

Wrangling

May 22, 2018

0.1 Tasks for Step 4

In [1]: *# TODO: re-load dataframes from SQL*

```
import csv
import pandas as pd
import numpy as np
import sqlite3
import matplotlib.pyplot as plt
import sklearn.metrics
import matplotlib.pyplot as plt; plt.rcParamsDefaults()
engine=sqlite3.connect('HW1_DB')
airlines_df=pd.read_sql('select* from airlines', engine)
airports_df=pd.read_sql('select* from airports', engine)
routes_df=pd.read_sql('select* from routes', engine)
incidents_df=pd.read_sql('select* from incidents', engine)
```

In [2]: *# TODO: airports_df.info(), airlines_df.info(), routes_df.info() and incidents_df.info()*

```
airports_df.info()
airlines_df.info()
routes_df.info()
incidents_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8107 entries, 0 to 8106
Data columns (total 12 columns):
airport_id      8107 non-null int64
airport_name    8107 non-null object
city            8107 non-null object
country         8107 non-null object
iata_ffa        8107 non-null object
icao             8107 non-null object
latitude        8107 non-null float64
longitude       8107 non-null float64
altitude        8107 non-null int64
timezone        8107 non-null float64
dst             8107 non-null object
tz             8107 non-null object
dtypes: float64(3), int64(2), object(7)
```

```

memory usage: 760.1+ KB
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6048 entries, 0 to 6047
Data columns (total 8 columns):
airline_id      6048 non-null int64
airline_name    6048 non-null object
airline_alias   6048 non-null object
iata            6048 non-null object
icao            6048 non-null object
airline_callsign 6048 non-null object
country        6048 non-null object
active         6048 non-null object
dtypes: int64(1), object(7)
memory usage: 378.1+ KB
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 66548 entries, 0 to 66547
Data columns (total 9 columns):
airline_iata_icao 66548 non-null object
airline_id       66548 non-null int64
source_airport_iata_icao 66548 non-null object
source_airport_id 66548 non-null int64
destination_iata_icao 66548 non-null object
destination_airport_id 66548 non-null int64
codeshare        66548 non-null object
number_of_stops  66548 non-null int64
Equipments       66548 non-null object
dtypes: int64(4), object(5)
memory usage: 4.6+ MB
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 322 entries, 0 to 321
Data columns (total 3 columns):
Airline      322 non-null object
Date         322 non-null object
FlightNum    322 non-null object
dtypes: object(3)
memory usage: 7.6+ KB

```

0.2 Step 4.0

```

In [3]: #TODO: output airports_df
        airports_df

```

```

Out[3]:

```

	airport_id	airport_name	city \
0	1	Goroka	Goroka
1	2	Madang	Madang
2	3	Mount Hagen	Mount Hagen
3	4	Nadzab	Nadzab

4	5	Port Moresby Jacksons Intl	Port Moresby
5	6	Wewak Intl	Wewak
6	7	Narsarsuaq	Narssarsuaq
7	8	Nuuk	Godthaab
8	9	Sondre Stromfjord	Sondrestrom
9	10	Thule Air Base	Thule
10	11	Akureyri	Akureyri
11	12	Egilsstadir	Egilsstadir
12	13	Hornafjordur	Hofn
13	14	Husavik	Husavik
14	15	Isafjordur	Isafjordur
15	16	Keflavik International Airport	Keflavik
16	17	Patreksfjordur	Patreksfjordur
17	18	Reykjavik	Reykjavik
18	19	Siglufjordur	Siglufjordur
19	20	Vestmannaeyjar	Vestmannaeyjar
20	21	Sault Ste Marie	Sault Sainte Marie
21	22	Winnipeg St Andrews	Winnipeg
22	23	Shearwater	Halifax
23	24	St Anthony	St. Anthony
24	25	Tofino	Tofino
25	26	Kugaaruk	Pelly Bay
26	27	Baie Comeau	Baie Comeau
27	28	Bagotville	Bagotville
28	29	Baker Lake	Baker Lake
29	30	Campbell River	Campbell River
...
8077	9512	Port of Seattle	Seattle
8078	9513	Ferry Terminal	Whittier
8079	9514	Ferry Terminal	Valdez
8080	9515	Gare Routiere	Saint-Louis
8081	9516	Bus Stop	La Riviere
8082	9517	Bus Stop	Cilaos
8083	9518	Bus Stop	Hell-Bourg
8084	9519	Bus Stop	Salazie
8085	9520	Gare	St Andre
8086	9521	L'Ocean Bus Terminal	Saint Denis
8087	9522	Gare Routiere	Saint Pierre
8088	9523	Baie Ste Anne	Praslin
8089	9524	La Passe	La Digue
8090	9525	Ferry Dock	Bainbridge Island
8091	9526	Bus	Phnom Penh
8092	9527	Bus	Siem Reap
8093	9528	Bus	Sihanoukville
8094	9529	Bus	Kampot
8095	9530	Bus	Kep
8096	9531	Main Station	Taipei
8097	9532	Prominent Hill	Prominent Hill

8098	9533	Chatsworth Station	Chatsworth
8099	9534	Algerciras Port	Algerciras
8100	9535	Ganges Water Aerodrome	Ganges
8101	9536	Pender Harbour Water Aerodrome	Pender Harbour
8102	9537	Mansons Landing Water Aerodrome	Mansons Landing
8103	9538	Port McNeill Airport	Port McNeill
8104	9539	Sullivan Bay Water Aerodrome	Sullivan Bay
8105	9540	Deer Harbor Seaplane	Deer Harbor
8106	9541	San Diego Old Town Transit Center	San Diego

	country	iata_ffa	icao	latitude	longitude	altitude	\
0	Papua New Guinea	GKA	AYGA	-6.081689	145.391881	5282	
1	Papua New Guinea	MAG	AYMD	-5.207083	145.788700	20	
2	Papua New Guinea	HGU	AYMH	-5.826789	144.295861	5388	
3	Papua New Guinea	LAE	AYNZ	-6.569828	146.726242	239	
4	Papua New Guinea	POM	AYPY	-9.443383	147.220050	146	
5	Papua New Guinea	WWK	AYWK	-3.583828	143.669186	19	
6	Greenland	UAK	BGBW	61.160517	-45.425978	112	
7	Greenland	GOH	BGGH	64.190922	-51.678064	283	
8	Greenland	SFJ	BGSF	67.016969	-50.689325	165	
9	Greenland	THU	BGTL	76.531203	-68.703161	251	
10	Iceland	AEY	BIAR	65.659994	-18.072703	6	
11	Iceland	EGS	BIEG	65.283333	-14.401389	76	
12	Iceland	HFN	BIHN	64.295556	-15.227222	24	
13	Iceland	HZK	BIHU	65.952328	-17.425978	48	
14	Iceland	IFJ	BIIS	66.058056	-23.135278	8	
15	Iceland	KEF	BIKF	63.985000	-22.605556	171	
16	Iceland	PFJ	BIPA	65.555833	-23.965000	11	
17	Iceland	RKV	BIRK	64.130000	-21.940556	48	
18	Iceland	SIJ	BISI	66.133333	-18.916667	10	
19	Iceland	VEY	BIVM	63.424303	-20.278875	326	
20	Canada	YAM	CYAM	46.485001	-84.509445	630	
21	Canada	YAV	CYAV	50.056389	-97.032500	760	
22	Canada	YAW	CYAW	44.639721	-63.499444	167	
23	Canada	YAY	CYAY	51.391944	-56.083056	108	
24	Canada	YAZ	CYAZ	49.082222	-125.772500	80	
25	Canada	YBB	CYBB	68.534444	-89.808056	56	
26	Canada	YBC	CYBC	49.132500	-68.204444	71	
27	Canada	YBG	CYBG	48.330555	-70.996391	522	
28	Canada	YBK	CYBK	64.298889	-96.077778	59	
29	Canada	YBL	CYBL	49.950832	-125.270833	346	
...	
8077	United States		PSEA	47.609178	-122.350605	0	
8078	United States		WHIT	60.776566	-148.683367	0	
8079	United States		VLDZ	61.123976	-146.365309	0	
8080	Reunion		STLS	-21.289604	55.407269	103	
8081	Reunion		LRIV	-21.272574	55.437706	190	
8082	Reunion		CILS	-21.135518	55.472087	4000	

8083	Reunion	HLBG	-21.064020	55.518453	3000	
8084	Reunion	SLZI	-21.027485	55.538955	3000	
8085	Reunion	ANDR	-20.963272	55.652508	100	
8086	Reunion	SDNS	-20.877683	55.457393	0	
8087	Reunion	STPR	-21.334240	55.471331	0	
8088	Seychelles	PLIN	-4.347119	55.765688	0	
8089	Seychelles	LADG	-4.347969	55.829069	0	
8090	United States	BAIN	47.622237	-122.509362	0	
8091	Cambodia	PNMP	11.561716	104.914276	0	
8092	Cambodia	SMRP	13.361002	103.859543	0	
8093	Cambodia	SNKV	10.607220	103.524886	0	
8094	Cambodia	KMPT	10.614922	104.177724	0	
8095	Cambodia	KEPC	10.480174	104.294228	0	
8096	Taiwan	TPEI	25.046176	121.517532	0	
8097	Australia	PXH	YPMH	-29.716667	135.521667	734
8098	United States	CWT		34.256944	-118.598889	978
8099	Spain			36.136000	-5.435000	0
8100	Canada	YGG		48.850000	-123.500000	0
8101	Canada	YPT		49.616667	-124.016667	0
8102	Canada	YMU		50.066667	-124.983333	0
8103	Canada	YMP		50.575556	-127.028611	225
8104	Canada	YTG		50.883333	-126.833333	0
8105	United States	DHB		48.618397	-123.005960	0
8106	United States	OLT		32.755200	-117.199500	0

	timezone	dst	tz
0	10.0	U	Pacific/Port_Moresby
1	10.0	U	Pacific/Port_Moresby
2	10.0	U	Pacific/Port_Moresby
3	10.0	U	Pacific/Port_Moresby
4	10.0	U	Pacific/Port_Moresby
5	10.0	U	Pacific/Port_Moresby
6	-3.0	E	America/Godthab
7	-3.0	E	America/Godthab
8	-3.0	E	America/Godthab
9	-4.0	E	America/Thule
10	0.0	N	Atlantic/Reykjavik
11	0.0	N	Atlantic/Reykjavik
12	0.0	N	Atlantic/Reykjavik
13	0.0	N	Atlantic/Reykjavik
14	0.0	N	Atlantic/Reykjavik
15	0.0	N	Atlantic/Reykjavik
16	0.0	N	Atlantic/Reykjavik
17	0.0	N	Atlantic/Reykjavik
18	0.0	N	Atlantic/Reykjavik
19	0.0	N	Atlantic/Reykjavik
20	-5.0	A	America/Toronto
21	-6.0	A	America/Winnipeg

22	-4.0	A	America/Halifax
23	-3.5	A	America/St_Johns
24	-8.0	A	America/Vancouver
25	-7.0	A	America/Edmonton
26	-5.0	A	America/Toronto
27	-5.0	A	America/Toronto
28	-6.0	A	America/Winnipeg
29	-8.0	A	America/Vancouver
...
8077	-8.0	A	America/Los_Angeles
8078	-9.0	A	America/Anchorage
8079	-9.0	A	America/Anchorage
8080	4.0	E	Indian/Reunion
8081	4.0	E	Indian/Reunion
8082	4.0	E	Indian/Reunion
8083	4.0	E	Indian/Reunion
8084	4.0	E	Indian/Reunion
8085	4.0	E	Indian/Reunion
8086	4.0	E	Indian/Reunion
8087	4.0	E	Indian/Reunion
8088	4.0	N	Indian/Mahe
8089	4.0	N	Indian/Mahe
8090	-8.0	A	America/Los_Angeles
8091	7.0	N	Asia/Phnom_Penh
8092	7.0	N	Asia/Phnom_Penh
8093	7.0	N	Asia/Phnom_Penh
8094	7.0	N	Asia/Phnom_Penh
8095	7.0	N	Asia/Phnom_Penh
8096	8.0	N	Asia/Taipei
8097	9.5	O	Australia/Adelaide
8098	-8.0	A	America/Los_Angeles
8099	1.0	E	Europe/Madrid
8100	-8.0	A	America/Vancouver
8101	-8.0	A	America/Vancouver
8102	-8.0	A	America/Vancouver
8103	-8.0	A	America/Vancouver
8104	-8.0	A	America/Vancouver
8105	-8.0	A	America/Los_Angeles
8106	-8.0	A	America/Los_Angeles

[8107 rows x 12 columns]

```
In [4]: #TODO: output airports_df.describe()
airports_df.describe()
```

```
Out[4]:
```

	airport_id	latitude	longitude	altitude	timezone
count	8107.000000	8107.000000	8107.000000	8107.000000	8107.000000
mean	4766.361046	26.817720	-3.921969	933.449365	0.169236

std	2943.205193	27.866953	85.900873	1624.740899	5.737326
min	1.000000	-89.999997	-179.877000	-1266.000000	-12.000000
25%	2091.500000	8.824928	-79.022498	38.000000	-5.000000
50%	4257.000000	34.987800	5.292028	272.000000	1.000000
75%	7508.500000	47.957599	49.785821	1020.000000	4.000000
max	9541.000000	82.517778	179.951000	14472.000000	13.000000

```
In [5]: #TODO: output airlines_df
airlines_df
```

```
Out[5]:
```

	airline_id	airline_name \
0	1	Private flight
1	2	135 Airways
2	3	1Time Airline
3	4	2 Sqn No 1 Elementary Flying Training School
4	5	213 Flight Unit
5	6	223 Flight Unit State Airline
6	7	224th Flight Unit
7	8	247 Jet Ltd
8	9	3D Aviation
9	10	40-Mile Air
10	11	4D Air
11	12	611897 Alberta Limited
12	13	Ansett Australia
13	14	Abacus International
14	15	Abelag Aviation
15	16	Army Air Corps
16	17	Aero Aviation Centre Ltd.
17	18	Aero Servicios Ejecutivos Internacionales
18	19	Aero Biniza
19	20	Aero Albatros
20	21	Aigle Azur
21	22	Aloha Airlines
22	23	Alaska Island Air
23	24	American Airlines
24	25	Aviation Management Corporation
25	26	Atlantis Airlines (USA)
26	27	Aerovista Airlines
27	28	Asiana Airlines
28	29	Askari Aviation
29	30	Australia Asia Airlines
...
6018	19651	CARICOM AIRWAYS (BARBADOS) INC.
6019	19674	Rainbow Air (RAI)
6020	19675	Rainbow Air Canada
6021	19676	Rainbow Air Polynesia
6022	19677	Rainbow Air Euro
6023	19678	Rainbow Air US

6024	19745	Transilvania
6025	19751	Dobrolet
6026	19774	Spike Airlines
6027	19776	Grand Cru Airlines
6028	19785	Go2Sky
6029	19803	All Argentina
6030	19804	All America
6031	19805	All Asia
6032	19806	All Africa
6033	19807	Regionalia México
6034	19808	All Europe
6035	19809	All Spain
6036	19810	Regional Air Iceland
6037	19811	British Air Ferries
6038	19812	Voestar
6039	19813	All Colombia
6040	19814	Regionalia Uruguay
6041	19815	Regionalia Venezuela
6042	19827	Regionalia Chile
6043	19828	Vuela Cuba
6044	19830	All Australia
6045	19831	Fly Europa
6046	19834	FlyPortugal
6047	19845	FTI Fluggesellschaft

	airline_alias	iata	icao	airline_callsign	country \
0			-		
1			GNL	GENERAL	United States
2	1T	RNX		NEXTIME	South Africa
3		WYT			United Kingdom
4		TFU			Russia
5		CHD		CHKALOVSK-AVIA	Russia
6		TTF		CARGO UNIT	Russia
7		TWF		CLOUD RUNNER	United Kingdom
8		SEC		SECUREX	United States
9	Q5	MLA		MILE-AIR	United States
10		QRT		QUARTET	Thailand
11		THD		DONUT	Canada
12	AN	AAA		ANSETT	Australia
13	1B				Singapore
14	W9	AAB		ABG	Belgium
15		AAC		ARMYAIR	United Kingdom
16		AAD		SUNRISE	Canada
17		SII		ASEISA	Mexico
18		BZS		BINIZA	Mexico
19		ABM		ALBATROS ESPANA	Spain
20	ZI	AAF		AIGLE AZUR	France
21	AQ	AAH		ALOHA	United States

22		AAK	ALASKA ISLAND	United States
23		AA AAL	AMERICAN	United States
24		AAM	AM CORP	United States
25		AAO	ATLANTIS AIR	United States
26		AAP	AEROVISTA GROUP	United Arab Emirates
27		OZ AAR	ASIANA	Republic of Korea
28		4K AAS	AL-AAS	Pakistan
29		AAU	AUSTASIA	Australia
...	
6018	CARICOM AIRWAYS	CCB		Barbados
6019	Rainbow Air (RAI)	RN RAB	Rainbow	United States
6020	Rainbow Air CAN	RY RAY	Rainbow CAN	Canada
6021	Rainbow Air POL	RX RPO	Rainbow Air	United States
6022	Rainbow Air EU	RU RUE	Rainbow Air	United Kingdom
6023	Rainbow Air US	RM RNY	Rainbow Air	United States
6024		TNS		Romania
6025	QD DOB		DOBROLET	Russia
6026	Aero Spike	SO SAL	Spike Air	United States
6027		GCA		Lithuania
6028		RLX	RELAX	Slovakia
6029	All Argentina	L1 AL1		Argentina
6030	All America	A2 AL2		United States
6031	All Asia	L9 AL3		China
6032	All Africa	9A 99F		South Africa
6033	Regionalia México	N4 J88		Mexico
6034	All Europe	N9 N99		United Kingdom
6035	All Spain	N7 N77		Spain
6036	Regional Air Iceland	9N N78		Iceland
6037		?? ??!		United Kingdom
6038	Voestar Brasil	8K K88		Brazil
6039	All Colombia	70 7KK		Colombia
6040	Regionalia Uruguay	2X 2K2		Uruguay
6041	Regionalia Venezuela	9X 9XX		Venezuela
6042	Regionalia Chile	9J CR1		Chile
6043	Vuela Cuba	6C 6CC		Cuba
6044	All Australia	88 8K8		Australia
6045		ER RWW		Spain
6046		PO FPT	FlyPortugal	Portugal
6047		FTI		Germany

active

0	Y
1	N
2	Y
3	N
4	N
5	N
6	N

7	N
8	N
9	Y
10	N
11	N
12	Y
13	Y
14	N
15	N
16	N
17	N
18	N
19	N
20	Y
21	Y
22	N
23	Y
24	N
25	N
26	N
27	Y
28	Y
29	N
...	...
6018	Y
6019	Y
6020	Y
6021	Y
6022	Y
6023	Y
6024	Y
6025	Y
6026	Y
6027	Y
6028	Y
6029	Y
6030	Y
6031	Y
6032	Y
6033	Y
6034	Y
6035	Y
6036	Y
6037	N
6038	Y
6039	Y
6040	Y
6041	Y

```

6042      Y
6043      Y
6044      Y
6045      Y
6046      Y
6047      N

```

```
[6048 rows x 8 columns]
```

```
In [6]: #TODO: output airlines_df.describe()
airlines_df.describe()
```

```
Out[6]:      airline_id
count    6048.000000
mean      3844.729167
std       3945.495487
min        -1.000000
25%       1512.750000
50%       3025.500000
75%       4543.250000
max      19845.000000

```

```
In [7]: #TODO: output routes_df
routes_df
```

```
Out[7]:      airline_iata_icao  airline_id  source_airport_iata_icao  \
0                2B           410                AER
1                2B           410                ASF
2                2B           410                ASF
3                2B           410                CEK
4                2B           410                CEK
5                2B           410                DME
6                2B           410                DME
7                2B           410                DME
8                2B           410                EGO
9                2B           410                EGO
10               2B           410                GYD
11               2B           410                KGD
12               2B           410                KZN
13               2B           410                KZN
14               2B           410                KZN
15               2B           410                KZN
16               2B           410                KZN
17               2B           410                KZN
18               2B           410                KZN
19               2B           410                LED
20               2B           410                LED
21               2B           410                LED
22               2B           410                MRV

```

23	2B	410	NBC
24	2B	410	NBC
25	2B	410	NBC
26	2B	410	NBC
27	2B	410	NJC
28	2B	410	NJC
29	2B	410	NUX
...
66518	ZL	4178	SYD
66519	ZL	4178	SYD
66520	ZL	4178	SYD
66521	ZL	4178	SYD
66522	ZL	4178	SYD
66523	ZL	4178	SYD
66524	ZL	4178	SYD
66525	ZL	4178	SYD
66526	ZL	4178	SYD
66527	ZL	4178	SYD
66528	ZL	4178	SYD
66529	ZL	4178	SYD
66530	ZL	4178	SYD
66531	ZL	4178	SYD
66532	ZL	4178	SYD
66533	ZL	4178	SYD
66534	ZL	4178	SYD
66535	ZL	4178	TRO
66536	ZL	4178	TRO
66537	ZL	4178	TSV
66538	ZL	4178	TSV
66539	ZL	4178	WGA
66540	ZL	4178	WGA
66541	ZL	4178	WIN
66542	ZL	4178	WIN
66543	ZL	4178	WYA
66544	ZM	19016	DME
66545	ZM	19016	FRU
66546	ZM	19016	FRU
66547	ZM	19016	OSS

	source_airport_id	destination_iata_icao	destination_airport_id \
0	2965	KZN	2990
1	2966	KZN	2990
2	2966	MRV	2962
3	2968	KZN	2990
4	2968	OVB	4078
5	4029	KZN	2990
6	4029	NBC	6969
7	4029	UUA	6160

8	6156	KGD	2952
9	6156	KZN	2990
10	2922	NBC	6969
11	2952	EGO	6156
12	2990	AER	2965
13	2990	ASF	2966
14	2990	CEK	2968
15	2990	DME	4029
16	2990	EGO	6156
17	2990	LED	2948
18	2990	SVX	2975
19	2948	KZN	2990
20	2948	NBC	6969
21	2948	UUA	6160
22	2962	ASF	2966
23	6969	DME	4029
24	6969	GYD	2922
25	6969	LED	2948
26	6969	SVX	2975
27	2972	SVX	2975
28	2972	UUA	6160
29	4364	SVX	2975
...
66518	3361	ABX	3333
66519	3361	ARM	4062
66520	3361	BHQ	6241
66521	3361	BHS	6249
66522	3361	BNK	4291
66523	3361	DBO	3358
66524	3361	GFF	6271
66525	3361	LSY	6287
66526	3361	MIM	6294
66527	3361	MQL	6298
66528	3361	MYA	6302
66529	3361	NRA	6307
66530	3361	NTL	4320
66531	3361	OAG	6793
66532	3361	PKE	6317
66533	3361	TRO	6794
66534	3361	WGA	3363
66535	6794	GFN	6792
66536	6794	SYD	3361
66537	3330	ISA	3324
66538	3330	WIN	6337
66539	3363	MEL	3339
66540	3363	SYD	3361
66541	6337	LRE	6289
66542	6337	TSV	3330

66543	6334	ADL	3341
66544	4029	FRU	2912
66545	2912	DME	4029
66546	2912	OSS	2913
66547	2913	FRU	2912

	codeshare	number_of_stops	Equipments
0		0	CR2
1		0	CR2
2		0	CR2
3		0	CR2
4		0	CR2
5		0	CR2
6		0	CR2
7		0	CR2
8		0	CR2
9		0	CR2
10		0	CR2
11		0	CR2
12		0	CR2
13		0	CR2
14		0	CR2
15		0	CR2
16		0	CR2
17		0	CR2
18		0	CR2
19		0	CR2
20		0	CR2
21		0	CR2
22		0	CR2
23		0	CR2
24		0	CR2
25		0	CR2
26		0	CR2
27		0	CR2
28		0	CR2
29		0	CR2
...
66518		0	SF3
66519		0	SF3
66520		0	SF3
66521		0	SF3
66522		0	SF3
66523		0	SF3
66524		0	SF3
66525		0	SF3
66526		0	SF3
66527		0	SF3

66528	0	SF3
66529	0	SF3
66530	0	SF3
66531	0	SF3
66532	0	SF3
66533	0	SF3
66534	0	SF3
66535	0	SF3
66536	0	SF3
66537	0	SF3
66538	0	SF3
66539	0	SF3
66540	0	SF3
66541	0	SF3
66542	0	SF3
66543	0	SF3
66544	0	734
66545	0	734
66546	0	734
66547	0	734

[66548 rows x 9 columns]

```
In [8]: #TODO: output routes_df.describe()
routes_df.describe()
```

```
Out[8]:
```

	airline_id	source_airport_id	destination_airport_id	\
count	66548.000000	66548.000000	66548.000000	
mean	3367.569754	2660.201929	2662.088192	
std	3182.024330	1577.965637	1578.661433	
min	10.000000	1.000000	1.000000	
25%	1317.000000	1382.000000	1382.000000	
50%	2942.000000	2958.000000	2958.000000	
75%	4608.000000	3670.000000	3670.000000	
max	19676.000000	9327.000000	9327.000000	

	number_of_stops
count	66548.000000
mean	0.000165
std	0.012856
min	0.000000
25%	0.000000
50%	0.000000
75%	0.000000
max	1.000000

```
In [9]: #TODO: output incidents_df
incidents_df
```

```

Out[9]:
      Airline      Date FlightNum
0      Comair  1997-01-09 00:00:00      3272
1  Stavropolskaya Aktsionernaya Avia  1997-03-18 00:00:00      1023
2      Merpati Nusantara Airlines  1997-04-19 00:00:00       106
3      China Southern Airlines  1997-05-08 00:00:00     3456
4      FedEx Express  1997-07-31 00:00:00       14
5      Sempati Air  1997-07-17 00:00:00      304
6      Korean Air  1997-08-06 00:00:00      801
7      Formosa Airlines  1997-08-10 00:00:00     7601
8      Vietnam Airlines  1997-09-03 00:00:00      815
9      Royal Brunei Airlines  1997-09-06 00:00:00      238
10     Garuda Indonesia  1997-09-26 00:00:00      152
11     Austral Líneas Aéreas  1997-10-10 00:00:00     2553
12     Tajikistan Airlines  1997-12-15 00:00:00     3183
13     Aerosvit  1997-12-17 00:00:00      241
14     SilkAir  1997-12-19 00:00:00      185
15     United Airlines  1997-12-28 00:00:00      826
16     Cebu Pacific  1998-02-02 00:00:00      387
17     China Airlines  1998-02-16 00:00:00      676
18     Philippine Airlines  1998-03-22 00:00:00      137
19     Air France  1998-04-20 00:00:00      422
20     PIA  1998-05-25 00:00:00      544
21     Propair  1998-06-18 00:00:00      420
22     Proteus Airlines  1998-07-30 00:00:00     706:
23     Myanma Airways  1998-08-24 00:00:00      635
24     Cubana de Aviación  1998-08-29 00:00:00      389
25     Swissair  1998-09-02 00:00:00      111
26     PauknAir  1998-09-25 00:00:00     4101
27     Lionair  1998-09-29 00:00:00      602
28     Thai Airways International  1998-12-11 00:00:00      261
29     China Southwest Airlines  1999-02-24 00:00:00     4509
..     ...
292    Ceiba Intercontinental Airlines  2015-09-05 00:00:00       71
293     British Airways  2015-09-08 00:00:00     2276
294     AviaStar  2015-10-02 00:00:00     7503
295     Dynamic Airways  2015-10-29 00:00:00      405
296     Metrojet  2015-10-31 00:00:00     9268
297     West Air Sweden  2016-01-08 00:00:00      294
298     Daallo Airlines  2016-02-02 00:00:00      159
299     Tara Air  2016-02-24 00:00:00      193
300     Flydubai  2016-03-19 00:00:00      981
301     EgyptAir  2016-03-29 00:00:00      181
302     Batik Air  2016-04-04 00:00:00     7703
303     EgyptAir  2016-05-19 00:00:00      804
304     Korean Air  2016-05-27 00:00:00     2708
305     Singapore Airlines  2016-06-27 00:00:00      368
306     Emirates  2016-08-03 00:00:00      521
307     ASL Airlines Hungary  2016-08-05 00:00:00     7332

```



```

308          Southwest Airlines  2016-08-27 00:00:00      3472
309          American Airlines  2016-10-28 00:00:00       383
310          FedEx Express      2016-10-28 00:00:00       910
311          LaMia Airlines      2016-11-28 00:00:00     2933
312  Pakistan International Airlines  2016-12-07 00:00:00       661
313          Aerosucre           2016-12-20 00:00:00     4544
314          Afriqiyah Airways    2016-12-23 00:00:00       209
315          Turkish Airlines     2017-01-16 00:00:00     6491
316          Peruvian Airlines    2017-03-28 00:00:00       112
317          Summit Air           2017-05-27 00:00:00       409
318          Air Canada           2017-07-07 00:00:00       759
319          Air France           2017-09-30 00:00:00        66
320          West Wind Aviation    2017-12-13 00:00:00       280
321          Pegasus Airlines     2018-01-13 00:00:00     8622

```

[322 rows x 3 columns]

```

In [10]: #TODO: output incidents_df.describe()
         incidents_df.describe()

```

```

Out[10]:
         Airline      Date FlightNum
count          322          322      322
unique          240          309      301
top  Turkish Airlines  2001-09-11 00:00:00      103
freq           6          4          3

```

0.3 Step 4.1

```

In [11]: # TODO: output dictionary {airports: ___, destinations: ___, sources: ___}
         s_id=routes_df.source_airport_id.unique()
         d_id=routes_df.destination_airport_id.unique()
         no_of_source_airports=len(s_id)
         no_of_dest_airports=len(d_id)
         total_airports_ids=len(np.unique(airports_df.airport_id))
         data_dict={'airports':total_airports_ids,'destinations':no_of_dest_airports,'sources':no_of_source_airports}
         data_dict

```

```

Out[11]: {'airports': 8107, 'destinations': 3170, 'sources': 3167}

```

0.4 Step 4.2.1

```

In [12]: # TODO: output airports you can fly to by not from
         x=(set(routes_df.destination_airport_id.unique()).difference(set(routes_df.source_airport_id.unique())))
         x

         list1=[]
         for i in x:
             list1.append(i)
         temp_dict={'temp_col':list1}

```

```
dict_df=pd.DataFrame.from_dict(temp_dict)
name_airports=pd.merge(airports_df,dict_df, left_on='airport_id', right_on='temp_col',h
name_airports
```

```
Out[12]: 0          Filippos
1          Coronel E Carvajal
2  Teniente Coronel Luis A Mantilla
3          Utila Airport
4  Abraham Lincoln Capital
5          Punta Gorda Airport
6          Dalanzadgad Airport
7          Pilot Point Airport
8          Karuluk Airport
9  Port Williams Seaplane Base
10         Zachar Bay Seaplane Base
11         Campo Alegre Airport
12         Breves Airport
13         El Porvenir
14         Oriximina Airport
Name: airport_name, dtype: object
```

0.5 Step 4.2.2

```
In [13]: #TODO: top 10 most popular airports
name_airports=pd.merge(airports_df,routes_df, left_on='airport_id', right_on='destinati

df_top_freq = name_airports.groupby(['airport_id', 'airport_name'])['destination_airpor
{"count": len}).sort_values(
    "count", ascending=False).head(10).reset_index()
df_top_freq
```

```
Out[13]:   airport_id      airport_name  count
0         3682  Hartsfield Jackson Atlanta Intl    905
1         3830    Chicago Ohare Intl    550
2         3364      Capital Intl    526
3          507        Heathrow    524
4         1382    Charles De Gaulle    517
5         3484    Los Angeles Intl    497
6          340    Frankfurt Main    493
7         3670  Dallas Fort Worth Intl    467
8         3797    John F Kennedy Intl    455
9          580      Schiphol    450
```

0.6 Step 4.3

```
In [16]: # TODO: Create dataframe of airline_iata, number of incidents, number of routes
data_frame=incidents_df
data_frame['Airline']=data_frame['Airline'].str.lower()
```

```

df1 = data_frame.groupby(['Airline'])['Airline'].agg(
    {"incident_count": len}).sort_values(
    "incident_count", ascending=False).reset_index()

data_frame1=airlines_df
data_frame1['airline_name']=data_frame1['airline_name'].str.lower()
merged_frame=pd.merge(df1,data_frame1, left_on='Airline', right_on='airline_name')

df2 = routes_df.groupby(['airline_id'])['airline_id'].agg(
    {"route_count": len}).sort_values(
    "route_count", ascending=False).reset_index()
#print(merged_frame.columns)
#print(df2.columns)
df3=pd.merge(merged_frame,df2, left_on='airline_id', right_on='airline_id')

iff=df3[['incident_count', 'iata', 'route_count']]

iff=iff[iff.incident_count > 2]
iff

```

```

Out[16]:
   incident_count iata  route_count
0                6   AA         2354
1                6   TK          652
2                5   MS          205
3                5   AF         1071
4                5   WN         1146
5                4   QF          430
6                4   CI          172
7                4   BA          549
8                3   OZ          262
9                3   GA          248
10               3   UA         2180
11               3   CU           59

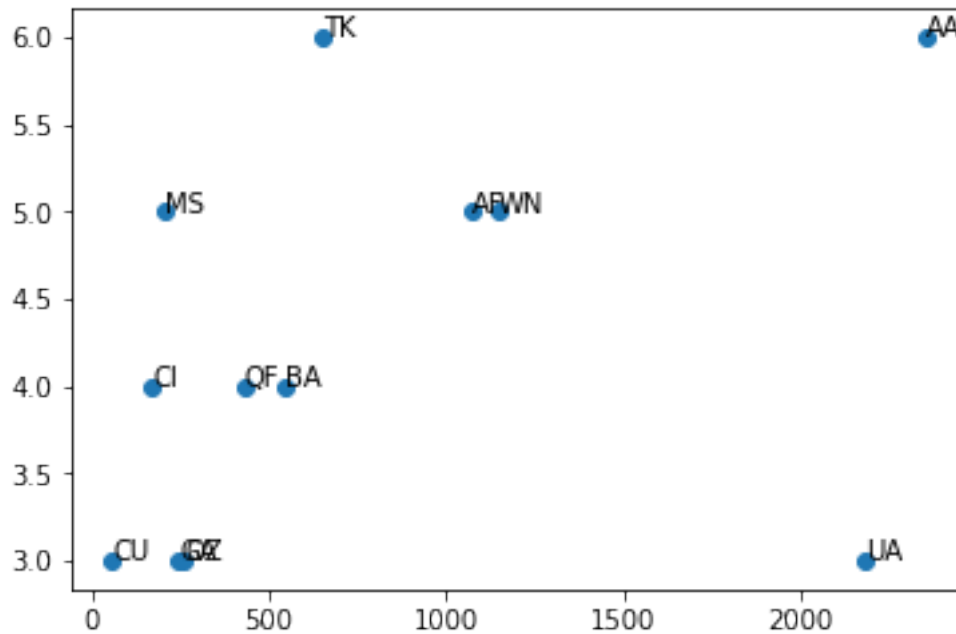
```

```

In [17]: # TODO: Scatter plot of number of routes vs. number of incidents
routes=list(iff.route_count)
no_of_incidents=list(iff.incident_count)
iata=list(iff.iata)

fig, ax = plt.subplots()
ax.scatter(routes,no_of_incidents)
for i, txt in enumerate(iata):
    ax.annotate(txt, (routes[i],no_of_incidents[i]))

```



0.7 Tasks for Step 5

```
In [18]: # TODO: get random sample of flights from SQL
flights_s_df=pd.read_sql_query('SELECT * FROM flights ORDER BY RANDOM() LIMIT 1000', co
flights_s_df
```

```
Out[18]:
```

	Year	Month	DayofMonth	AirlineID	Carrier	FlightNum	Origin	Dest	\
0	2016	10	30	19393	WN	3071	SAT	PHX	
1	2016	10	7	19790	DL	839	SRQ	ATL	
2	2016	10	26	19805	AA	768	MCO	PHL	
3	2016	10	27	19977	UA	1075	LAS	EWR	
4	2016	10	4	19393	WN	2285	MCO	BUF	
5	2016	10	2	20409	B6	1551	ACK	BOS	
6	2016	10	25	19805	AA	320	SAT	DFW	
7	2016	10	14	20304	OO	5899	RNO	SFO	
8	2016	10	15	20304	OO	5254	TUS	ORD	
9	2016	10	23	21171	VX	251	JFK	LAS	
10	2016	10	5	19790	DL	2371	MSP	SMF	
11	2016	10	10	20366	EV	5528	TYS	ATL	
12	2016	10	16	19393	WN	1103	BNA	MSY	
13	2016	10	5	19393	WN	2852	SAN	SMF	
14	2016	10	8	19393	WN	1007	AUS	DEN	
15	2016	10	22	19393	WN	3314	BUR	OAK	
16	2016	10	13	19977	UA	663	EWR	IAH	
17	2016	10	2	19805	AA	1936	CLT	EWR	
18	2016	10	28	19393	WN	951	MSY	BWI	

19	2016	10	7	19393	WN	1840	RNO	LAX
20	2016	10	27	19977	UA	380	SFO	BOS
21	2016	10	23	19805	AA	350	LAX	SFO
22	2016	10	10	20304	OO	5020	LAX	SBP
23	2016	10	19	19790	DL	585	DTW	LAX
24	2016	10	24	19393	WN	259	OKC	HOU
25	2016	10	6	19393	WN	2884	DAL	ABQ
26	2016	10	2	19790	DL	1726	ATL	FLL
27	2016	10	27	19977	UA	1659	CVG	ORD
28	2016	10	20	19393	WN	1291	MDW	GRR
29	2016	10	16	19977	UA	640	ORD	PHL
..
970	2016	10	6	20304	OO	4585	MSN	SLC
971	2016	10	21	19393	WN	3653	LAX	SMF
972	2016	10	13	19790	DL	1386	ATL	DSM
973	2016	10	25	19393	WN	2006	LAS	SFO
974	2016	10	23	19393	WN	2082	OMA	STL
975	2016	10	21	20409	B6	260	PHL	BOS
976	2016	10	3	19790	DL	630	DTW	ATL
977	2016	10	30	20366	EV	5263	RIC	BOS
978	2016	10	2	19790	DL	1739	ATL	TUS
979	2016	10	9	19393	WN	3146	PDX	MDW
980	2016	10	16	19790	DL	1903	ATL	FAY
981	2016	10	17	20304	OO	4752	SLC	EUG
982	2016	10	19	19393	WN	2357	OAK	SEA
983	2016	10	18	20304	OO	4759	SLC	PSC
984	2016	10	7	19790	DL	2478	ATL	JAN
985	2016	10	21	19790	DL	1363	GRR	DTW
986	2016	10	22	19805	AA	2278	LAX	ORD
987	2016	10	20	20304	OO	5151	SGU	DEN
988	2016	10	19	20409	B6	1507	JFK	IAD
989	2016	10	5	19393	WN	1668	SAN	SMF
990	2016	10	7	19805	AA	1205	SNA	ORD
991	2016	10	30	20366	EV	2845	ACT	DFW
992	2016	10	14	19790	DL	1307	RIC	ATL
993	2016	10	11	20366	EV	4430	EWB	SAV
994	2016	10	24	20409	B6	24	LAX	JFK
995	2016	10	15	19977	UA	261	SFO	LAX
996	2016	10	10	20366	EV	5264	CRW	ATL
997	2016	10	3	19393	WN	884	MDW	MSP
998	2016	10	22	20304	OO	4747	CID	DTW
999	2016	10	22	20436	F9	1637	MCO	MKE

	ArrDelayMinutes	Cancelled
0	41.0	0.0
1	0.0	0.0
2	71.0	0.0
3	134.0	0.0

4	0.0	0.0
5	0.0	0.0
6	0.0	0.0
7	NaN	1.0
8	0.0	0.0
9	0.0	0.0
10	0.0	0.0
11	0.0	0.0
12	7.0	0.0
13	6.0	0.0
14	0.0	0.0
15	0.0	0.0
16	0.0	0.0
17	0.0	0.0
18	0.0	0.0
19	5.0	0.0
20	0.0	0.0
21	0.0	0.0
22	0.0	0.0
23	0.0	0.0
24	0.0	0.0
25	0.0	0.0
26	0.0	0.0
27	0.0	0.0
28	0.0	0.0
29	0.0	0.0
..
970	0.0	0.0
971	0.0	0.0
972	0.0	0.0
973	0.0	0.0
974	0.0	0.0
975	0.0	0.0
976	0.0	0.0
977	45.0	0.0
978	0.0	0.0
979	0.0	0.0
980	0.0	0.0
981	0.0	0.0
982	0.0	0.0
983	0.0	0.0
984	0.0	0.0
985	19.0	0.0
986	0.0	0.0
987	1.0	0.0
988	0.0	0.0
989	0.0	0.0
990	0.0	0.0

991	0.0	0.0
992	0.0	0.0
993	0.0	0.0
994	0.0	0.0
995	0.0	0.0
996	0.0	0.0
997	48.0	0.0
998	0.0	0.0
999	5.0	0.0

[1000 rows x 10 columns]

```
In [19]: # TODO: create nested map of Jaccard distance
#df_r=routes_df.drop_duplicates(subset=None, keep='first', inplace=False)
#df_f=flights_s_df.drop_duplicates(subset=None, keep='first', inplace=False)

dict2={}
for element in routes_df:
    element_rcol=routes_df[element].drop_duplicates()
    dict1={}
    my_list=[]
    for item in flights_s_df:
        if np.issubdtype(flights_s_df[item].dtype,np.number):
            continue
        else:
            item_fcol=flights_s_df[item].drop_duplicates()
            intersection=pd.Series(list(set(element_rcol).intersection(set(item_fcol))))
            union=pd.Series(list(set(element_rcol).union(set(item_fcol))))
            if len(intersection)/len(union)==0:
                continue
            else:
                my_list.append(len(intersection)/len(union))
                dict1.update({item:my_list})
                dict2.update({element:dict1})

jaccard_map=dict2
```

0.8 Step 5.1

```
In [20]: # TODO: output nested map
jaccard_map
```

```
Out[20]: {'airline_iata_icao': {'Carrier': [0.01845018450184502]},
'destination_iata_icao': {'Dest': [0.04604225796278776, 0.04290220820189274]},
'Origin': [0.04604225796278776, 0.04290220820189274]},
'source_airport_iata_icao': {'Dest': [0.045755758914484065,
```

```
0.0429428481212504],
'Origin': [0.045755758914484065, 0.0429428481212504]]}}
```

0.9 Step 5.2.1

```
In [21]: # TODO: create joint DataFrame flights_s_df
         flights_s_df=routes_df.merge(flights_s_df, left_on=['airline_iata_icao','destination_ia
         flights_s_df
```

```
Out[21]:
```

	airline_iata_icao	airline_id	source_airport_iata_icao	source_airport_id	\
0	AA	24	ATL	3682	
1	AA	24	ATL	3682	
2	AA	24	AUS	3673	
3	AA	24	BOS	3448	
4	AA	24	BWI	3849	
5	AA	24	BWI	3849	
6	AA	24	BWI	3849	
7	AA	24	BWI	3849	
8	AA	24	CLT	3876	
9	AA	24	CLT	3876	
10	AA	24	CLT	3876	
11	AA	24	CLT	3876	
12	AA	24	CLT	3876	
13	AA	24	CLT	3876	
14	AA	24	CLT	3876	
15	AA	24	CLT	3876	
16	AA	24	CLT	3876	
17	AA	24	CLT	3876	
18	AA	24	CLT	3876	
19	AA	24	DCA	3520	
20	AA	24	DCA	3520	
21	AA	24	DCA	3520	
22	AA	24	DEN	3751	
23	AA	24	DFW	3670	
24	AA	24	DFW	3670	
25	AA	24	DFW	3670	
26	AA	24	DFW	3670	
27	AA	24	DFW	3670	
28	AA	24	DFW	3670	
29	AA	24	DFW	3670	
...	
734	WN	4547	SJC	3748	
735	WN	4547	SJC	3748	
736	WN	4547	SJC	3748	
737	WN	4547	SJC	3748	
738	WN	4547	SLC	3536	
739	WN	4547	SLC	3536	
740	WN	4547	SLC	3536	

741	WN	4547	SLC	3536
742	WN	4547	SLC	3536
743	WN	4547	SMF	3817
744	WN	4547	SMF	3817
745	WN	4547	SMF	3817
746	WN	4547	SMF	3817
747	WN	4547	SMF	3817
748	WN	4547	SMF	3817
749	WN	4547	SNA	3867
750	WN	4547	STL	3678
751	WN	4547	STL	3678
752	WN	4547	STL	3678
753	WN	4547	STL	3678
754	WN	4547	STL	3678
755	WN	4547	STL	3678
756	WN	4547	TPA	3646
757	WN	4547	TPA	3646
758	WN	4547	TPA	3646
759	WN	4547	TPA	3646
760	WN	4547	TUL	3855
761	WN	4547	TUS	3636
762	WN	4547	TUS	3636
763	WN	4547	TUS	3636

	destination_iata_icao	destination_airport_id	codeshare	number_of_stops	\
0	CLT	3876	Y	0	
1	DFW	3670		0	
2	DFW	3670		0	
3	ORD	3830		0	
4	CLT	3876		0	
5	CLT	3876		0	
6	CLT	3876		0	
7	CLT	3876		0	
8	BDL	3825		0	
9	BNA	3690		0	
10	BNA	3690		0	
11	BWI	3849		0	
12	DTW	3645		0	
13	DTW	3645		0	
14	JFK	3797		0	
15	LAS	3877		0	
16	LAX	3484		0	
17	MCO	3878		0	
18	TPA	3646		0	
19	DFW	3670		0	
20	MIA	3576		0	
21	MIA	3576		0	
22	PHX	3462		0	

23	BDL	3825	0
24	CMH	3759	0
25	COS	3819	0
26	HNL	3728	0
27	LAX	3484	0
28	MCI	3458	0
29	MIA	3576	0
..
734	BUR	3644	0
735	BUR	3644	0
736	LAS	3877	0
737	LAX	3484	0
738	DEN	3751	0
739	LAS	3877	0
740	MDW	3747	0
741	OAK	3453	0
742	OAK	3453	0
743	BUR	3644	0
744	LAS	3877	0
745	LAX	3484	0
746	ONT	3734	0
747	SAN	3731	0
748	SAN	3731	0
749	SFO	3469	0
750	BOS	3448	0
751	FLL	3533	0
752	LAS	3877	0
753	OMA	3454	0
754	PHX	3462	0
755	TPA	3646	0
756	BNA	3690	0
757	LAS	3877	0
758	MDW	3747	0
759	MSY	3861	0
760	PHX	3462	0
761	LAS	3877	0
762	LAX	3484	0
763	LAX	3484	0

			Equipments	Year	Month	DayofMonth	AirlineID	Carrier	\
0		319	320 CR9 321	2016	10	30	19805	AA	
1			M80 319 M83	2016	10	22	19805	AA	
2			M80 738 M83	2016	10	23	19805	AA	
3			738	2016	10	19	19805	AA	
4	E90	321	757 319 737 320	2016	10	7	19805	AA	
5	E90	321	757 319 737 320	2016	10	22	19805	AA	
6	E90	321	757 319 737 320	2016	10	27	19805	AA	
7	E90	321	757 319 737 320	2016	10	24	19805	AA	

8		319	321	320	2016	10	4	19805	AA				
9			319	CR7	2016	10	6	19805	AA				
10				319	CR7	2016	10	26	19805				
11	319	737	321	757	320	E90	2016	10	26	19805	AA		
12				319	321	320	2016	10	3	19805	AA		
13				319	321	320	2016	10	6	19805	AA		
14					321	319	2016	10	29	19805	AA		
15						321	2016	10	4	19805	AA		
16							321	2016	10	18	19805	AA	
17				321	767	757	319	2016	10	16	19805	AA	
18				320	321	319	737	2016	10	2	19805	AA	
19					738	M80	2016	10	26	19805	AA		
20						738	2016	10	8	19805	AA		
21						738	2016	10	25	19805	AA		
22					320	321	2016	10	27	19805	AA		
23						738	2016	10	26	19805	AA		
24						M83	M80	2016	10	12	19805	AA	
25						M80	M83	2016	10	26	19805	AA	
26							763	2016	10	21	19805	AA	
27				738	763	757	M83	2016	10	17	19805	AA	
28						M80	M83	2016	10	22	19805	AA	
29						738	763	757	2016	10	23	19805	AA
..					
734						73W	2016	10	3	19393	WN		
735						73W	2016	10	22	19393	WN		
736				73W	733	73C	73H	2016	10	9	19393	WN	
737				73W	733	73C	73H	738	2016	10	4	19393	WN
738				73W	738	733	73C	73H	2016	10	30	19393	WN
739						73W	73C	2016	10	14	19393	WN	
740						73H	73W	2016	10	13	19393	WN	
741						73W	73C	733	2016	10	10	19393	WN
742						73W	73C	733	2016	10	13	19393	WN
743							73W	2016	10	13	19393	WN	
744				73H	73W	733	73C	738	2016	10	3	19393	WN
745				73C	73W	733	73H	738	2016	10	21	19393	WN
746					733	73W	73C	73H	2016	10	16	19393	WN
747					73W	73H	73C	733	2016	10	5	19393	WN
748					73W	73H	73C	733	2016	10	5	19393	WN
749							73W	2016	10	3	19393	WN	
750						73C	73W	2016	10	3	19393	WN	
751						733	73W	2016	10	20	19393	WN	
752				738	73W	73C	73H	733	2016	10	15	19393	WN
753						733	73C	73W	2016	10	23	19393	WN
754						73W	733	73H	2016	10	28	19393	WN
755						73W	73C	73H	2016	10	24	19393	WN
756						73C	73W	733	2016	10	1	19393	WN
757							73H	73W	2016	10	10	19393	WN
758							73H	73W	2016	10	4	19393	WN

759		73W 733	2016	10	31	19393	WN
760		73W	2016	10	12	19393	WN
761		73W 73C	2016	10	30	19393	WN
762	73C 73W 733 73H		2016	10	2	19393	WN
763	73C 73W 733 73H		2016	10	18	19393	WN

	FlightNum	Origin	Dest	ArrDelayMinutes	Cancelled
0	589	CLT	ATL	10.0	0.0
1	1473	DFW	ATL	0.0	0.0
2	1561	DFW	AUS	0.0	0.0
3	1205	ORD	BOS	0.0	0.0
4	1998	CLT	BWI	0.0	0.0
5	703	CLT	BWI	0.0	0.0
6	1924	CLT	BWI	0.0	0.0
7	1960	CLT	BWI	0.0	0.0
8	1837	BDL	CLT	0.0	0.0
9	2065	BNA	CLT	3.0	0.0
10	2065	BNA	CLT	0.0	0.0
11	1798	BWI	CLT	7.0	0.0
12	756	DTW	CLT	3.0	0.0
13	1713	DTW	CLT	0.0	0.0
14	148	JFK	CLT	0.0	0.0
15	431	LAS	CLT	33.0	0.0
16	696	LAX	CLT	0.0	0.0
17	746	MCO	CLT	0.0	0.0
18	1719	TPA	CLT	0.0	0.0
19	2290	DFW	DCA	0.0	0.0
20	2239	MIA	DCA	7.0	0.0
21	1768	MIA	DCA	0.0	0.0
22	1704	PHX	DEN	12.0	0.0
23	1581	BDL	DFW	0.0	0.0
24	938	CMH	DFW	0.0	0.0
25	160	COS	DFW	0.0	0.0
26	8	HNL	DFW	0.0	0.0
27	2473	LAX	DFW	11.0	0.0
28	1046	MCI	DFW	0.0	0.0
29	1541	MIA	DFW	0.0	0.0
..
734	2542	BUR	SJC	6.0	0.0
735	610	BUR	SJC	0.0	0.0
736	562	LAS	SJC	10.0	0.0
737	1796	LAX	SJC	20.0	0.0
738	3125	DEN	SLC	0.0	0.0
739	416	LAS	SLC	11.0	0.0
740	1078	MDW	SLC	0.0	0.0
741	2933	OAK	SLC	0.0	0.0
742	2933	OAK	SLC	0.0	0.0
743	1754	BUR	SMF	0.0	0.0

744	232	LAS	SMF	0.0	0.0
745	3653	LAX	SMF	0.0	0.0
746	747	ONT	SMF	39.0	0.0
747	2852	SAN	SMF	6.0	0.0
748	1668	SAN	SMF	0.0	0.0
749	2212	SFO	SNA	0.0	0.0
750	3099	BOS	STL	11.0	0.0
751	826	FLL	STL	0.0	0.0
752	3499	LAS	STL	0.0	0.0
753	2082	OMA	STL	0.0	0.0
754	1924	PHX	STL	23.0	0.0
755	499	TPA	STL	0.0	0.0
756	1209	BNA	TPA	0.0	0.0
757	2955	LAS	TPA	0.0	0.0
758	806	MDW	TPA	1.0	0.0
759	1487	MSY	TPA	33.0	0.0
760	2570	PHX	TUL	55.0	0.0
761	556	LAS	TUS	25.0	0.0
762	2436	LAX	TUS	0.0	0.0
763	2436	LAX	TUS	11.0	0.0

[764 rows x 19 columns]

```
In [24]: # TODO: output for each airline IATA code and destination city IATA code, the number of
fre_df=flights_s_df.groupby(['airline_iata_icao','destination_iata_icao'])['airline_iat
{"count": len}).sort_values(
    "count", ascending=False).reset_index()
fre_df
```

```
Out[24]:
```

	airline_iata_icao	destination_iata_icao	count
0	DL	ATL	56
1	UA	ORD	20
2	AA	DFW	19
3	AA	CLT	16
4	WN	LAS	15
5	WN	LAX	13
6	WN	MDW	13
7	B6	BOS	13
8	AA	MIA	13
9	DL	MSP	13
10	WN	DEN	12
11	B6	JFK	12
12	DL	DTW	12
13	UA	EWB	12
14	AA	PHX	11
15	AS	SEA	11
16	UA	DEN	10
17	WN	BWI	10

18	UA	IAH	10
19	WN	MCO	9
20	DL	LAX	8
21	WN	SJC	8
22	DL	SLC	8
23	WN	SAN	8
24	UA	SFO	7
25	WN	OAK	7
26	AA	ORD	7
27	WN	TPA	6
28	UA	LAX	6
29	WN	PHX	6
..
214	B6	PBI	1
215	UA	PHL	1
216	B6	BDL	1
217	B6	BUF	1
218	UA	OMA	1
219	UA	MSY	1
220	B6	DCA	1
221	B6	DEN	1
222	B6	DFW	1
223	B6	EWR	1
224	B6	HPN	1
225	UA	DTW	1
226	B6	LAX	1
227	B6	OAK	1
228	UA	BNA	1
229	HA	LAX	1
230	UA	ABQ	1
231	NK	RSW	1
232	NK	PHL	1
233	B6	PDX	1
234	NK	OAK	1
235	NK	MYR	1
236	B6	PSE	1
237	NK	LAS	1
238	B6	RIC	1
239	B6	ROC	1
240	NK	DEN	1
241	HA	OGG	1
242	HA	OAK	1
243	DL	RIC	1

[244 rows x 3 columns]

0.10 Step 5.2.2

In [32]: *# TODO: output bad flights*

```
df_1=flights_s_df[(flights_s_df.ArrDelayMinutes >= 30)]
df_2=flights_s_df[flights_s_df.Cancelled > 0 ]
df1=df_1.append(df_2)
fre_df1=df1.groupby(['airline_iata_icao','destination_iata_icao'])['airline_iata_icao']
        {"count": len}).sort_values(
        "count", ascending=False).reset_index()
fre_df1
```

```
Out[32]:
```

	airline_iata_icao	destination_iata_icao	count
0	B6	BOS	5
1	WN	LAX	4
2	UA	ORD	3
3	AA	MIA	3
4	DL	ATL	3
5	WN	MDW	2
6	UA	BOS	2
7	AA	DFW	2
8	F9	DEN	2
9	B6	JFK	2
10	UA	EWR	2
11	AA	CLT	2
12	AA	PHX	2
13	WN	SAT	2
14	WN	MSY	1
15	WN	ONT	1
16	WN	PDX	1
17	WN	DEN	1
18	WN	BOS	1
19	VX	SFO	1
20	VX	IAD	1
21	WN	PHX	1
22	UA	SFO	1
23	WN	PIT	1
24	UA	LAS	1
25	HA	LAX	1
26	UA	BWI	1
27	NK	LGA	1
28	F9	MCO	1
29	DL	SLC	1
30	DL	LAX	1
31	DL	IAD	1
32	DL	CLT	1
33	B6	HPN	1
34	B6	FLL	1
35	B6	DCA	1
36	AS	SEA	1

37	AS	EWR	1
38	AA	SAT	1
39	AA	MCO	1
40	AA	LAS	1
41	WN	SNA	1

0.11 Step 5.2.3

```
In [36]: # TODO: pairs with more than one bad flight
freq=fre_df1[(fre_df1['count'])>1]
freq
```

```
Out[36]:
```

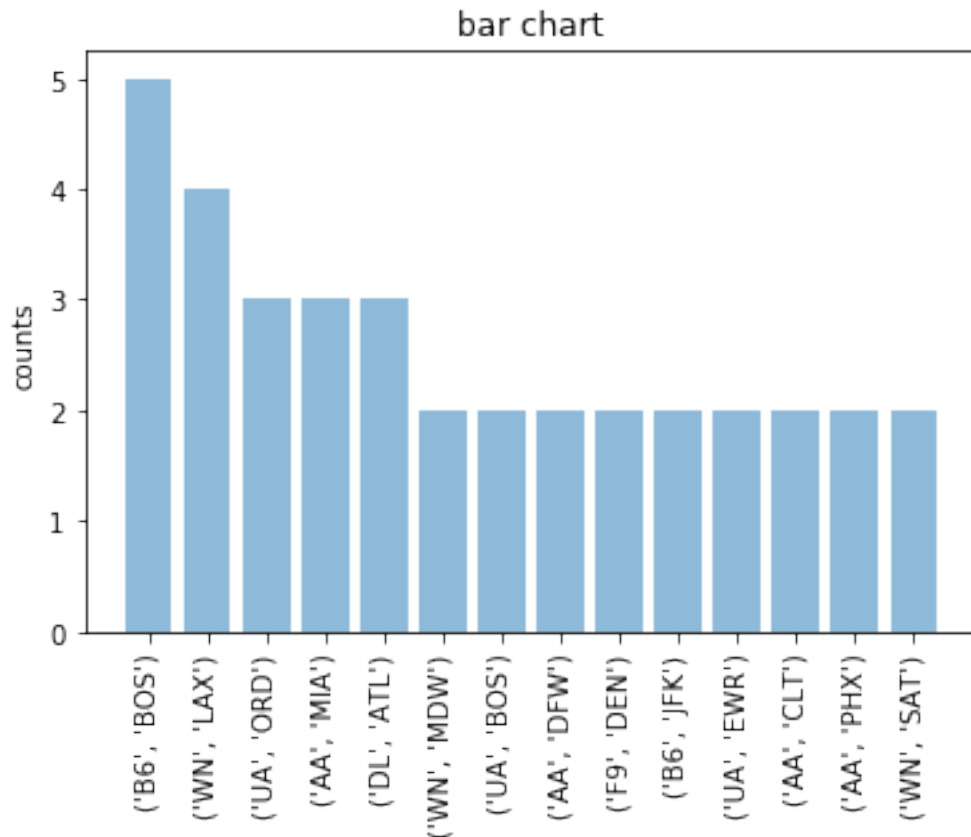
	airline_iata_icao	destination_iata_icao	count
0	B6	BOS	5
1	WN	LAX	4
2	UA	ORD	3
3	AA	MIA	3
4	DL	ATL	3
5	WN	MDW	2
6	UA	BOS	2
7	AA	DFW	2
8	F9	DEN	2
9	B6	JFK	2
10	UA	EWR	2
11	AA	CLT	2
12	AA	PHX	2
13	WN	SAT	2

```
In [53]: # TODO: bar chart for bad flights
delay_flight=list(freq['count'])
airline_iata=list(freq.airline_iata_icao)
dest_iata=list(freq.destination_iata_icao)
pos_df=pd.DataFrame({'air':airline_iata,'dest':dest_iata})

pos=[]
for i in range(len(pos_df)):
    a=pos_df.ix[i,'air']
    d=pos_df.ix[i,'dest']
    pos.append((a,d))

plt.bar(np.arange(len(delay_flight)), delay_flight, align='center', alpha=0.5)
plt.xticks(np.arange(len(delay_flight)), pos,rotation='vertical')
plt.ylabel('counts')
plt.title('bar chart')

plt.show()
```

0.12 Step 5.2.4

In []: *# TODO: delayed flights by airline*

```
n_df=flights_s_df[(flights_s_df.ArrDelayMinutes >= 30)]
n_df
```

```
In [38]: fre_df1=df1.groupby(['airline_iata_icao'])['airline_iata_icao'].agg(
        {"count": len}).sort_values(
        "count", ascending=False).reset_index()
fre_df1
```

```
Out[38]:   airline_iata_icao  count
0           WN         16
1           AA         12
2           B6         10
3           UA         10
4           DL          7
5           F9          3
6           AS          2
7           VX          2
```

8	HA	1
9	NK	1