popularity of presidential candidate's names.

 Example: Dwight Eisenhower's name Dwight is not popular today, with only 5 babies born with this name in California in 2020.

To solve this problem, we'll have to join tables.

This will be almost exactly like Table.join from data 8
 (http://data8.org/datascience/_autosummary/datascience.tables.Table.join.html)



Creating Table 1: Babynames in 2020



Let's set aside names from 2020 first:

```
babynames_2020 = babynames[babynames["Year"] == 2020]
```

babynames_2020

	Name	Sex	Count	Year
0	Olivia	F	17641	2020
1	Emma	F	15656	2020
2	Ava	F	13160	2020
3	Charlotte	F	13065	2020
4	Sophia	F	13036	2020
31448	Zykell	М	5	2020
31449	Zylus	М	5	2020
31450	Zymari	М	5	2020
31451	Zyn	М	5	2020
31452	Zyran	М	5	2020
31453 rc	ows × 4 col	umns		

Creating Table 2: Presidents with First Names



To join our table, we'll also need to set aside the first names of each candidate.

 You'll have a chance to write this code again on lab, so don't worry about the details too much.

elections["First Name"] = elections["Candidate"].str.split().str[0]

	Year	Candidate	Party	Popular vote	Result	%	First Name

177	2016	Jill Stein	Green	1457226	loss	1.073699	Jill
178	2020	Joseph Biden	Democratic	81268924	win	51.311515	Joseph
179	2020	Donald Trump	Republican	74216154	loss	46.858542	Donald
180	2020	Jo Jorgensen	Libertarian	1865724	loss	1.177979	Jo
181	2020	Howard Hawkins	Green	405035	loss	0.255731	Howard



Joining Our Tables



	Year_x	Candidate	Party	Popular vote	Result	%	First Name	State	Sex	Year_y	Name	Count
0	1824	Andrew Jackson	Democratic- Republican	151271	loss	57.210122	Andrew	CA	М	2020	Andrew	867
1	1828	Andrew Jackson	Democratic	642806	win	56.203927	Andrew	CA	М	2020	Andrew	867
2	1832	Andrew Jackson	Democratic	702735	win	54.574789	Andrew	CA	М	2020	Andrew	867
3	1824	John Quincy Adams	Democratic- Republican	113142	win	42.789878	John	CA	М	2020	John	617
4	1828	John Quincy Adams	National Republican	500897	loss	43.796073	John	CA	М	2020	John	617





LECTURE 4

Pandas, Part III

Content credit: Lisa Yan, Josh Hug

