

app_my

July 21, 2020

```
[1]: import pandas as pd
```

```
[2]: #Wczytanie pliku csv przy pomocy biblioteki Pandas
flights = pd.read_csv('flight_data_2016.csv')
```

```
[3]: #Sprawdzenie kolumn w danych
flights.columns
```

```
[3]: Index(['QUARTER', 'MONTH', 'DAY_OF_MONTH', 'DAY_OF_WEEK', 'FL_DATE',
'UNIQUE_CARRIER', 'FL_NUM', 'ORIGIN_AIRPORT_ID',
'ORIGIN_AIRPORT_SEQ_ID', 'ORIGIN_CITY_MARKET_ID', 'ORIGIN',
'ORIGIN_CITY_NAME', 'ORIGIN_STATE_ABR', 'ORIGIN_STATE_NM',
'DEST_AIRPORT_ID', 'DEST_AIRPORT_SEQ_ID', 'DEST_CITY_MARKET_ID', 'DEST',
'DEST_CITY_NAME', 'DEST_STATE_ABR', 'DEST_STATE_NM', 'CRS_DEP_TIME',
'DEP_TIME', 'DEP_DELAY', 'DEP_DELAY_NEW', 'WHEELS_ON', 'TAXI_IN',
'CRS_ARR_TIME', 'ARR_TIME', 'ARR_DELAY', 'ARR_DELAY_NEW',
'CRS_ELAPSED_TIME', 'ACTUAL_ELAPSED_TIME', 'AIR_TIME', 'FLIGHTS',
'DISTANCE', 'CARRIER_DELAY', 'WEATHER_DELAY', 'NAS_DELAY',
'SECURITY_DELAY', 'LATE_AIRCRAFT_DELAY', 'FIRST_DEP_TIME',
'Unnamed: 42'],
dtype='object')
```

```
[4]: #Wyświetlenie pierwszych 5 pozycji
flights.head()
```

```
[4]:
```

	QUARTER	MONTH	DAY_OF_MONTH	DAY_OF_WEEK	FL_DATE	UNIQUE_CARRIER	\
0	1	1	3	7	2016-01-03	F9	
1	1	1	3	7	2016-01-03	F9	
2	1	1	3	7	2016-01-03	F9	
3	1	1	3	7	2016-01-03	F9	
4	1	1	3	7	2016-01-03	F9	

	FL_NUM	ORIGIN_AIRPORT_ID	ORIGIN_AIRPORT_SEQ_ID	ORIGIN_CITY_MARKET_ID	\
0	694	11292	1129202	30325	
1	809	14027	1402702	34027	
2	907	15356	1535602	35356	
3	908	14492	1449202	34492	

4 915 15356 1535602 35356

	...	AIR_TIME	FLIGHTS	DISTANCE	CARRIER_DELAY	WEATHER_DELAY	NAS_DELAY	\
0	...	87.0	1.0	692.0	NaN	NaN	NaN	
1	...	224.0	1.0	1679.0	19.0	0.0	0.0	
2	...	60.0	1.0	373.0	NaN	NaN	NaN	
3	...	57.0	1.0	373.0	NaN	NaN	NaN	
4	...	107.0	1.0	693.0	2.0	0.0	18.0	

	SECURITY_DELAY	LATE_AIRCRAFT_DELAY	FIRST_DEP_TIME	Unnamed: 42
0	NaN	NaN	NaN	NaN
1	0.0	0.0	NaN	NaN
2	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN
4	0.0	0.0	NaN	NaN

[5 rows x 43 columns]

```
[5]: #Wyciągnięcie interesujących danych
flights =
    ↳ flights[['QUARTER', 'MONTH', 'DAY_OF_MONTH', 'DAY_OF_WEEK', 'UNIQUE_CARRIER', 'ARR_DELAY', 'ORIGIN_CITY_NAME', 'DEST_CITY_NAME', 'DISTANCE', 'AIR_TIME']]
flights
```

[5]:	QUARTER	MONTH	DAY_OF_MONTH	DAY_OF_WEEK	UNIQUE_CARRIER	ARR_DELAY	\
0	1	1	3	7	F9	-5.0	
1	1	1	3	7	F9	19.0	
2	1	1	3	7	F9	-2.0	
3	1	1	3	7	F9	-5.0	
4	1	1	3	7	F9	20.0	
...	
1856056	4	12	30	5	DL	-5.0	
1856057	4	12	30	5	DL	3.0	
1856058	4	12	30	5	DL	-29.0	
1856059	4	12	30	5	DL	-3.0	
1856060	4	12	30	5	DL	-10.0	

	ORIGIN_CITY_NAME	DEST_CITY_NAME	DISTANCE	\
0	Denver, CO	Cedar Rapids/Iowa City, IA	692.0	
1	West Palm Beach/Palm Beach, FL	Denver, CO	1679.0	
2	Trenton, NJ	Raleigh/Durham, NC	373.0	
3	Raleigh/Durham, NC	Trenton, NJ	373.0	
4	Trenton, NJ	Chicago, IL	693.0	
...	
1856056	Fort Lauderdale, FL	Atlanta, GA	581.0	
1856057	Atlanta, GA	Milwaukee, WI	669.0	
1856058	Milwaukee, WI	Atlanta, GA	669.0	

1856059	Atlanta, GA	Fort Myers, FL	515.0
1856060	Fort Myers, FL	Atlanta, GA	515.0

	AIR_TIME
0	87.0
1	224.0
2	60.0
3	57.0
4	107.0
...	...
1856056	88.0
1856057	105.0
1856058	89.0
1856059	71.0
1856060	85.0

[1856061 rows x 10 columns]

```
[6]: #Sprawdzenie ilosci wartosci NULL w kazdej kolumnie
      flights.isna().sum(axis = 0)
```

```
[6]: QUARTER          0
      MONTH           0
      DAY_OF_MONTH    0
      DAY_OF_WEEK      0
      UNIQUE_CARRIER  0
      ARR_DELAY       31658
      ORIGIN_CITY_NAME  0
      DEST_CITY_NAME   0
      DISTANCE         0
      AIR_TIME        31658
      dtype: int64
```

```
[7]: import numpy as np
```

```
[8]: #Usuniecie wartosci NULL
      flights['ARR_DELAY'].replace('', np.nan, inplace=True)
      flights.dropna(subset=['ARR_DELAY'], inplace=True)

      flights['AIR_TIME'].replace('', np.nan, inplace=True)
      flights.dropna(subset=['AIR_TIME'], inplace=True)
```

```
[9]: #Sprawdzenie wartosci NULL w kazdej kolumnie
      flights.isnull().sum(axis = 0)
```

```
[9]: QUARTER          0
      MONTH           0
```

```

DAY_OF_MONTH      0
DAY_OF_WEEK       0
UNIQUE_CARRIER   0
ARR_DELAY         0
ORIGIN_CITY_NAME   0
DEST_CITY_NAME     0
DISTANCE          0
AIR_TIME          0
dtype: int64

```

```

[10]: #Sprawdzenie ilosci wierszy
count_row = flights.shape[0]
count_row

```

```

[10]: 1824403

```

```

[11]: #dodanie kolumny indeksow na podstawie ilosci wierszy
flights['New_ID'] = range(0, 0 + flights.shape[0])
#Wyciagniecie interesujacych danych w odpowiedniej kolejnosci (indeks na
↳poczatku)
flights =
↳flights[['New_ID', 'QUARTER', 'MONTH', 'DAY_OF_MONTH', 'DAY_OF_WEEK', 'UNIQUE_CARRIER', 'ARR_DELA
↳'DEST_CITY_NAME', 'DISTANCE', 'AIR_TIME']]
flights

```

```

[11]:
      New_ID  QUARTER  MONTH  DAY_OF_MONTH  DAY_OF_WEEK  UNIQUE_CARRIER  \
0           0        1      1             3           7             F9
1           1        1      1             3           7             F9
2           2        1      1             3           7             F9
3           3        1      1             3           7             F9
4           4        1      1             3           7             F9
...         ...     ...     ...           ...         ...
1856056  1824398        4     12            30          5             DL
1856057  1824399        4     12            30          5             DL
1856058  1824400        4     12            30          5             DL
1856059  1824401        4     12            30          5             DL
1856060  1824402        4     12            30          5             DL

      ARR_DELAY      ORIGIN_CITY_NAME  \
0          -5.0          Denver, CO
1          19.0  West Palm Beach/Palm Beach, FL
2          -2.0          Trenton, NJ
3          -5.0      Raleigh/Durham, NC
4          20.0          Trenton, NJ
...         ...         ...
1856056      -5.0      Fort Lauderdale, FL
1856057       3.0          Atlanta, GA

```

1856058	-29.0	Milwaukee, WI
1856059	-3.0	Atlanta, GA
1856060	-10.0	Fort Myers, FL

	DEST_CITY_NAME	DISTANCE	AIR_TIME
0	Cedar Rapids/Iowa City, IA	692.0	87.0
1	Denver, CO	1679.0	224.0
2	Raleigh/Durham, NC	373.0	60.0
3	Trenton, NJ	373.0	57.0
4	Chicago, IL	693.0	107.0
...
1856056	Atlanta, GA	581.0	88.0
1856057	Milwaukee, WI	669.0	105.0
1856058	Atlanta, GA	669.0	89.0
1856059	Fort Myers, FL	515.0	71.0
1856060	Atlanta, GA	515.0	85.0

[1824403 rows x 11 columns]

```
[12]: #Wyliczenie dodatkowo szybkości samolotu
flights['air_speed (mph)'] = flights['DISTANCE'] / (flights['AIR_TIME'] / 60)
flights
```

/Users/monikajanocha/opt/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
[12]:
```

	New_ID	QUARTER	MONTH	DAY_OF_MONTH	DAY_OF_WEEK	UNIQUE_CARRIER	\
0	0	1	1	3	7	F9	
1	1	1	1	3	7	F9	
2	2	1	1	3	7	F9	
3	3	1	1	3	7	F9	
4	4	1	1	3	7	F9	
...	
1856056	1824398	4	12	30	5	DL	
1856057	1824399	4	12	30	5	DL	
1856058	1824400	4	12	30	5	DL	
1856059	1824401	4	12	30	5	DL	
1856060	1824402	4	12	30	5	DL	

	ARR_DELAY	ORIGIN_CITY_NAME	\
0	-5.0	Denver, CO	

```

1          19.0 West Palm Beach/Palm Beach, FL
2          -2.0                      Trenton, NJ
3          -5.0                      Raleigh/Durham, NC
4          20.0                      Trenton, NJ
...
1856056    -5.0                      Fort Lauderdale, FL
1856057     3.0                      Atlanta, GA
1856058   -29.0                      Milwaukee, WI
1856059    -3.0                      Atlanta, GA
1856060   -10.0                      Fort Myers, FL

```

	DEST_CITY_NAME	DISTANCE	AIR_TIME	air_speed (mph)
0	Cedar Rapids/Iowa City, IA	692.0	87.0	477.241379
1	Denver, CO	1679.0	224.0	449.732143
2	Raleigh/Durham, NC	373.0	60.0	373.000000
3	Trenton, NJ	373.0	57.0	392.631579
4	Chicago, IL	693.0	107.0	388.598131
...
1856056	Atlanta, GA	581.0	88.0	396.136364
1856057	Milwaukee, WI	669.0	105.0	382.285714
1856058	Atlanta, GA	669.0	89.0	451.011236
1856059	Fort Myers, FL	515.0	71.0	435.211268
1856060	Atlanta, GA	515.0	85.0	363.529412

[1824403 rows x 12 columns]

```
[13]: #wygenerowanie nowego pliku csv
flights.to_csv('flight_data_2016_nowe.csv', index=False)
```

```
[13]: #Wczytanie pliku csv przy pomocy biblioteki Pandas
flights2 = pd.read_csv('flight_data_2016_nowe.csv', index_col='New_ID')
```

```

/Users/monikajanocha/opt/anaconda3/lib/python3.7/site-
packages/numpy/lib/arraysetops.py:569: FutureWarning: elementwise comparison
failed; returning scalar instead, but in the future will perform elementwise
comparison
    mask |= (ar1 == a)

```

```
[14]: flights2
```

```

[14]:
   QUARTER  MONTH  DAY_OF_MONTH  DAY_OF_WEEK  UNIQUE_CARRIER  ARR_DELAY  \
New_ID
0         1      1             3           7              F9        -5.0
1         1      1             3           7              F9        19.0
2         1      1             3           7              F9        -2.0
3         1      1             3           7              F9        -5.0
4         1      1             3           7              F9        20.0

```

...
1824398	4	12	30	5	DL	-5.0
1824399	4	12	30	5	DL	3.0
1824400	4	12	30	5	DL	-29.0
1824401	4	12	30	5	DL	-3.0
1824402	4	12	30	5	DL	-10.0

	ORIGIN_CITY_NAME	DEST_CITY_NAME	DISTANCE \
New_ID			
0	Denver, CO	Cedar Rapids/Iowa City, IA	692.0
1	West Palm Beach/Palm Beach, FL	Denver, CO	1679.0
2	Trenton, NJ	Raleigh/Durham, NC	373.0
3	Raleigh/Durham, NC	Trenton, NJ	373.0
4	Trenton, NJ	Chicago, IL	693.0
...
1824398	Fort Lauderdale, FL	Atlanta, GA	581.0
1824399	Atlanta, GA	Milwaukee, WI	669.0
1824400	Milwaukee, WI	Atlanta, GA	669.0
1824401	Atlanta, GA	Fort Myers, FL	515.0
1824402	Fort Myers, FL	Atlanta, GA	515.0

	AIR_TIME	air_speed (mph)
New_ID		
0	87.0	477.241379
1	224.0	449.732143
2	60.0	373.000000
3	57.0	392.631579
4	107.0	388.598131
...
1824398	88.0	396.136364
1824399	105.0	382.285714
1824400	89.0	451.011236
1824401	71.0	435.211268
1824402	85.0	363.529412

[1824403 rows x 11 columns]

```
[15]: flights2_okrojone = flights2.iloc[:,4:]
```

```
[16]: flights2_okrojone
```

```
[16]:
```

	UNIQUE_CARRIER	ARR_DELAY	ORIGIN_CITY_NAME \
New_ID			
0	F9	-5.0	Denver, CO
1	F9	19.0	West Palm Beach/Palm Beach, FL
2	F9	-2.0	Trenton, NJ
3	F9	-5.0	Raleigh/Durham, NC

4	F9	20.0	Trenton, NJ
...
1824398	DL	-5.0	Fort Lauderdale, FL
1824399	DL	3.0	Atlanta, GA
1824400	DL	-29.0	Milwaukee, WI
1824401	DL	-3.0	Atlanta, GA
1824402	DL	-10.0	Fort Myers, FL

New_ID	DEST_CITY_NAME	DISTANCE	AIR_TIME	air_speed (mph)
0	Cedar Rapids/Iowa City, IA	692.0	87.0	477.241379
1	Denver, CO	1679.0	224.0	449.732143
2	Raleigh/Durham, NC	373.0	60.0	373.000000
3	Trenton, NJ	373.0	57.0	392.631579
4	Chicago, IL	693.0	107.0	388.598131
...
1824398	Atlanta, GA	581.0	88.0	396.136364
1824399	Milwaukee, WI	669.0	105.0	382.285714
1824400	Atlanta, GA	669.0	89.0	451.011236
1824401	Fort Myers, FL	515.0	71.0	435.211268
1824402	Atlanta, GA	515.0	85.0	363.529412

[1824403 rows x 7 columns]

```
[17]: #Wyliczenie jakie sa wartosci statystyczne dla kazdej z linii lotniczej o
      ↪konkretnej trasie
flights2_grouped = flights2_okrojone.
      ↪groupby(['UNIQUE_CARRIER', 'ORIGIN_CITY_NAME', 'DEST_CITY_NAME']).
      ↪agg(['count', 'mean', 'min', 'max']).reset_index()
flights2_grouped
```

```
[17]:
```

	UNIQUE_CARRIER	ORIGIN_CITY_NAME	DEST_CITY_NAME	ARR_DELAY \
				count
0	AA	Albany, NY	Charlotte, NC	324
1	AA	Albuquerque, NM	Dallas/Fort Worth, TX	533
2	AA	Amarillo, TX	Dallas/Fort Worth, TX	56
3	AA	Anchorage, AK	Dallas/Fort Worth, TX	30
4	AA	Anchorage, AK	Los Angeles, CA	22
...
7417	WN	Wichita, KS	Chicago, IL	53
7418	WN	Wichita, KS	Dallas, TX	57
7419	WN	Wichita, KS	Las Vegas, NV	122
7420	WN	Wichita, KS	Phoenix, AZ	91
7421	WN	Wichita, KS	St. Louis, MO	171

			DISTANCE				AIR_TIME \
	mean	min	max	count	mean	min	max
				count			count

0	-7.904321	-49.0	206.0	324	646.0	646.0	646.0	324
1	6.744841	-35.0	1260.0	533	569.0	569.0	569.0	533
2	-1.625000	-31.0	130.0	56	312.0	312.0	312.0	56
3	114.733333	-26.0	1295.0	30	3043.0	3043.0	3043.0	30
4	-33.318182	-54.0	43.0	22	2345.0	2345.0	2345.0	22
...
7417	4.320755	-21.0	178.0	53	589.0	589.0	589.0	53
7418	4.245614	-21.0	313.0	57	333.0	333.0	333.0	57
7419	-6.229508	-43.0	298.0	122	986.0	986.0	986.0	122
7420	-7.494505	-45.0	108.0	91	870.0	870.0	870.0	91
7421	-2.192982	-31.0	325.0	171	392.0	392.0	392.0	171

	air_speed (mph)				\	
	mean	min	max	count	mean	min
0	106.395062	87.0	138.0	324	367.229228	280.869565
1	77.645403	67.0	112.0	533	441.701742	304.821429
2	49.071429	42.0	67.0	56	384.090308	279.402985
3	351.800000	332.0	375.0	30	519.408818	486.880000
4	277.136364	264.0	292.0	22	508.006802	481.849315
...
7417	82.679245	75.0	103.0	53	428.576997	343.106796
7418	56.368421	50.0	63.0	57	355.535563	317.142857
7419	138.237705	115.0	166.0	122	430.175175	356.385542
7420	126.340659	103.0	147.0	91	415.328312	355.102041
7421	54.099415	47.0	93.0	171	437.516893	252.903226

	max
0	445.517241
1	509.552239
2	445.714286
3	549.939759
4	532.954545
...	...
7417	471.200000
7418	399.600000
7419	514.434783
7420	506.796117
7421	500.425532

[7422 rows x 19 columns]

```
[18]: carrier_stats2 = flights2_okrojone.groupby('UNIQUE_CARRIER')['air_speed (mph)'].
      ↳describe().reset_index().rename(columns={'UNIQUE_CARRIER': 'airline',
      ↳'count': 'number_of_flights', '50%': 'median'})
      carrier_stats2
```

```
[18]:
```

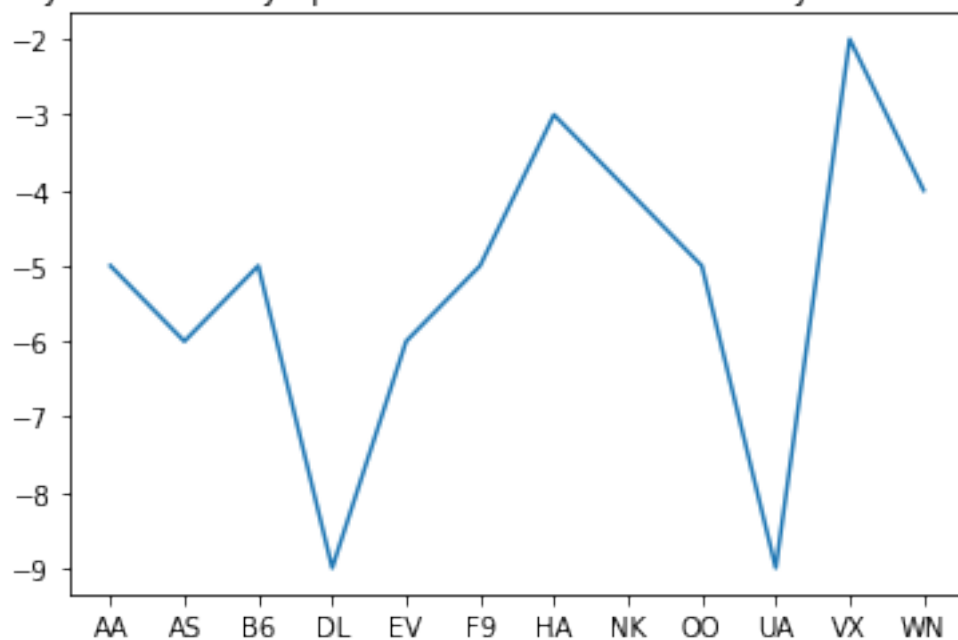
	airline	number_of_flights	mean	std	min	25%	\
0	AA	295833.0	423.912774	68.590616	70.751445	384.761905	
1	AS	58543.0	437.894440	62.692879	89.062500	409.606299	
2	B6	93313.0	420.127606	72.145404	97.500000	379.487179	
3	DL	296594.0	417.086086	64.261109	115.675676	377.647059	
4	EV	154224.0	361.223625	66.158543	50.308530	318.260870	
5	F9	31236.0	448.124893	49.197357	165.397590	416.097561	
6	HA	25683.0	341.098756	88.909812	149.268293	272.727273	
7	NK	46196.0	435.862743	54.175735	131.111111	404.457831	
8	OO	194789.0	367.814386	74.208374	70.985915	322.857143	
9	UA	178621.0	444.691363	64.347271	95.714286	407.272727	
10	VX	23063.0	442.336155	65.358191	190.754717	392.701422	
11	WN	426308.0	415.021506	59.775816	77.513514	374.444444	

	median	75%	max
0	429.310345	469.629630	760.000000
1	440.571429	474.279221	762.105263
2	432.857143	467.213115	628.358209
3	419.318182	458.365385	628.369565
4	365.357143	407.899160	731.250000
5	445.263158	478.656716	619.072848
6	322.105263	370.285714	616.267606
7	439.800000	471.190476	674.838710
8	375.428571	420.000000	726.000000
9	447.532468	488.000000	647.088608
10	440.597015	493.214286	625.603448
11	414.827586	455.368421	786.346154

```
[20]: import matplotlib.pyplot as plt
```

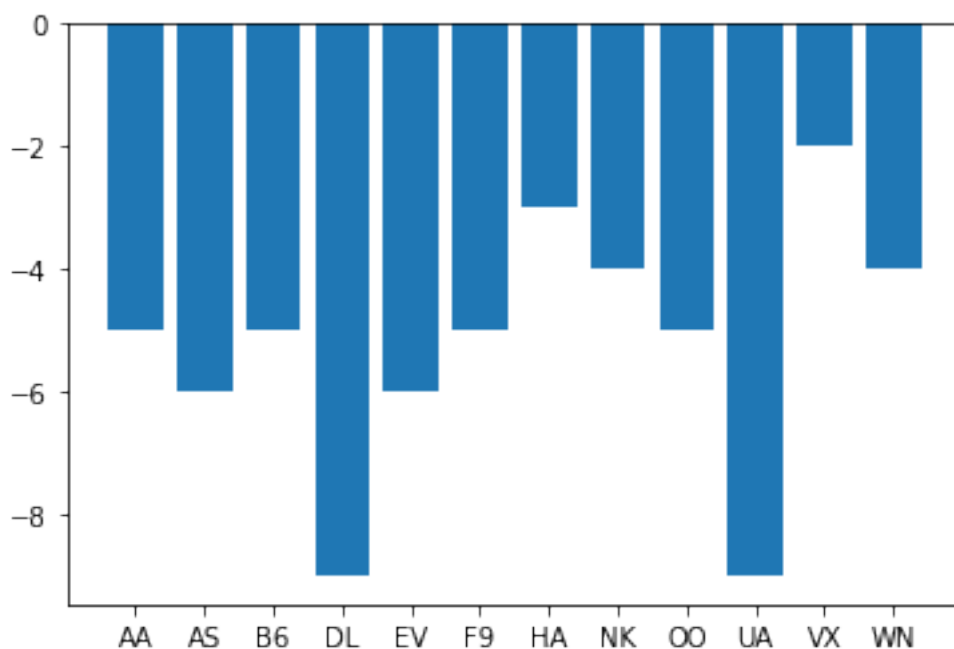
```
[21]: x = carrier_stats['airline']
y = carrier_stats['median']
plt.plot(x, y)
plt.title('Wykres mediany opóźnień w zależności od nazwy linii lotniczej')
plt.show()
```

Wykres mediany opóźnień w zależności od nazwy linii lotniczej



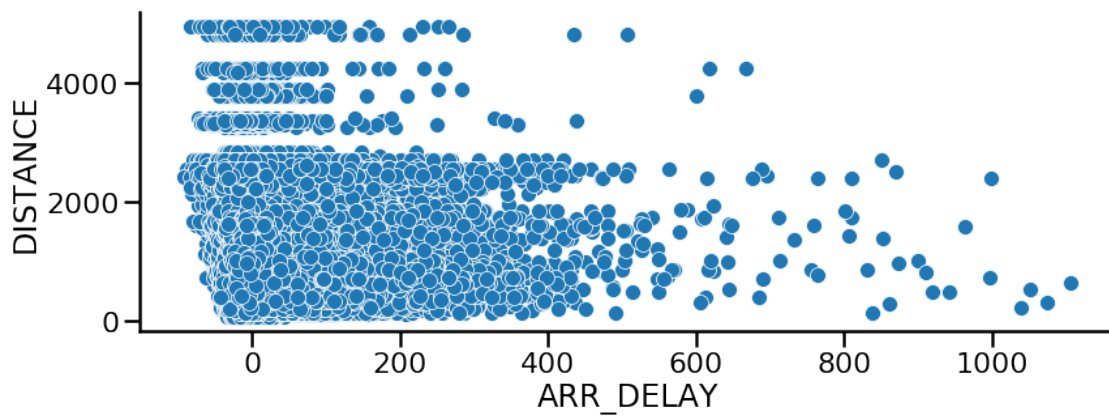
```
[22]: x=carrier_stats['airline']  
y=carrier_stats['median']  
plt.bar(x, y)
```

[22]: <BarContainer object of 12 artists>

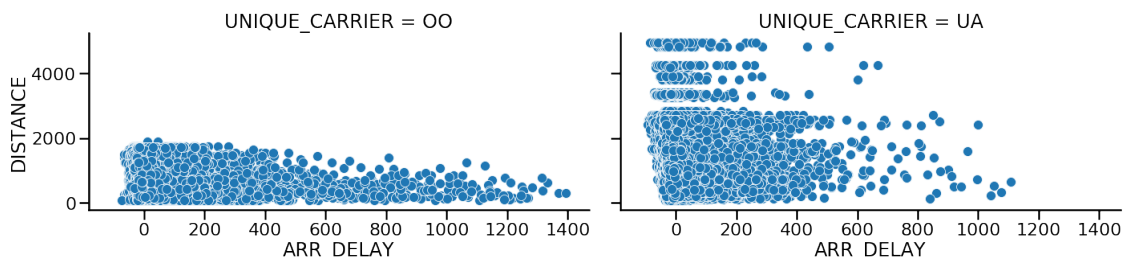


```
[23]: import seaborn as sns
```

```
[24]: sns.set_context('poster')
sns.relplot(x="ARR_DELAY",
            y="DISTANCE",
            aspect=2.5,
            data=flights[flights['UNIQUE_CARRIER']=='UA'],
            kind="scatter");
plt.show()
```



```
[27]: sns.set_context('poster')
sns.relplot(data=flights[(flights['UNIQUE_CARRIER'] == 'OO') |
↪ (flights['UNIQUE_CARRIER'] == 'UA') ],
            x="ARR_DELAY",
            y="DISTANCE",
            aspect=2,
            kind="scatter",
            col='UNIQUE_CARRIER')
plt.show()
```



```
[26]: sns.set_context('paper')
sns.catplot(x="UNIQUE_CARRIER",
            data=flights,
            aspect=2.5,
            kind='count')
plt.show()
```

