

# Question 1

Step 1

Step 2 and 3

	user_id	age	gender	occupation	zip_code
0	1	24	M	technician	85711
1	2	53	F	other	94043
2	3	23	M	writer	32067
3	4	24	M	technician	43537
4	5	33	F	other	15213

Step 4

Mean age per occupation:

	occupation	mean_age
0	administrator	38.746835
1	artist	31.392857
2	doctor	43.571429
3	educator	42.010526
4	engineer	36.388060
5	entertainment	29.222222
6	executive	38.718750
7	healthcare	41.562500
8	homemaker	32.571429
9	lawyer	36.750000
10	librarian	40.000000
11	marketing	37.615385
12	none	26.555556
13	other	34.523810
14	programmer	33.121212
15	retired	63.071429
16	salesman	35.666667
17	scientist	35.548387
18	student	22.081633
19	technician	33.148148
20	writer	36.311111

Step 5

Male ratio per occupation (sorted descending):

occupation	
doctor	1.000000
engineer	0.970149
technician	0.962963
retired	0.928571
programmer	0.909091
executive	0.906250
scientist	0.903226
entertainment	0.888889
lawyer	0.833333
salesman	0.750000
educator	0.726316
student	0.693878
other	0.657143
marketing	0.615385
writer	0.577778
none	0.555556

```
administrator    0.544304
artist           0.535714
librarian        0.431373
healthcare       0.312500
homemaker        0.142857
dtype: float64
```

## Step 6

For each occupation, the minimum and maximum ages:

	min_age	max_age
occupation		
administrator	21	70
artist	19	48
doctor	28	64
educator	23	63
engineer	22	70
entertainment	15	50
executive	22	69
healthcare	22	62
homemaker	20	50
lawyer	21	53
librarian	23	69
marketing	24	55
none	11	55
other	13	64
programmer	20	63
retired	51	73
salesman	18	66
scientist	23	55
student	7	42
technician	21	55
writer	18	60

## Step 7

For each combination of occupation and sex, the mean age:

		mean_age
occupation	gender	
administrator	F	40.638889
	M	37.162791
artist	F	30.307692
	M	32.333333
doctor	M	43.571429
educator	F	39.115385
	M	43.101449
engineer	F	29.500000
	M	36.600000
entertainment	F	31.000000
	M	29.000000
executive	F	44.000000
	M	38.172414
healthcare	F	39.818182
	M	45.400000
homemaker	F	34.166667
	M	23.000000
lawyer	F	39.500000
	M	36.200000
librarian	F	40.000000
	M	40.000000
marketing	F	37.200000
	M	37.875000
none	F	36.500000
	M	18.600000
other	F	35.472222

	M	34.028986
programmer	F	32.166667
	M	33.216667
retired	F	70.000000
	M	62.538462
salesman	F	27.000000
	M	38.555556
scientist	F	28.333333
	M	36.321429
student	F	20.750000
	M	22.669118
technician	F	38.000000
	M	32.961538
writer	F	37.631579
	M	35.346154

## Step 8

For each occupation the percentage of women and men:

occupation	gender	
administrator	F	45.569620
	M	54.430380
artist	F	46.428571
	M	53.571429
doctor	M	100.000000
educator	F	27.368421
	M	72.631579
engineer	F	2.985075
	M	97.014925
entertainment	F	11.111111
	M	88.888889
executive	F	9.375000
	M	90.625000
healthcare	F	68.750000
	M	31.250000
homemaker	F	85.714286
	M	14.285714
lawyer	F	16.666667
	M	83.333333
librarian	F	56.862745
	M	43.137255
marketing	F	38.461538
	M	61.538462
none	F	44.444444
	M	55.555556
other	F	34.285714
	M	65.714286
programmer	F	9.090909
	M	90.909091
retired	F	7.142857
	M	92.857143
salesman	F	25.000000
	M	75.000000
scientist	F	9.677419
	M	90.322581
student	F	30.612245
	M	69.387755
technician	F	3.703704
	M	96.296296
writer	F	42.222222
	M	57.777778

dtype: float64

## Question 2

Step 1

Step 2 and 3

Euro12 data:

	Team	Goals	Shots on target	Shots off target	Shooting Accuracy \
0	Croatia	4	13	12	51.9%
1	Czech Republic	4	13	18	41.9%
2	Denmark	4	10	10	50.0%
3	England	5	11	18	50.0%
4	France	3	22	24	37.9%

	% Goals-to-shots	Total shots (inc. Blocked)	Hit Woodwork	Penalty goals \
0	16.0%	32	0	0
1	12.9%	39	0	0
2	20.0%	27	1	0
3	17.2%	40	0	0
4	6.5%	65	1	0

	Penalties not scored ...	Saves made	Saves-to-shots ratio	Fouls Won \
0	0 ...	13	81.3%	41
1	0 ...	9	60.1%	53
2	0 ...	10	66.7%	25
3	0 ...	22	88.1%	43
4	0 ...	6	54.6%	36

	Fouls Conceded	Offsides	Yellow Cards	Red Cards	Subs on	Subs off \
0	62	2	9	0	9	9
1	73	8	7	0	11	11
2	38	8	4	0	7	7
3	45	6	5	0	11	11
4	51	5	6	0	11	11

	Players Used
0	16
1	19
2	15
3	16
4	19

[5 rows x 35 columns]

(16, 35)

Step 4

Only the goals column:

	Goals
0	4
1	4
2	4
3	5
4	3
5	10
6	5
7	6
8	2
9	2

```
10      6
11      1
12      5
13     12
14      5
15      2
```

#### Step 5

Teams participated in the Euro2012:

```
['Croatia' 'Czech Republic' 'Denmark' 'England' 'France' 'Germany'
 'Greece' 'Italy' 'Netherlands' 'Poland' 'Portugal' 'Republic of Ireland'
 'Russia' 'Spain' 'Sweden' 'Ukraine']
```

Total teams: 16

#### Step 6

Number of columns in the dataset: 35

#### Step 7

Discipline dataframe:

	Team	Yellow Cards	Red Cards
0	Croatia	9	0
1	Czech Republic	7	0
2	Denmark	4	0
3	England	5	0
4	France	6	0
5	Germany	4	0
6	Greece	9	1
7	Italy	16	0
8	Netherlands	5	0
9	Poland	7	1
10	Portugal	12	0
11	Republic of Ireland	6	1
12	Russia	6	0
13	Spain	11	0
14	Sweden	7	0
15	Ukraine	5	0

#### Step 8

Sorted discipline:

	Team	Yellow Cards	Red Cards
6	Greece	9	1
9	Poland	7	1
11	Republic of Ireland	6	1
7	Italy	16	0
10	Portugal	12	0
13	Spain	11	0
0	Croatia	9	0
1	Czech Republic	7	0
14	Sweden	7	0
4	France	6	0
12	Russia	6	0
3	England	5	0
8	Netherlands	5	0
15	Ukraine	5	0
2	Denmark	4	0
5	Germany	4	0

```
<ipython-input-16-2b310c00d462>:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_](https://pandas.pydata.org/pandas-docs/stable/user_)

## Step 9

Mean yellow cards per team:

	mean_yellow_cards
Team	
Croatia	9.0
Czech Republic	7.0
Denmark	4.0
England	5.0
France	6.0
Germany	4.0
Greece	9.0
Italy	16.0
Netherlands	5.0
Poland	7.0
Portugal	12.0
Republic of Ireland	6.0
Russia	6.0
Spain	11.0
Sweden	7.0
Ukraine	5.0

## Step 10

Teams with more than 6 goals:

	Team	Goals	Shots on target	Shots off target	Shooting Accuracy	\
5	Germany	10	32	32	47.8%	
13	Spain	12	42	33	55.9%	

	% Goals-to-shots	Total shots (inc. Blocked)	Hit Woodwork	Penalty goals	\
5	15.6%	80	2	1	
13	16.0%	100	0	1	

	Penalties not scored	...	Saves made	Saves-to-shots ratio	Fouls Won	\
5	0	...	10	62.6%	63	
13	0	...	15	93.8%	102	

	Fouls Conceded	Offsides	Yellow Cards	Red Cards	Subs on	Subs off	\
5	49	12	4	0	15	15	
13	83	19	11	0	17	17	

	Players Used
5	17
13	18

[2 rows x 35 columns]

## Step 11

Teams starting with a G:

	Team	Goals	Shots on target	Shots off target	Shooting Accuracy	\
5	Germany	10	32	32	47.8%	
6	Greece	5	8	18	30.7%	

	% Goals-to-shots	Total shots (inc. Blocked)	Hit Woodwork	Penalty goals	\
5	15.6%	80	2	1	
6	19.2%	32	1	1	

	Penalties not scored	...	Saves made	Saves-to-shots ratio	Fouls Won	\
5	0	...	10	62.6%	63	

6	1	...	13	65.1%	67	
	Fouls Conceded	Offsides	Yellow Cards	Red Cards	Subs on	Subs off \
5	49	12	4	0	15	15
6	48	12	9	1	12	12
	Players Used					
5	17					
6	20					

[2 rows x 35 columns]

## Step 12

First 7 columns from euro12:

	Team	Goals	Shots on target	Shots off target \
0	Croatia	4	13	12
1	Czech Republic	4	13	18
2	Denmark	4	10	10
3	England	5	11	18
4	France	3	22	24
5	Germany	10	32	32
6	Greece	5	8	18
7	Italy	6	34	45
8	Netherlands	2	12	36
9	Poland	2	15	23
10	Portugal	6	22	42
11	Republic of Ireland	1	7	12
12	Russia	5	9	31
13	Spain	12	42	33
14	Sweden	5	17	19
15	Ukraine	2	7	26

	Shooting Accuracy %	Goals-to-shots	Total shots (inc. Blocked)
0	51.9%	16.0%	32
1	41.9%	12.9%	39
2	50.0%	20.0%	27
3	50.0%	17.2%	40
4	37.9%	6.5%	65
5	47.8%	15.6%	80
6	30.7%	19.2%	32
7	43.0%	7.5%	110
8	25.0%	4.1%	60
9	39.4%	5.2%	48
10	34.3%	9.3%	82
11	36.8%	5.2%	28
12	22.5%	12.5%	59
13	55.9%	16.0%	100
14	47.2%	13.8%	39
15	21.2%	6.0%	38

## Step 13

All columns except last 3 columns from euro12:

	Team	Goals	Shots on target	Shots off target \
0	Croatia	4	13	12
1	Czech Republic	4	13	18
2	Denmark	4	10	10
3	England	5	11	18
4	France	3	22	24
5	Germany	10	32	32
6	Greece	5	8	18
7	Italy	6	34	45
8	Netherlands	2	12	36
9	Poland	2	15	23

10	Portugal	6	22	42
11	Republic of Ireland	1	7	12
12	Russia	5	9	31
13	Spain	12	42	33
14	Sweden	5	17	19
15	Ukraine	2	7	26

	Shooting Accuracy %	Goals-to-shots	Total shots (inc. Blocked)	\
0	51.9%	16.0%		32
1	41.9%	12.9%		39
2	50.0%	20.0%		27
3	50.0%	17.2%		40
4	37.9%	6.5%		65
5	47.8%	15.6%		80
6	30.7%	19.2%		32
7	43.0%	7.5%		110
8	25.0%	4.1%		60
9	39.4%	5.2%		48
10	34.3%	9.3%		82
11	36.8%	5.2%		28
12	22.5%	12.5%		59
13	55.9%	16.0%		100
14	47.2%	13.8%		39
15	21.2%	6.0%		38

	Hit Woodwork	Penalty goals	Penalties not scored	...	Clean Sheets	\
0	0	0	0	...	0	
1	0	0	0	...	1	
2	1	0	0	...	1	
3	0	0	0	...	2	
4	1	0	0	...	1	
5	2	1	0	...	1	
6	1	1	1	...	1	
7	2	0	0	...	2	
8	2	0	0	...	0	
9	0	0	0	...	0	
10	6	0	0	...	2	
11	0	0	0	...	0	
12	2	0	0	...	0	
13	0	1	0	...	5	
14	3	0	0	...	1	
15	0	0	0	...	0	

	Blocks	Goals conceded	Saves made	Saves-to-shots ratio	Fouls Won	\
0	10	3	13	81.3%	41	
1	10	6	9	60.1%	53	
2	10	5	10	66.7%	25	
3	29	3	22	88.1%	43	
4	7	5	6	54.6%	36	
5	11	6	10	62.6%	63	
6	23	7	13	65.1%	67	
7	18	7	20	74.1%	101	
8	9	5	12	70.6%	35	
9	8	3	6	66.7%	48	
10	11	4	10	71.5%	73	
11	23	9	17	65.4%	43	
12	8	3	10	77.0%	34	
13	8	1	15	93.8%	102	
14	12	5	8	61.6%	35	
15	4	4	13	76.5%	48	

	Fouls Conceded	Offsides	Yellow Cards	Red Cards
0	62	2	9	0
1	73	8	7	0
2	38	8	4	0
3	45	6	5	0



4	51	5	6	0
5	49	12	4	0
6	48	12	9	1
7	89	16	16	0
8	30	3	5	0
9	56	3	7	1
10	90	10	12	0
11	51	11	6	1
12	43	4	6	0
13	83	19	11	0
14	51	7	7	0
15	31	4	5	0

[16 rows x 32 columns]

Step 14

Shooting accuracy from England, Italy and Russia:

	Team	Shooting Accuracy
3	England	50.0%
7	Italy	43.0%
12	Russia	22.5%

# Question 3

## Step 1

```
Series 1:
0      4
1      3
2      3
3      2
4      3
..
95     2
96     3
97     3
98     3
99     1
Length: 100, dtype: int64
Series 2:
0      2
1      2
2      1
3      2
4      2
..
95     1
96     2
97     1
98     2
99     2
Length: 100, dtype: int64
Series 3:
0      18589
1      10167
2      10414
3      25928
4      14552
...
95      13936
96      18983
97      28013
98      20548
99      24990
Length: 100, dtype: int64
```

## Step 3

```
0  1  2
0  4  2  18589
1  3  2  10167
2  3  1  10414
3  2  2  25928
4  3  2  14552
```

## Step 4

```
After column name changed:
   bedrs  bathrs  price_sqr_meter
0      4      2          18589
1      3      2          10167
2      3      1          10414
```

3	2	2	25928
4	3	2	14552

## Step 5

Bigcolumn dataframe:

0	4
1	3
2	3
3	2
4	3

	...
95	13936
96	18983
97	28013
98	20548
99	24990

Length: 300, dtype: int64

## Step 6

Bigcolumn index:

```
Int64Index([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9,
            ...,
            90, 91, 92, 93, 94, 95, 96, 97, 98, 99],
           dtype='int64', length=300)
```

## Step 7

After index reset:

0	4
1	3
2	3
3	2
4	3

	...
295	13936
296	18983
297	28013
298	20548
299	24990

Length: 300, dtype: int64

Date	RPT	VAL	ROS	KIL	SHA	BIR	DUB	CLA	MUL	\
------	-----	-----	-----	-----	-----	-----	-----	-----	-----	---

1961-01-01	15.04	14.96	13.17	9.29	NaN	9.87	13.67	10.25	10.83
1961-01-02	14.71	NaN	10.83	6.50	12.62	7.67	11.50	10.04	9.79
1961-01-03	18.50	16.88	12.33	10.13	11.17	6.17	11.25	NaN	8.50
1961-01-04	10.58	6.63	11.75	4.58	4.54	2.88	8.63	1.79	5.83
1961-01-05	13.33	13.25	11.42	6.17	10.71	8.21	11.92	6.54	10.92

	CLO	BEL	MAL
Date			
1961-01-01	12.58	18.50	15.04
1961-01-02	9.67	17.54	13.83
1961-01-03	7.67	12.75	12.71
1961-01-04	5.88	5.46	10.88
1961-01-05	10.34	12.92	11.83

Step 6

Missing values per location:

```

RPT      6
VAL      3
ROS      2
KIL      5
SHA      2
BIR      0
DUB      3
CLA      2
MUL      3
CLO      1
BEL      0
MAL      4
dtype: int64

```

Step 7

Non-missing values in total:

78857

Step 8

Mean windspeeds of the windspeeds over all the locations and all the times:

10.22788376428218

Step 9

Min, max and mean windspeeds and standard deviations of the windspeeds at each location:

	min	std	mean	max
RPT	0.67	5.618413	12.362987	35.80
VAL	0.21	5.267356	10.644314	33.37
ROS	1.50	5.008450	11.660526	33.84
KIL	0.00	3.605811	6.306468	28.46
SHA	0.13	4.936125	10.455834	37.54
BIR	0.00	3.968683	7.092254	26.16
DUB	0.00	4.977555	9.797343	30.37
CLA	0.00	4.499449	8.495053	31.08
MUL	0.00	4.166872	8.493590	25.88
CLO	0.04	4.503954	8.707332	28.21
BEL	0.13	5.835037	13.121007	42.38
MAL	0.67	6.699794	15.599079	42.54

Step 10

day\_stats: min, max and mean windspeed and standard deviations of the windspeeds across all the locations at each day:

	min	max	mean	std
Day				

1	2.402500	29.994167	10.532620	4.418214
2	3.222500	26.236667	10.390365	4.514749
3	3.165000	21.517500	9.920201	3.939575
4	2.598333	21.715000	9.806661	4.015167
5	2.282500	23.771667	10.114915	3.917657
6	2.062500	23.283333	10.015224	4.453977
7	2.840833	25.624167	10.101673	4.395235
8	2.802500	20.981667	9.902049	4.065742
9	2.890000	24.764167	10.399803	4.197974
10	3.086667	23.653333	10.529325	4.491358
11	3.212500	24.160000	10.560683	4.481103
12	2.325833	25.039167	10.495718	4.871524
13	3.037500	23.613333	9.889074	4.289324
14	3.226667	25.355833	10.372138	4.528647
15	2.498333	26.139167	10.073495	4.417735
16	3.035000	24.403333	10.288977	4.590038
17	3.295833	28.841667	10.374973	4.746188
18	2.926667	21.880000	10.152459	4.396713
19	2.567500	24.673333	10.220950	4.376312
20	3.017500	23.721667	10.068786	4.209427
21	2.287500	23.703333	9.774354	4.289666
22	2.750833	24.384167	10.006636	4.350735
23	2.514167	23.600000	10.505355	3.999849
24	2.234167	23.663333	10.396698	4.136423
25	2.590000	25.467500	10.435988	4.261661
26	2.653333	23.975000	9.927810	4.319961
27	2.496667	26.843333	10.345127	4.636004
28	2.485000	23.921667	10.209516	4.166212
29	3.373333	22.349167	10.341877	4.254877
30	3.650000	22.736667	10.629337	4.044755
31	2.773333	22.740000	10.410893	4.227855

Step 11

Average windspeed for January month for each location:

```

RPT    14.847325
VAL    12.914560
ROS    13.299624
KIL     7.199498
SHA    11.667734
BIR     8.054839
DUB    11.819355
CLA     9.512047
MUL     9.543208
CLO    10.053566
BEL    14.550520
MAL    18.028763
dtype: float64

```

Step 12

Downsample to a yearly frequency for each location:

	RPT	VAL	ROS	KIL	SHA	BIR \
Year						
1961	12.299583	10.351796	11.362369	6.958227	10.881763	7.729726
1962	12.246923	10.110438	11.732712	6.960440	10.657918	7.393068
1963	12.813452	10.836986	12.541151	7.330055	11.724110	8.434712
1964	12.363661	10.920164	12.104372	6.787787	11.454481	7.570874
1965	12.451370	11.075534	11.848767	6.858466	11.024795	7.478110
1966	13.461973	11.557205	12.020630	7.345726	11.805041	7.793671
1967	12.737151	10.990986	11.739397	7.143425	11.630740	7.368164
1968	11.835628	10.468197	11.409754	6.477678	10.760765	6.067322
1969	11.166356	9.723699	10.902000	5.767973	9.873918	6.189973
1970	12.600329	10.726932	11.730247	6.217178	10.567370	7.609452
1971	11.273123	9.095178	11.088329	5.241507	9.440329	6.097151

1972	12.463962	10.561311	12.058333	5.929699	9.430410	6.358825
1973	11.828466	10.680493	10.680493	5.547863	9.640877	6.548740
1974	13.643096	11.811781	12.336356	6.427041	11.110986	6.809781
1975	12.008575	10.293836	11.564712	5.269096	9.190082	5.668521
1976	11.737842	10.203115	10.761230	5.109426	8.846339	6.311038
1977	13.099616	11.144493	12.627836	6.073945	10.003836	8.586438
1978	12.504356	11.044274	11.380000	6.082356	10.167233	7.650658

	DUB	CLA	MUL	CLO	BEL	MAL
Year						
1961	9.733923	8.858788	8.647652	9.835577	13.502795	13.680773
1962	11.020712	8.793753	8.316822	9.676247	12.930685	14.323956
1963	11.075699	10.336548	8.903589	10.224438	13.638877	14.999014
1964	10.259153	9.467350	7.789016	10.207951	13.740546	14.910301
1965	10.618712	8.879918	7.907425	9.918082	12.964247	15.591644
1966	10.579808	8.835096	8.514438	9.768959	14.265836	16.307260
1967	10.652027	9.325616	8.645014	9.547425	14.774548	17.135945
1968	8.859180	8.255519	7.224945	7.832978	12.808634	15.017486
1969	8.564493	7.711397	7.924521	7.754384	12.621233	15.762904
1970	9.609890	8.334630	9.297616	8.289808	13.183644	16.456027
1971	8.385890	6.757315	7.915370	7.229753	12.208932	15.025233
1972	9.704508	7.680792	8.357295	7.515273	12.727377	15.028716
1973	8.482110	7.614274	8.245534	7.812411	12.169699	15.441096
1974	10.084603	9.896986	9.331753	8.736356	13.252959	16.947671
1975	8.562603	7.843836	8.797945	7.382822	12.631671	15.307863
1976	9.149126	7.146202	8.883716	7.883087	12.332377	15.471448
1977	11.523205	8.378384	9.098192	8.821616	13.459068	16.590849
1978	9.489342	8.800466	9.089753	8.301699	12.967397	16.771370

Step 13

Downsample to a monthly frequency for each location:

	RPT	VAL	ROS	KIL	SHA	BIR \
Month						
1	14.841333	11.988333	13.431613	7.736774	11.072759	8.588065
2	16.269286	14.975357	14.441481	9.230741	13.852143	10.937500
3	10.890000	11.296452	10.752903	7.284000	10.509355	8.866774
4	10.722667	9.427667	9.998000	5.830667	8.435000	6.495000
5	9.860968	8.850000	10.818065	5.905333	9.490323	6.574839
...	...	...	...	...	...	...
8	9.645161	8.259355	9.032258	4.502903	7.368065	5.935161
9	10.913667	10.895000	10.635000	5.725000	10.372000	9.278333
10	9.897742	8.670968	9.295806	4.721290	8.525161	6.774194
11	16.151667	14.802667	13.508000	7.317333	11.475000	8.743000
12	16.175484	13.748065	15.635161	7.094839	11.398710	9.241613

	DUB	CLA	MUL	CLO	BEL	MAL
Month						
1	11.184839	9.245333	9.085806	10.107419	13.880968	14.703226
2	11.890714	11.846071	11.821429	12.714286	18.583214	15.411786
3	9.644194	9.829677	10.294138	11.251935	16.410968	15.720000
4	6.925333	7.094667	7.342333	7.237000	11.147333	10.278333
5	7.604000	8.177097	8.039355	8.499355	11.900323	12.011613
...	...	...	...	...	...	...
8	5.650323	5.417742	7.241290	5.536774	10.466774	12.054194
9	10.790333	9.583000	10.069333	8.939000	15.680333	19.391333
10	8.115484	7.337742	8.297742	8.243871	13.776774	17.150000
11	11.492333	9.657333	10.701333	10.676000	17.404667	20.723000
12	12.077419	10.194839	10.616774	11.028710	13.859677	21.371613

[216 rows x 12 columns]

Step 14

Downsample to a weekly frequency for each location:

	RPT	VAL	ROS	KIL	SHA	BIR \
Week						
1	15.040000	14.960000	13.170000	9.290000	NaN	9.870000
1	13.541429	11.486667	10.487143	6.417143	9.474286	6.435714
1	12.468571	8.967143	11.958571	4.630000	7.351429	5.072857
1	13.204286	9.862857	12.982857	6.328571	8.966667	7.417143
1	19.880000	16.141429	18.225714	12.720000	17.432857	14.828571
...	...	...	...	...	...	...
12	14.934286	11.232857	13.941429	5.565714	10.215714	8.618571
12	20.740000	19.190000	17.034286	9.777143	15.287143	12.774286
12	16.758571	14.692857	14.987143	6.917143	11.397143	7.272857
12	11.155714	8.008571	13.172857	4.004286	7.825714	6.290000
12	14.951429	11.801429	16.035714	6.507143	9.660000	8.620000

	DUB	CLA	MUL	CLO	BEL	MAL
Week						
1	13.670000	10.250000	10.830000	12.580000	18.500000	15.040000
1	11.061429	6.616667	8.434286	8.497143	12.481429	13.238571
1	7.535714	6.820000	5.712857	7.571429	11.125714	11.024286
1	9.257143	7.875714	7.145714	8.124286	9.821429	11.434286
1	15.528571	15.160000	14.480000	15.640000	20.930000	22.530000
...	...	...	...	...	...	...
12	9.642857	7.685714	9.011429	9.547143	11.835714	18.728571
12	14.437143	12.488571	13.870000	14.082857	18.517143	23.061429
12	10.208571	7.967143	9.168571	8.565714	11.102857	15.562857
12	7.798571	8.667143	7.151429	8.072857	11.845714	18.977143
12	13.708571	10.477143	10.868571	11.471429	12.947143	26.844286

[940 rows x 12 columns]

## Step 15

	min	max	mean	std
Week				
01	7.895000	18.020000	12.311667	2.909304
02	6.167143	13.458571	9.640496	2.575846
03	4.624286	13.017143	8.391310	2.842770
04	7.150000	13.392857	9.925556	2.183014
05	12.357143	22.340000	16.729028	2.988786
06	9.311429	18.582857	12.918214	3.030879
07	10.770000	22.152857	15.569405	3.305013
08	8.281429	19.234286	12.590655	2.902097
09	8.315714	17.005714	12.920714	2.492463
10	7.300000	16.444286	10.751290	2.313917
11	7.494286	18.057143	11.617877	2.945308
12	7.815714	18.577143	11.831429	3.109227
13	5.505714	17.312857	9.403571	3.148063
14	7.528333	14.024286	10.773651	1.728760
15	5.798571	12.547143	8.938095	2.207989
16	4.458571	9.828571	6.573690	1.837791
17	8.227143	13.595714	10.643214	1.838235
18	5.515714	11.021429	8.193690	1.998838
19	10.090000	17.552857	13.620238	2.145580
20	4.314286	8.702857	6.670972	1.374801
21	4.061429	11.411429	7.300000	2.351056
22	4.780000	12.702857	8.064008	2.344159
23	5.708571	12.960000	8.978690	2.271478
24	5.115714	12.114286	8.172500	2.081590
25	6.698571	16.155714	10.533690	2.820740
26	5.572857	16.660000	10.036071	3.213309
27	7.201429	15.904286	10.340813	2.563225
28	6.832857	13.797143	9.196647	2.078194
29	8.198571	16.054286	11.594643	2.312691
30	2.852857	7.987143	5.314048	1.695954
31	5.918571	14.267143	9.520357	2.282987



32	6.077143	13.218571	9.377857	2.037304
33	5.900000	12.315000	9.351726	1.956986
34	9.011429	17.435714	13.237262	2.512534
35	7.215714	16.328571	11.718929	2.644762
36	4.321429	11.472857	6.804048	2.053783
37	6.387143	12.648571	9.065476	2.038477
38	10.588571	18.920000	14.296667	2.942733
39	4.985714	12.870000	7.977381	2.279029
40	8.762857	16.580000	11.704921	2.617872
41	5.204286	11.682857	8.090833	2.294498
42	6.554286	16.944286	10.174821	3.229020
43	11.114286	23.695714	15.876905	3.559088
44	7.778571	17.778571	11.994266	3.020417
45	7.064286	18.178571	10.990595	3.129724
46	4.744286	14.255714	8.477381	2.962276
47	3.948571	14.381429	6.750238	3.059382
48	5.000000	12.328571	8.290020	2.348823
49	7.310000	16.630000	11.318333	2.960049
50	9.088571	17.150000	13.507619	2.783435
51	7.007143	13.675714	10.988452	2.135537
52	6.457143	16.815714	10.121190	3.215027

# Question 5

Step 1

Step 2 and 3

Chipo:

```
order_id  quantity  item_name \
0         1         1    Chips and Fresh Tomato Salsa
1         1         1                      Izze
2         1         1          Nantucket Nectar
3         1         1  Chips and Tomatillo-Green Chili Salsa
4         2         2          Chicken Bowl
```

```
choice_description  item_price
0                  NaN      $2.39
1          [Clementine]      $3.39
2              [Apple]      $3.39
3                  NaN      $2.39
4  [Tomatillo-Red Chili Salsa (Hot), [Black Beans...  $16.98
```

Out[119]:

	order_id	quantity	item_name	choice_description	item_price
0	1	1	Chips and Fresh Tomato Salsa	NaN	\$2.39
1	1	1	Izze	[Clementine]	\$3.39
2	1	1	Nantucket Nectar	[Apple]	\$3.39
3	1	1	Chips and Tomatillo-Green Chili Salsa	NaN	\$2.39
4	2	2	Chicken Bowl	[Tomatillo-Red Chili Salsa (Hot), [Black Beans...	\$16.98

Step 4

First 10 chipo entries:

```
order_id  quantity  item_name \
0         1         1    Chips and Fresh Tomato Salsa
1         1         1                      Izze
2         1         1          Nantucket Nectar
3         1         1  Chips and Tomatillo-Green Chili Salsa
4         2         2          Chicken Bowl
5         3         1          Chicken Bowl
6         3         1        Side of Chips
7         4         1        Steak Burrito
8         4         1    Steak Soft Tacos
9         5         1        Steak Burrito
```

```
choice_description  item_price
0                  NaN      $2.39
1          [Clementine]      $3.39
2              [Apple]      $3.39
3                  NaN      $2.39
4  [Tomatillo-Red Chili Salsa (Hot), [Black Beans...  $16.98
5  [Fresh Tomato Salsa (Mild), [Rice, Cheese, Sou...  $10.98
6                  NaN      $1.69
7  [Tomatillo Red Chili Salsa, [Fajita Vegetables...  $11.75
8  [Tomatillo Green Chili Salsa, [Pinto Beans, Ch...  $9.25
9  [Fresh Tomato Salsa, [Rice, Black Beans, Pinto...  $9.25
```

Step 5

Number of observations in Chipo: 4622

Step 6

Number of columns in Chipo: 5

Step 7

All the column names:

```
['order_id' 'quantity' 'item_name' 'choice_description' 'item_price']
```

Step 8

Dataset indexing type: RangeIndex(start=0, stop=4622, step=1)

Step 9

Most ordered item name: Chicken Bowl

Step 10

Most ordered item count: 761

Step 11

Most ordered item from choice description column: Rice

Step 12

Total items that were ordered: 2497

Step 13

Item price type: object

	order_id	quantity	item_name \
0	1	1	Chips and Fresh Tomato Salsa
1	1	1	Izze
2	1	1	Nantucket Nectar
3	1	1	Chips and Tomatillo-Green Chili Salsa
4	2	2	Chicken Bowl

	choice_description	item_price
0	[]	2.39
1	[Clementine]	3.39
2	[Apple]	3.39
3	[]	2.39
4	[Tomatillo-Red Chili Salsa (Hot), Black Beans,...	16.98

Item price type: float64

Step 14

Revenue for the period in the dataset: 39237.02 \$

Step 15

Total number of orders: 1834

Step 16

Average revenue amount per order: 21.39423118865867 \$

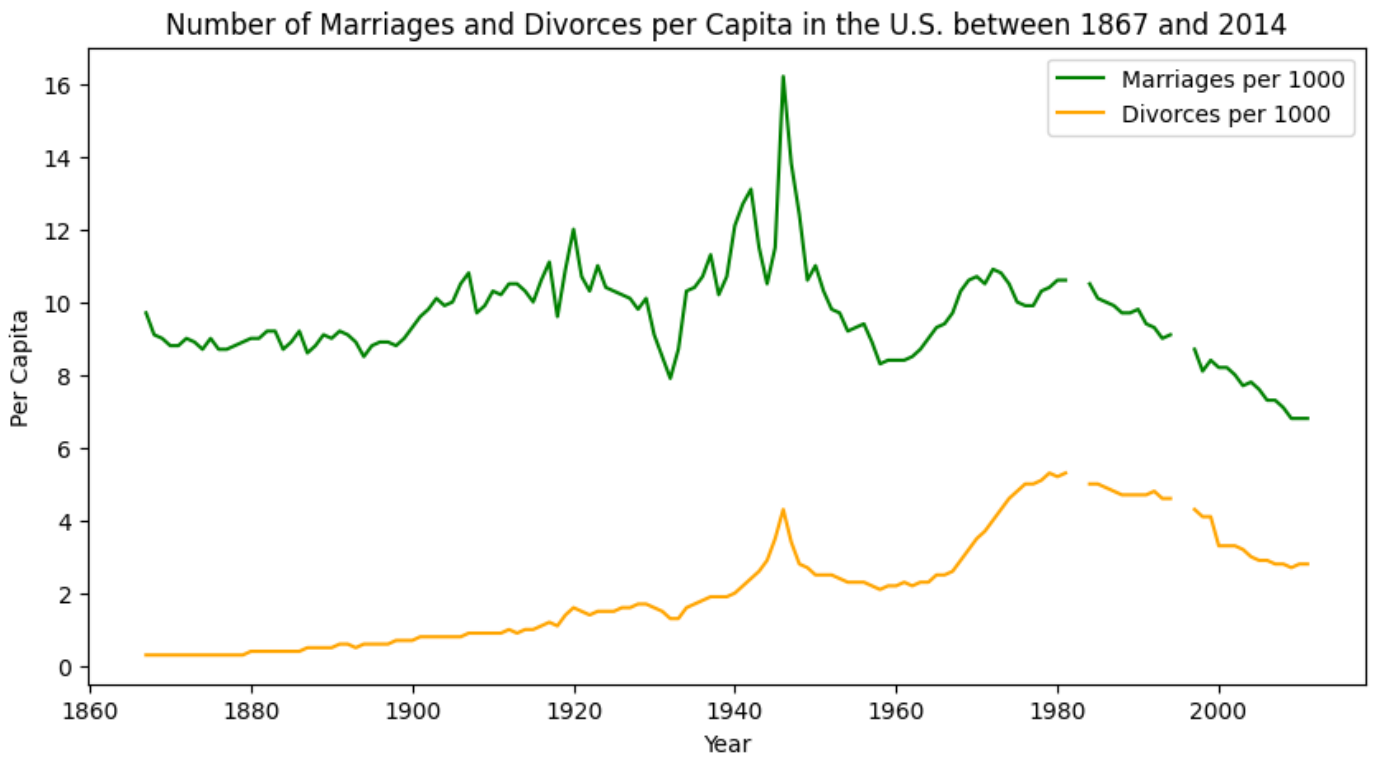
Step 17

Total number of different items sold: 50

## Question 6

	Year	Marriages	Divorces	Population	Marriages_per_1000	\
0	1867	357000.0	10000.0	36970000	9.7	
1	1868	345000.0	10000.0	37885000	9.1	
2	1869	348000.0	11000.0	38870000	9.0	
3	1870	352000.0	11000.0	39905000	8.8	
4	1871	359000.0	12000.0	41010000	8.8	

	Divorces_per_1000
0	0.3
1	0.3
2	0.3
3	0.3
4	0.3

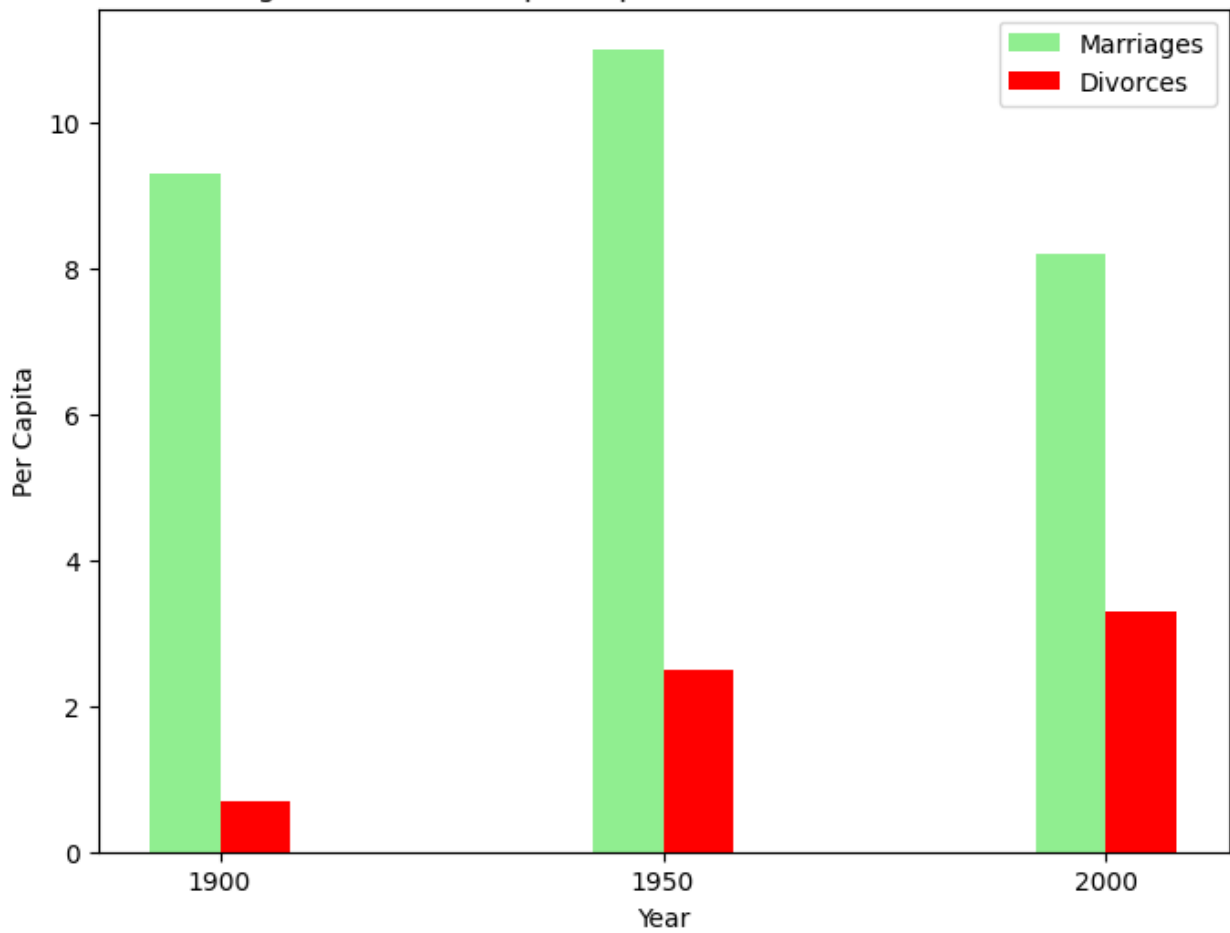


## Question 7

	Year	Marriages	Divorces	Population	Marriages_per_1000	\
0	1867	357000.0	10000.0	36970000	9.7	
1	1868	345000.0	10000.0	37885000	9.1	
2	1869	348000.0	11000.0	38870000	9.0	
3	1870	352000.0	11000.0	39905000	8.8	
4	1871	359000.0	12000.0	41010000	8.8	

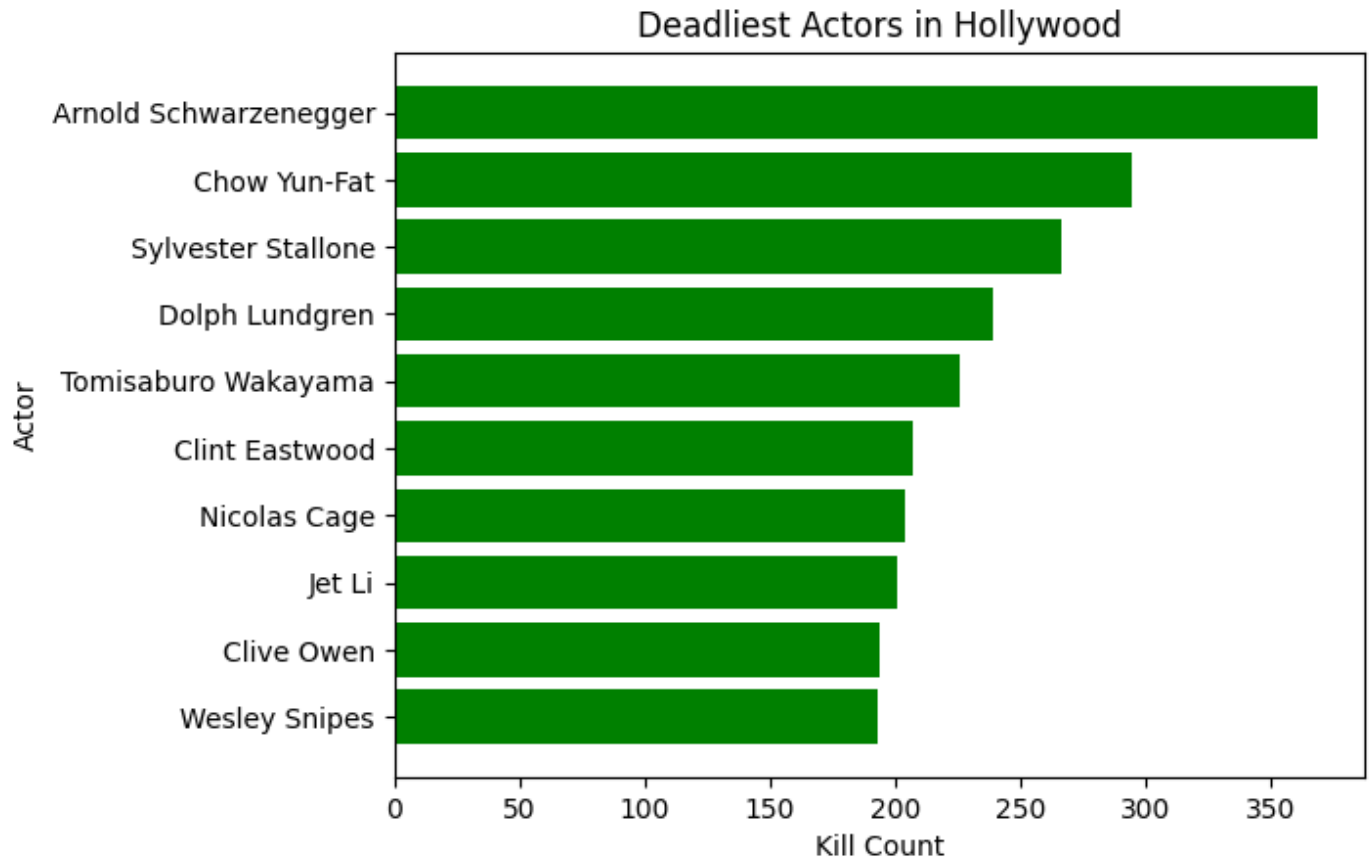
	Divorces_per_1000
0	0.3
1	0.3
2	0.3
3	0.3
4	0.3

Number of marriages and divorces per capita in the U.S. between 1900, 1950, and 2000.



## Question 8

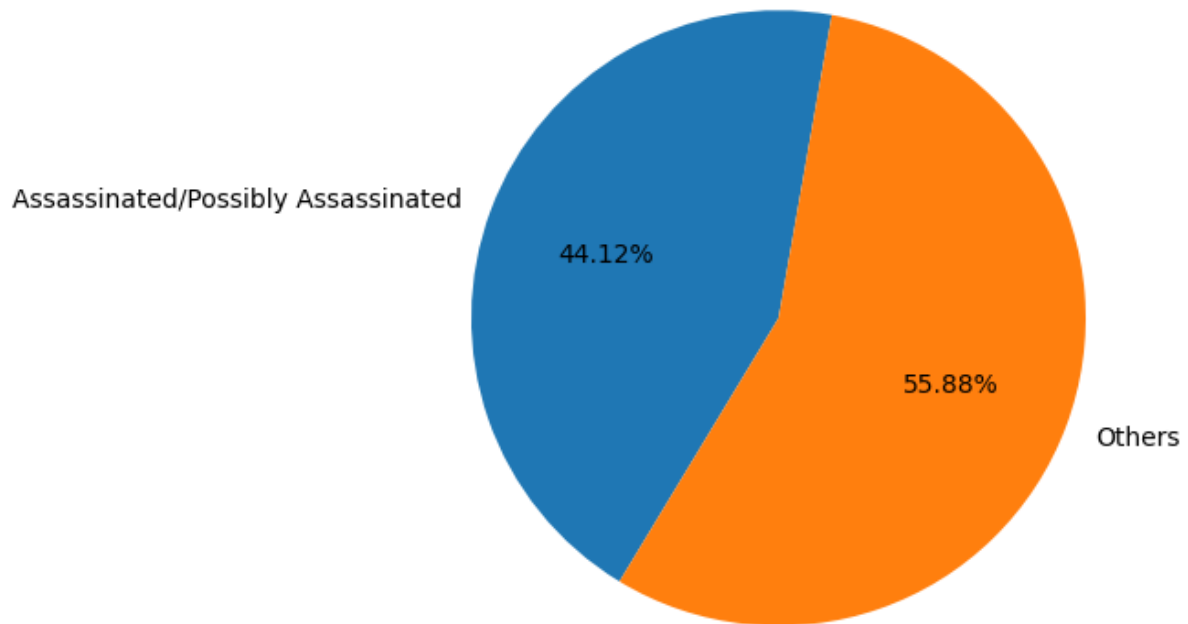
	Actor	Count
0	Arnold Schwarzenegger	369
1	Chow Yun-Fat	295
2	Clint Eastwood	207
3	Clive Owen	194
4	Dolph Lundgren	239



## Question 9

	Emperor	Length_of_Reign	Cause_of_Death
0	Augustus	40.58	Possibly assassinated
1	Tiberius	22.50	Possibly assassinated
2	Caligula	4.83	Assassinated
3	Claudius	13.75	Possibly assassinated
4	Nero	13.67	Suicide

fraction of all Roman Emperors that were assassinated/possibly assassinated





## Question 10

Relationship between revenue from Arcades and the number of Computer Science PhDs between 2000-2009

