

United States Airlines Analysis

January 30, 2023

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

1 Project Task: Week 1

1.1 Applied data science with Python

1.2 1. Import and aggregate data:

a. Collect information related to flights, airports (e.g., type of airport and elevation), and runways (e.g., length_ft, width_ft, surface, and number of runways). Gather all fields you believe might cause avoidable delays in one dataset. Hint: In this case, you would have to determine the keys to join the tables. A data description will be useful.

```
[2]: airport = pd.read_excel('airports.xlsx')
airlines = pd.read_excel('Airlines.xlsx')
runways = pd.read_excel('runways.xlsx')
```

```
[3]: airport.head()
```

```
[3]:
```

	id	ident	type	name \
0	6523	00A	heliport	Total Rf Heliport
1	323361	00AA	small_airport	Aero B Ranch Airport
2	6524	00AK	small_airport	Lowell Field
3	6525	00AL	small_airport	Epps Airpark
4	6526	00AR	closed	Newport Hospital & Clinic Heliport

	latitude_deg	longitude_deg	elevation_ft	continent	iso_country	iso_region \
0	40.070801	-74.933601	11.0	NaN	US	US-PA
1	38.704022	-101.473911	3435.0	NaN	US	US-KS
2	59.947733	-151.692524	450.0	NaN	US	US-AK
3	34.864799	-86.770302	820.0	NaN	US	US-AL

4	35.608700	-91.254898	237.0	NaN	US	US-AR
---	-----------	------------	-------	-----	----	-------

	municipality	scheduled_service	gps_code	iata_code	local_code	home_link	\
0	Bensalem	no	00A	NaN	00A	NaN	
1	Leoti	no	00AA	NaN	00AA	NaN	
2	Anchor Point	no	00AK	NaN	00AK	NaN	
3	Harvest	no	00AL	NaN	00AL	NaN	
4	Newport	no	NaN	NaN	NaN	NaN	

	wikipedia_link	keywords
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	00AR

```
[4]: airport.shape
```

```
[4]: (73805, 18)
```

```
[5]: airlines.head()
```

	id	Airline	Flight	AirportFrom	AirportTo	DayOfWeek	Time	Length	Delay
0	1	CO	269	SFO	IAH	3	15	205	1
1	2	US	1558	PHX	CLT	3	15	222	1
2	3	AA	2400	LAX	DFW	3	20	165	1
3	4	AA	2466	SFO	DFW	3	20	195	1
4	5	AS	108	ANC	SEA	3	30	202	0

```
[6]: airlines.shape
```

```
[6]: (518556, 9)
```

```
[7]: runways.head()
```

	id	airport_ref	airport_ident	length_ft	width_ft	surface	lighted	\
0	269408	6523	00A	80.0	80.0	ASPH-G	1	
1	255155	6524	00AK	2500.0	70.0	GRVL	0	
2	254165	6525	00AL	2300.0	200.0	TURF	0	
3	270932	6526	00AR	40.0	40.0	GRASS	0	
4	322128	322127	00AS	1450.0	60.0	Turf	0	

	closed	le_ident	le_latitude_deg	le_longitude_deg	le_elevation_ft	\
0	0	H1	NaN	NaN	NaN	
1	0	N	NaN	NaN	NaN	
2	0	1	NaN	NaN	NaN	
3	0	H1	NaN	NaN	NaN	

4	0	1	NaN	NaN	NaN
---	---	---	-----	-----	-----

	le_heading_degT	le_displaced_threshold_ft	he_ident	he_latitude_deg	\
0	NaN	NaN	NaN	NaN	
1	NaN	NaN	S	NaN	
2	NaN	NaN	19	NaN	
3	NaN	NaN	H1	NaN	
4	NaN	NaN	19	NaN	

	he_longitude_deg	he_elevation_ft	he_heading_degT	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	

	he_displaced_threshold_ft
0	NaN
1	NaN
2	NaN
3	NaN
4	NaN

```
[8]: runways.shape
```

```
[8]: (43977, 20)
```

Before merging the data lets drop the columns that will not play an important role in the model building

```
[10]: runways.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 43977 entries, 0 to 43976
Data columns (total 20 columns):
#   Column              Non-Null Count  Dtype
---  -
0   id                  43977 non-null  int64
1   airport_ref         43977 non-null  int64
2   airport_ident       43977 non-null  object
3   length_ft           43753 non-null  float64
4   width_ft            41088 non-null  float64
5   surface             43520 non-null  object
6   lighted             43977 non-null  int64
7   closed              43977 non-null  int64
8   le_ident            43793 non-null  object
9   le_latitude_deg     15016 non-null  float64
10  le_longitude_deg    15000 non-null  float64
```

```

11 le_elevation_ft          12781 non-null float64
12 le_heading_degT          14624 non-null float64
13 le_displaced_threshold_ft 2883 non-null float64
14 he_ident                 37332 non-null object
15 he_latitude_deg          14971 non-null float64
16 he_longitude_deg         14973 non-null float64
17 he_elevation_ft          12620 non-null float64
18 he_heading_degT          16428 non-null float64
19 he_displaced_threshold_ft 3176 non-null float64
dtypes: float64(12), int64(4), object(4)
memory usage: 6.7+ MB

```

```

[11]: runways.drop(['le_ident', 'le_latitude_deg', 'le_longitude_deg',
↳ 'le_elevation_ft', 'le_heading_degT',
        'le_displaced_threshold_ft', 'he_ident',
↳ 'he_latitude_deg', 'he_longitude_deg', 'he_elevation_ft', 'he_heading_degT',
        'he_displaced_threshold_ft'], axis = 1, inplace=True)

```

```

[12]: runways

```

```

[12]:
      id  airport_ref  airport_ident  length_ft  width_ft  surface \
0    269408         6523           00A      80.0     80.0   ASPH-G
1    255155         6524           00AK     2500.0     70.0    GRVL
2    254165         6525           00AL     2300.0    200.0    TURF
3    270932         6526           00AR      40.0     40.0   GRASS
4    322128        322127           00AS     1450.0     60.0    Turf
...
43972  235186        27243           ZYTX     10499.0    148.0     CON
43973  235169        27244           ZYYJ     8530.0    148.0     CON
43974  354997        317861           ZYYK     8202.0      NaN     NaN
43975  346789        346788      ZZ-0003     1800.0     15.0    Turf
43976  313663        313629           ZZZZ     1713.0     82.0  concrete

```

```

      lighted  closed
0           1        0
1           0        0
2           0        0
3           0        0
4           0        0
...
43972        1        0
43973        1        0
43974        0        0
43975        0        0
43976        0        0

```

```

[43977 rows x 8 columns]

```

Now lets remove the feature from the airport data that is not usefull.

```
[14]: airport.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 73805 entries, 0 to 73804
Data columns (total 18 columns):
#   Column                Non-Null Count  Dtype
---  -
0   id                    73805 non-null  int64
1   ident                 73805 non-null  object
2   type                  73805 non-null  object
3   name                  73805 non-null  object
4   latitude_deg          73805 non-null  float64
5   longitude_deg          73805 non-null  float64
6   elevation_ft          59683 non-null  float64
7   continent              38086 non-null  object
8   iso_country            73546 non-null  object
9   iso_region            73805 non-null  object
10  municipality           68739 non-null  object
11  scheduled_service      73805 non-null  object
12  gps_code               42996 non-null  object
13  iata_code              9160 non-null   object
14  local_code             32975 non-null  object
15  home_link              3492 non-null   object
16  wikipedia_link         10705 non-null  object
17  keywords               13951 non-null  object
dtypes: float64(3), int64(1), object(14)
memory usage: 10.1+ MB
```

```
[15]: airport.drop(['continent', 'iso_country', 'iso_region', 'municipality',
→ 'gps_code', 'local_code', 'home_link',
→ 'wikipedia_link', 'keywords'], axis=1, inplace=True)
airport
```

```
[15]:
```

	id	ident	type	name \
0	6523	00A	heliport	Total Rf Heliport
1	323361	00AA	small_airport	Aero B Ranch Airport
2	6524	00AK	small_airport	Lowell Field
3	6525	00AL	small_airport	Epps Airpark
4	6526	00AR	closed	Newport Hospital & Clinic Heliport
...
73800	46378	ZZ-0001	heliport	Sealand Helipad
73801	307326	ZZ-0002	small_airport	Glorioso Islands Airstrip
73802	346788	ZZ-0003	small_airport	Fainting Goat Airport
73803	342102	ZZZW	closed	Scandium City Heliport
73804	313629	ZZZZ	small_airport	Satsuma I qjima Airport

	latitude_deg	longitude_deg	elevation_ft	scheduled_service	iata_code
0	40.070801	-74.933601	11.0	no	NaN
1	38.704022	-101.473911	3435.0	no	NaN
2	59.947733	-151.692524	450.0	no	NaN
3	34.864799	-86.770302	820.0	no	NaN
4	35.608700	-91.254898	237.0	no	NaN
...
73800	51.894444	1.482500	40.0	no	NaN
73801	-11.584278	47.296389	11.0	no	NaN
73802	32.110587	-97.356312	690.0	no	NaN
73803	69.355287	-138.939310	4.0	no	ZYW
73804	30.784722	130.270556	338.0	no	NaN

[73805 rows x 9 columns]

Now lets merge the runways and airport data.

```
[17]: airport_runway = pd.merge(airport, runways, left_on = "ident", right_on = "ident",
    ↪ "airport_ident")
```

```
[19]: airport_runway.drop(['id_x', 'id_y'], axis=1, inplace=True)
```

```
[20]: airport_runway
```

```
[20]:
```

	ident	type	name \
0	00A	heliport	Total Rf Heliport
1	00AK	small_airport	Lowell Field
2	00AL	small_airport	Epps Airpark
3	00AR	closed	Newport Hospital & Clinic Heliport
4	00AS	small_airport	Fulton Airport
...
43972	ZYTX	large_airport	Shenyang Taoxian International Airport
43973	ZYYJ	medium_airport	Yanji Chaoyangchuan Airport
43974	ZYYK	medium_airport	Yingkou Lanqi Airport
43975	ZZ-0003	small_airport	Fainting Goat Airport
43976	ZZZZ	small_airport	Satsuma Içjima Airport

	latitude_deg	longitude_deg	elevation_ft	scheduled_service	iata_code \
0	40.070801	-74.933601	11.0	no	NaN
1	59.947733	-151.692524	450.0	no	NaN
2	34.864799	-86.770302	820.0	no	NaN
3	35.608700	-91.254898	237.0	no	NaN
4	34.942803	-97.818019	1100.0	no	NaN
...
43972	41.639801	123.483002	198.0	yes	SHE
43973	42.882801	129.451004	624.0	yes	YNJ
43974	40.542524	122.358600	NaN	yes	YKH

43975	32.110587	-97.356312	690.0	no	NaN
43976	30.784722	130.270556	338.0	no	NaN

	airport_ref	airport_ident	length_ft	width_ft	surface	lighted	\
0	6523	00A	80.0	80.0	ASPH-G	1	
1	6524	00AK	2500.0	70.0	GRVL	0	
2	6525	00AL	2300.0	200.0	TURF	0	
3	6526	00AR	40.0	40.0	GRASS	0	
4	322127	00AS	1450.0	60.0	Turf	0	
...		
43972	27243	ZYTX	10499.0	148.0	CON	1	
43973	27244	ZYYJ	8530.0	148.0	CON	1	
43974	317861	ZYYK	8202.0	NaN	NaN	0	
43975	346788	ZZ-0003	1800.0	15.0	Turf	0	
43976	313629	ZZZZ	1713.0	82.0	concrete	0	

	closed
0	0
1	0
2	0
3	0
4	0
...	...
43972	0
43973	0
43974	0
43975	0
43976	0

[43977 rows x 15 columns]

Now lets merge the final column airline.

```
[22]: final_df = pd.merge(airlines,airport_runway,how = "inner", left_on = "id",
    ↪ "AirportFrom", right_on = "iata_code" )
```

```
[23]: final_df.drop_duplicates(subset=['id'], keep='first', inplace=True)
final_df
```

	id	Airline	Flight	AirportFrom	AirportTo	DayOfWeek	Time	\
0	1	CO	269	SFO	IAH	3	15	
4	4	AA	2466	SFO	DFW	3	20	
8	9	DL	2606	SFO	MSP	3	35	
12	129	DL	1580	SFO	DTW	3	345	
16	150	UA	756	SFO	DEN	3	348	
...		
2160266	451344	CO	2	GUM	HNL	1	400	

2160268	469866	CO	2	GUM	HNL	2	400
2160270	488365	CO	2	GUM	HNL	3	400
2160272	506855	CO	2	GUM	HNL	4	400
2160274	525138	CO	2	GUM	HNL	5	400

	Length	Delay	ident	...	elevation_ft	scheduled_service	iata_code	\
0	205	1	KSFO	...	13.0	yes	SFO	
4	195	1	KSFO	...	13.0	yes	SFO	
8	216	1	KSFO	...	13.0	yes	SFO	
12	270	0	KSFO	...	13.0	yes	SFO	
16	158	0	KSFO	...	13.0	yes	SFO	
...	
2160266	430	1	PGUM	...	298.0	yes	GUM	
2160268	430	1	PGUM	...	298.0	yes	GUM	
2160270	430	0	PGUM	...	298.0	yes	GUM	
2160272	430	1	PGUM	...	298.0	yes	GUM	
2160274	430	1	PGUM	...	298.0	yes	GUM	

	airport_ref	airport_ident	length_ft	width_ft	surface	lighted	\
0	3878	KSFO	7500.0	200.0	ASP	1	
4	3878	KSFO	7500.0	200.0	ASP	1	
8	3878	KSFO	7500.0	200.0	ASP	1	
12	3878	KSFO	7500.0	200.0	ASP	1	
16	3878	KSFO	7500.0	200.0	ASP	1	
...	
2160266	5433	PGUM	12015.0	150.0	asphalt	1	
2160268	5433	PGUM	12015.0	150.0	asphalt	1	
2160270	5433	PGUM	12015.0	150.0	asphalt	1	
2160272	5433	PGUM	12015.0	150.0	asphalt	1	
2160274	5433	PGUM	12015.0	150.0	asphalt	1	

	closed
0	0
4	0
8	0
12	0
16	0
...	...
2160266	0
2160268	0
2160270	0
2160272	0
2160274	0

[518525 rows x 24 columns]

- 1.2.1 b. When it comes to on-time arrivals, different airlines perform differently based on the amount of experience they have. The major airlines in this field include US Airways Express (founded in 1967) Continental Airlines (founded in 1934), and Express Jet (founded in 1986). Pull such information specific to various airlines from the Wikipedia page link given below. https://en.wikipedia.org/wiki/List_of_airlines_of_the_United_States.

Hint: Here, you should use web scraping to learn how long an airline has been operating.

Now lets use the web scrapping to import the data from the wikipedia.

```
[24]: url = "https://en.wikipedia.org/wiki/List_of_airlines_of_the_United_States"
      tables = pd.read_html(url)
```

```
[25]: print(tables)
```

	Airline	Image	IATA	ICAO	Callsign	\
0	Alaska Airlines	NaN	AS	ASA	ALASKA	
1	Allegiant Air	NaN	G4	AAY	ALLEGIANT	
2	American Airlines	NaN	AA	AAL	AMERICAN	
3	Avelo Airlines	NaN	XP	VXP	AVELO	
4	Breeze Airways	NaN	MX	MXV	MOXY	
5	Delta Air Lines	NaN	DL	DAL	DELTA	
6	Eastern Airlines	NaN	2D	EAL	EASTERN	
7	Frontier Airlines	NaN	F9	FFT	FRONTIER FLIGHT	
8	Hawaiian Airlines	NaN	HA	HAL	HAWAIIAN	
9	JetBlue	NaN	B6	JBU	JETBLUE	
10	Southwest Airlines	NaN	WN	SWA	SOUTHWEST	
11	Spirit Airlines	NaN	NK	NKS	SPIRIT WINGS	
12	Sun Country Airlines	NaN	SY	SCX	SUN COUNTRY	
13	United Airlines	NaN	UA	UAL	UNITED	

	Primary hubs,	Secondary hubs	Founded	\
0	Seattle/Tacoma	AnchoragePortland (OR)San Franci...	1932	
1	Las Vegas	CincinnatiFort Walton BeachIndianapol...	1997	
2	Dallas/Fort Worth	CharlotteChicago-O'HareLos An...	1926	
3		BurbankNew HavenOrlando	1987	
4	Charleston	HartfordNew OrleansNorfolkProvoTampa	2018	
5	Atlanta	BostonDetroitLos AngelesMinneapolis/St...	1924	
6		MiamiNew York-JFK	2010	
7	Denver	AtlantaChicago-O'HareCincinnatiCleveland...	1994	
8		HonoluluKahului	1929	
9	New York-JFK	BostonLos AngelesFort LauderdaleOr...	1998	
10	Dallas-Love	AtlantaBaltimoreChicago-MidwayDenve...	1967	
11	Atlantic City	DetroitLas VegasFort LauderdaleCh...	1980	
12	Minneapolis/St. Paul	Dallas/Fort WorthLas Vegas	1982	
13	Chicago-O'Hare	DenverGuamHouston-Intercontinent...	1926	

	Notes					
0	Founded as McGee Airways and commenced operati...					
1	Founded as WestJet Express and commenced opera...					
2	Founded as American Airways and commenced oper...					
3	First did business as Casino Express Airlines ...					
4	NaN					
5	Founded as Huff Daland Dusters and commenced o...					
6	NaN					
7	NaN					
8	Founded as Inter-Island Airways in early 1929 ...					
9	Founded as New Air and commenced operations in...					
10	Founded as Air Southwest and commenced operati...					
11	Founded as Charter One.					
12	Commenced operations in 1983.Operates some Ama...					
13	Founded as Varney Air Lines and commenced oper... ,					Airline
	Image	IATA	ICAO	Callsign \		
0	Air Wisconsin	NaN	ZW	AWI	WISCONSIN	
1	Cape Air	NaN	9K	KAP	CAIR	
2	CommutAir	NaN	C5	UCA	COMMUTAIR	
3	Contour Airlines	NaN	LF	VTE	VOLUNTEER	
4	Elite Airways	NaN	7Q	MNU	MAINER	
5	Endeavor Air	NaN	9E	EDV	ENDEAVOR	
6	Envoy Air	NaN	MQ	ENY	ENVOY	
7	GoJet Airlines	NaN	G7	GJS	LINDBERGH	
8	Horizon Air	NaN	QX	QXE	HORIZON	
9	Mesa Airlines	NaN	YV	ASH	AIR SHUTTLE	
10	Piedmont Airlines	NaN	PT	PDT	PIEDMONT	
11	PSA Airlines	NaN	OH	JIA	BLUE STREAK	
12	Republic Airways	NaN	YX	RPA	BRICKYARD	
13	Silver Airways	NaN	3M	SIL	SILVER WINGS	
14	SkyWest Airlines	NaN	OO	SKW	SKYWEST	
	Primary Hubs, Secondary Hubs					Founded \
0	Appleton	Chicago-O'Hare	Columbia	Milwaukee	Washington...	1965
1	Hyannis	Billings	Boston	Nantucket	St. Louis	San Jua...
2		Denver	Newark	Washington	Dulles	1989
3				Smyrna (TN)		1982
4		Melbourne/Orlando	Newark	Portland (Maine)		2006
5	Minneapolis/St. Paul	Atlanta	Cincinnati	Detroit	N...	1985
6		Dallas/Fort Worth	Chicago-O'Hare	Miami		1984
7			Chicago-O'Hare	Denver		2004
8		Seattle/Tacoma	Portland (OR)			1981
9	As American Eagle:	Phoenix-Sky Harbor	Dallas/For...			1980
10		Charlotte	Philadelphia	Washington-National		1961
11		Charlotte	Philadelphia	Washington-National		1979
12	As American Eagle:	Indianapolis	Columbus (OH)	Kan...		1998
13		Fort Lauderdale	Orlando	Tampa		2011
14	As Delta Connection:	Atlanta	Boise	Colorado Sprin...		1972

						Notes
0						Operates as United Express
1						NaN
2						Operates as United Express.
3						NaN
4						Commenced operations in 2014.
5						Founded as Express Airlines I. Operates as Del...
6						Founded as American Eagle Airlines. Operates a...
7						Commenced operations in 2005. Operates as Unit...
8						Operates as Alaska Airlines.
9						Founded as Mesa Air Shuttle. All but one aircr...
10						Founded as Henson Aviation and commenced opera...
11						Founded as Vee Neal Airlines. Operates as Amer...
12						Commenced operations in 2005. Operates as Amer...
13						NaN
14						Operates as Delta Connection, United Express, ... ,
Airline	Image	IATA	ICAO	Callsign	\	
0				Advanced Air	NaN AN WSN	WINGSPAN
1				Air Sunshine	NaN YI RSI	AIR SUNSHINE
2				Bering Air	NaN 8E BRG	BERING AIR
3				Boutique Air	NaN 4B BTQ	BOUTIQUE
4				Everts Air	NaN 5V VTS	EVERTS
5				Gem Air	NaN NaN NaN	NaN
6				Grand Canyon Airlines	NaN YR CVU	CANYON VIEW
7				Grand Canyon Scenic Airlines	NaN YR SCE	SCENIC
8				Grant Aviation	NaN GV GUN	HOOT
9				Griffing Flying Service	NaN NaN NaN	NaN
10				Island Airways	NaN NaN NaN	NaN
11				JSX	NaN XE JSX	BIGSTRIPE
12				Kenmore Air	NaN M5 KEN	KENMORE
13				Key Lime Air	NaN KG LYM	KEY LIME
14				Mokulele Airlines	NaN MW MHO	MAHALO
15				New England Airlines	NaN EJ NEA	NEW ENGLAND
16				Penobscot Island Air	NaN NaN NaN	NaN
17				Reliant Air	NaN NaN RLI	RELIANT
18				San Juan Airlines	NaN NaN NaN	SKYFERRY
19				Servant Air	NaN 8D NaN	NaN
20				Southern Airways Express	NaN 9X FDY	FRIENDLY
21				Surf Air	NaN NaN UF	SURFAIR
22				Taquan Air	NaN K3 TQN	TAQUAN
23				Tradewind Aviation	NaN TJ GPD	GOODSPEED
24				Ultimate Air Shuttle	NaN UE UJC	ULTIMATE
25				Utah Airways	NaN NaN NaN	NaN
26				Warbelow's Air Ventures	NaN 4W WAV	WARBELOW
27				Wright Air Service	NaN 8V WRF	WRIGHT FLYER

Primary Hubs, Secondary Hubs Founded \

0	Hawthorne	2005
1	San Juan	1982
2	NomeKotzebueUnalakleet	1979
3	Dallas/Fort WorthDenverPhoenix-Sky Harbor	2007
4	FairbanksAnchorage	1978
5	Salmon	2014
6	Boulder CityGrand CanyonPage	1927
7	Grand Canyon	1967
8	AnchorageBethelCold BayDillinghamEmmonakKenaiK...	1971
9	Port Clinton	1937
10	Charlevoix	1945
11	BurbankOaklandLas VegasSanta AnaPhoenixConcord	2016
12	KenmoreSeattle-Lake UnionSeattle-Boeing	1946
13	Denver-CentennialDenverDenver-Rocky MountainGr...	1997
14	Kailua-KonaKahului	1994
15	Westerly	1970
16	Rockland	2004
17	Danbury	1988
18	Bellingham	2002
19	Kodiak	2003
20	MemphisDestinPittsburghWashington-Dulles	2013
21	HawthorneOaklandSan CarlosSanta BarbaraTruckee	2012
22	Ketchikan Harbor	1977
23	Oxford (CT)San Juan White Plains	2001
24	Cincinnati-Lunken	2009
25	Ogden	2015
26	Fairbanks	1958
27	Fairbanks	1966

Notes

0	Has the EAS contract to serve Grant County Air...
1	NaN
2	NaN
3	NaN
4	Founded as Tatonduk Flying Service.
5	NaN
6	Founded as Scenic Airways.
7	Founded as Scenic Airlines.
8	Founded as Delta Air Services.
9	NaN
10	Founded as McPhillips Flying Service.
11	NaN
12	Founded as Mines Collins Munro.
13	Operates as Denver Air Connection.
14	Founded as Mokulele Flight Service.
15	NaN
16	NaN
17	NaN

18						NaN
19						NaN
20						NaN
21						NaN
22						NaN
23						NaN
24						NaN
25						NaN
26						NaN
27						NaN ,

Airline	Image	IATA	ICAO	Callsign	\		
0				Air Charter Bahamas	NaN	NaN	NaN
1				Air Flight Charters	NaN	NaN	FLL
2				Airshare	NaN	NaN	XSR
3				Berry Aviation	NaN	NaN	BYA
4				Bighorn Airways	NaN	NaN	BHR
5				Charter Air Transport	NaN	VC	SRY
6				Choice Airways	NaN	NaN	CSX
7				ExcelAire	NaN	NaN	XLS
8				Global Crossing Airlines	NaN	G6	GXA
9				Great Lakes Air	NaN	NaN	NaN
10				Gryphon Airlines	NaN	Y3	VOS
11				IAero Airways	NaN	WQ	SWQ
12				IBC Airways	NaN	II	CSQ
13				L-3 Flight International Aviation	NaN	NaN	RTD
14				Liberty Jet Management	NaN	NaN	LRT
15				NetJets	NaN	1I	EJA
16				Omni Air International	NaN	X9	OAE
17				Omni Air Transport	NaN	NaN	DRL
18				Pacific Coast Jet	NaN	NaN	PXT
19				Pentastar Aviation	NaN	NaN	DCX
20				Phoenix Air	NaN	NaN	PHA
21				PlaneSense	NaN	NaN	CNS
22				Presidential Airways	NaN	NaN	PRD
23				Sierra Pacific Airlines	NaN	SI	SPA
24				Skymax	NaN	NaN	SMX
25				Songbird Airways	NaN	SK	SGB
26				Stampede Aviation	NaN	NaN	NaN
27				Superior Air Charter	NaN	NaN	RSP
28				Superior Aviation	NaN	SO	HKA
29				Talkeetna Air Taxi	NaN	NaN	NaN
30				Tropic Ocean Airways	NaN	NaN	NaN
31				World Atlantic Airlines	NaN	K8	WAL
32				XOJET Aviation LLC	NaN	NaN	XOJ

	Primary Hubs, Secondary Hubs	Founded	\
0		NaN	NaN
1	Fort Lauderdale	1987.0	

2		NaN	2000.0
3		San Marcos	1983.0
4		Sheridan	1947.0
5		Cleveland-Lakefront	1997.0
6		Fort Lauderdale-Executive	2009.0
7		Long Island/Islip	1993.0
8	Atlantic City	Las Vegas	Miami 2019.0
9		St. Ignace	NaN
10		NaN	NaN
11		Miami	1997.0
12		Fort Lauderdale	1991.0
13		Newport News	1972.0
14		Long Island/Islip	2006.0
15		Columbus	1964.0
16		Tulsa	1993.0
17		Tulsa	NaN
18		NaN	2006.0
19		Waterford	1964.0
20		Cartersville	1978.0
21		Portsmouth (NH)	1992.0
22		Melbourne/Orlando	NaN
23		Tucson	1970.0
24		Fort Lauderdale	1997.0
25		Miami	1990.0
26		Healy/Denali NP	2011.0
27		NaN	2006.0
28		Lansing	1979.0
29		Talkeetna	1947.0
30		Fort Lauderdale	2009.0
31		Miami	2002.0
32		Sacramento-McClellan	2006.0

	Notes
0	NaN
1	NaN
2	Founded as Executive Flight Services
3	NaN
4	NaN
5	NaN
6	NaN
7	NaN
8	NaN
9	NaN
10	NaN
11	Founded as Swift Air
12	NaN
13	NaN
14	NaN

29	National Airlines	NaN	N8	NCR	NATIONAL CARGO
30	Northern Air Cargo	NaN	NC	NAC	YUKON
31	Polar Air Cargo	NaN	PO	PAC	POLAR
32	Royal Air Freight	NaN	NaN	RAX	AIR ROYAL
33	Ryan Air Services	NaN	7S	RYA	RYAN AIR
34	Sky Lease Cargo	NaN	GG	KYE	SKY CUBE
35	Skyway Enterprises	NaN	KI	SKZ	SKYWAY-INC
36	Strat Air	NaN	NaN	NaN	NaN
37	Tepper Aviation	NaN	NaN	NaN	NaN
38	Trans Executive Airlines	NaN	KH	MUI	RHOADES EXPRESS
39	UPS Airlines	NaN	5X	UPS	UPS
40	USA Jet Airlines	NaN	UJ	JUS	JET USA
41	West Air	NaN	NaN	PCM	PAC VALLEY
42	Western Global Airlines	NaN	KD	WGN	WESTERN GLOBAL
43	Wiggins Airways	NaN	WG	WIG	WIGGINS AIRWAYS

	Primary Hubs,	Secondary Hubs	Founded \
0		Miami	2014.0
1	Wilmington (OH)	Cincinnati	1980.0
2		Milwaukee	1986.0
3		Columbus-Rickenbacker	1974.0
4		Wilmington (OH)	1978.0
5		Anchorage	1996.0
6		Honolulu	1946.0
7		Provo	1971.0
8	Fort Worth/Alliance	Cincinnati	2015.0
9		Dallas/Fort Worth	1968.0
10		Miami	1974.0
11	Dallas-Addison	El Paso	2000.0
12		Guam	1998.0
13	New York-JFK	Anchorage	1992.0
14		Bemidji	1946.0
15		Akron/Canton	1986.0
16		Billings	1981.0
17		Iron Mountain	1998.0
18		Coeur d' Alene	1977.0
19		Fairbanks	1995.0
20	Memphis	Anchorage	1971.0
21		Milwaukee	1985.0
22		Waterford	1983.0
23	Ypsilanti	Anchorage	1967.0
24		Ypsilanti	NaN
25		Anchorage	1995.0
26		Addison	1978.0
27		Billings	1983.0
28		Kinston	1974.0
29		Orlando/Sanford	1985.0
30		Anchorage	1956.0

31	AnchorageCincinnatiHong KongHonoluluLos Angeles...	1993.0
32	Waterford	1961.0
33	AnchorageAniakBethelEmmonakKotzebueNomeSt. Mar...	1953.0
34	Miami	1969.0
35	NaN	1981.0
36	Miami	2018.0
37	Crestview	1987.0
38	Honolulu	1982.0
39	LouisvilleChicago/RockfordCologne/BonnColumbia...	1988.0
40	YpsilantiLaredo	1994.0
41	Las VegasOaklandOntarioSacramentoSan Diego	1988.0
42	Miami Liege, Belgium; AnchorageFort Myers, FL	2013.0
43	Manchester	1929.0

	Notes
0	NaN
1	Founded as Airborne Express. Operates some Ama...
2	Commenced operations in 1980.
3	Founded as Financial Air Express.
4	Founded as US Airways and commenced operations...
5	NaN
6	Founded as Trans-Pacific Airlines and separate...
7	NaN
8	Formerly Amazon Prime Air
9	Founded as California Air Charter.
10	NaN
11	NaN
12	NaN
13	Commenced operations in 1993. Operates some Am...
14	Commenced operations in 1947.
15	NaN
16	NaN
17	NaN
18	NaN
19	NaN
20	Founded as Federal Express and commenced opera...
21	NaN
22	Founded as Air Contract Cargo.
23	Founded as American International Airways.
24	NaN
25	NaN
26	NaN
27	NaN
28	NaN
29	Commenced operations in 1986.
30	NaN
31	NaN
32	NaN

33 Founded as Unalakleet Air Taxi.

Airline	Image	IATA	ICAO	Callsign	\
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Primary Hubs, Secondary Hubs Founded

Airline	Image	IATA	ICAO	\
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Callsign	Primary Hubs	Secondary Hubs	Founded	\
1				
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Notes

9	E
10	F
11	G
12	H
13	I
14	J
15	K
16	L
17	M
18	N
19	O
20	P
21	Q
22	R
23	S
24	T
25	U Uganda Ukraine United Arab Emirates United K...
26	U
27	V
28	W
29	Y
30	Z
31	See also

vteLists of airlines.1

0	All 0-9 A B C D E F G H I J K L M N O P Q R S ...
1	Africa Americas Asia Europe Oceania
2	vteExpand for full listA Abkhazia Afghanistan ...
3	vteExpand for full list
4	A Abkhazia Afghanistan Akrotiri and Dhekelia Å...
5	Abkhazia Afghanistan Akrotiri and Dhekelia Åla...
6	The Bahamas Bahrain Bangladesh Barbados Belaru...
7	Cambodia Cameroon Canada Cape Verde Cayman Isl...
8	Denmark Dhekelia Djibouti Dominica Dominican R...
9	East Timor Ecuador Egypt El Salvador Equatoria...
10	Falkland Islands Faroe Islands Fiji Finland Fr...
11	Gabon The Gambia Georgia Germany Ghana Gibralt...
12	Haiti Honduras Hong Kong Hungary
13	Iceland India Indonesia Iran Iraq Ireland Isra...
14	Jamaica Japan Jersey Jordan
15	Kazakhstan Kenya Kiribati North Korea South Ko...
16	Laos Latvia Lebanon Lesotho Liberia Libya Liec...
17	Macau Macedonia, Republic of Madagascar Malawi...
18	Namibia Nauru Nepal Netherlands Netherlands An...
19	Oman
20	Pakistan Palau Palestine Panama Papua New Guin...
21	Qatar
22	Romania Russia Rwanda

23 Sahrawi Arab Democratic Republic Saint Barthél...
 24 Taiwan Tajikistan Tanzania Thailand Togo Tokel...
 25 U Uganda Ukraine United Arab Emirates United K...
 26 Uganda Ukraine United Arab Emirates United Kin...
 27 Vanuatu Vatican City Venezuela Vietnam British...
 28 Wallis and Futuna
 29 Yemen
 30 Zambia Zimbabwe
 31 List of airline holding companies List of airl... ,
 vteExpand for full list \

0	A Abkhazia Afghanistan Akrotiri and Dhekelia Å...	A
1		A
2		B
3		C
4		D
5		E
6		F
7		G
8		H
9		I
10		J
11		K
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13		M
14		N
15		O
16		P
17		Q
18		R
19		S
20		T
21	U Uganda Ukraine United Arab Emirates United K...	U
22		U
23		V
24		W
25		Y
26		Z

vteExpand for full list.1

0	A Abkhazia Afghanistan Akrotiri and Dhekelia Å...
1	Abkhazia Afghanistan Akrotiri and Dhekelia Åla...
2	The Bahamas Bahrain Bangladesh Barbados Belaru...
3	Cambodia Cameroon Canada Cape Verde Cayman Isl...
4	Denmark Dhekelia Djibouti Dominica Dominican R...
5	East Timor Ecuador Egypt El Salvador Equatoria...
6	Falkland Islands Faroe Islands Fiji Finland Fr...
7	Gabon The Gambia Georgia Germany Ghana Gibralt...
8	Haiti Honduras Hong Kong Hungary

9 Iceland India Indonesia Iran Iraq Ireland Isra...
 10 Jamaica Japan Jersey Jordan
 11 Kazakhstan Kenya Kiribati North Korea South Ko...
 12 Laos Latvia Lebanon Lesotho Liberia Libya Liec...
 13 Macau Macedonia, Republic of Madagascar Malawi...
 14 Namibia Nauru Nepal Netherlands Netherlands An...
 15 Oman
 16 Pakistan Palau Palestine Panama Papua New Guin...
 17 Qatar
 18 Romania Russia Rwanda
 19 Sahrawi Arab Democratic Republic Saint Barthél...
 20 Taiwan Tajikistan Tanzania Thailand Togo Tokel...
 21 U Uganda Ukraine United Arab Emirates United K...
 22 Uganda Ukraine United Arab Emirates United Kin...
 23 Vanuatu Vatican City Venezuela Vietnam British...
 24 Wallis and Futuna
 25 Yemen
 26 Zambia Zimbabwe , 0

1
 0 A Abkhazia Afghanistan Akrotiri and Dhekelia Åla...
 1 B The Bahamas Bahrain Bangladesh Barbados Belaru...
 2 C Cambodia Cameroon Canada Cape Verde Cayman Isl...
 3 D Denmark Dhekelia Djibouti Dominica Dominican R...
 4 E East Timor Ecuador Egypt El Salvador Equatoria...
 5 F Falkland Islands Faroe Islands Fiji Finland Fr...
 6 G Gabon The Gambia Georgia Germany Ghana Gibralt...
 7 H Haiti Honduras Hong Kong Hungary
 8 I Iceland India Indonesia Iran Iraq Ireland Isra...
 9 J Jamaica Japan Jersey Jordan
 10 K Kazakhstan Kenya Kiribati North Korea South Ko...
 11 L Laos Latvia Lebanon Lesotho Liberia Libya Liec...
 12 M Macau Macedonia, Republic of Madagascar Malawi...
 13 N Namibia Nauru Nepal Netherlands Netherlands An...
 14 O Oman
 15 P Pakistan Palau Palestine Panama Papua New Guin...
 16 Q Qatar
 17 R Romania Russia Rwanda
 18 S Sahrawi Arab Democratic Republic Saint Barthél...
 19 T Taiwan Tajikistan Tanzania Thailand Togo Tokel..., 0
 1
 0 U Uganda Ukraine United Arab Emirates United Kin...
 1 V Vanuatu Vatican City Venezuela Vietnam British...
 2 W Wallis and Futuna
 3 Y Yemen
 4 Z Zambia Zimbabwe,

vteAirlines of the United States \

0 Mainline
 1 Regional

```

2             Affiliated
3             Independent
4             Cargo
5             Charter
6             Air taxi and tours
7             Air ambulance
8             Government
9 List of airline holding companies List of airl...

           vteAirlines of the United States.1
0 Alaska Airlines Allegiant Air American Airline...
1 Affiliated Air Wisconsin CommutAir Endeavor Ai...
2 Air Wisconsin CommutAir Endeavor Air Envoy Air...
3 Advanced Air Air Flamenco Air Sunshine Bering ...
4 ABX Air Air Cargo Carriers Air Transport Inter...
5 Air Charter Bahamas Airstream Jets Alerion Avi...
6 Gem Air Grand Canyon Scenic Airlines Griffing ...
7 Air Evac Lifeteam AirMed International Air Met...
8             Comco Janet JPATS Patriot Express
9 List of airline holding companies List of airl... , 0
1
0 Affiliated Air Wisconsin CommutAir Endeavor Air Envoy Air...
1 Independent Advanced Air Air Flamenco Air Sunshine Bering ...,
vteList of airlines of the Americas \
0 North America Caribbean Central America Northe...
1             Sovereign states
2             Dependencies andother territories

           vteList of airlines of the Americas.1 \
0 North America Caribbean Central America Northe...
1 Antigua and Barbuda Argentina Bahamas Barbados...
2 Anguilla Aruba Bermuda Bonaire British Virgin ...

           vteList of airlines of the Americas.2
0 North America Caribbean Central America Northe...
1             NaN
2             NaN ,
0             1
0 Authority control: National libraries Israel United States]

```

[26]: tables[0]

```

[26]:
      Airline  Image IATA ICAO      Callsign \
0   Alaska Airlines   NaN  AS  ASA      ALASKA
1   Allegiant Air     NaN  G4  AAY    ALLEGIANT
2   American Airlines  NaN  AA  AAL    AMERICAN
3   Avelo Airlines    NaN  XP  VXP     AVELO

```

4	Breeze Airways	NaN	MX	MXY	MOXY
5	Delta Air Lines	NaN	DL	DAL	DELTA
6	Eastern Airlines	NaN	2D	EAL	EASTERN
7	Frontier Airlines	NaN	F9	FFT	FRONTIER FLIGHT
8	Hawaiian Airlines	NaN	HA	HAL	HAWAIIAN
9	JetBlue	NaN	B6	JBU	JETBLUE
10	Southwest Airlines	NaN	WN	SWA	SOUTHWEST
11	Spirit Airlines	NaN	NK	NKS	SPIRIT WINGS
12	Sun Country Airlines	NaN	SY	SCX	SUN COUNTRY
13	United Airlines	NaN	UA	UAL	UNITED

	Primary hubs	Secondary hubs	Founded	\
0	Seattle/Tacoma	AnchoragePortland (OR)San Franci...	1932	
1	Las Vegas	CincinnatiFort Walton BeachIndianapol...	1997	
2	Dallas/Fort Worth	CharlotteChicago-O'HareLos An...	1926	
3		BurbankNew HavenOrlando	1987	
4	Charleston	HartfordNew OrleansNorfolkProvoTampa	2018	
5	Atlanta	BostonDetroitLos AngelesMinneapolis/St...	1924	
6		MiamiNew York-JFK	2010	
7	Denver	AtlantaChicago-O'HareCincinnatiCleveland...	1994	
8		HonoluluKahului	1929	
9	New York-JFK	BostonLos AngelesFort LauderdaleOr...	1998	
10	Dallas-Love	AtlantaBaltimoreChicago-MidwayDenve...	1967	
11	Atlantic City	DetroitLas VegasFort LauderdaleCh...	1980	
12	Minneapolis/St. Paul	Dallas/Fort WorthLas Vegas	1982	
13	Chicago-O'Hare	DenverGuamHouston-Intercontinent...	1926	

	Notes
0	Founded as McGee Airways and commenced operati...
1	Founded as WestJet Express and commenced opera...
2	Founded as American Airways and commenced oper...
3	First did business as Casino Express Airlines ...
4	NaN
5	Founded as Huff Daland Dusters and commenced o...
6	NaN
7	NaN
8	Founded as Inter-Island Airways in early 1929 ...
9	Founded as New Air and commenced operations in...
10	Founded as Air Southwest and commenced operati...
11	Founded as Charter One.
12	Commenced operations in 1983.Operates some Ama...
13	Founded as Varney Air Lines and commenced oper...

[32]: tables[6]

	Airline	Image	IATA	ICAO	\
0	Comco	NaN	NaN	NaN	

1		Janet	NaN	NaN	WWW
2	Justice Prisoner and Alien Transportation System		NaN	NaN	JUD

	Callsign	Primary Hubs,	Secondary Hubs	Founded	\
0	NaN		NaN	2002	
1	JANET		Las Vegas	1972	
2	JUSTICE		Oklahoma City	1980	

	Notes
0	NaN
1	NaN
2	Commenced operations in 1995.

```
[33]: # Lets first merge all wikipedia table.
wiki_table =
↳ [tables[0], tables[1], tables[2], tables[3], tables[4], tables[5], tables[6]]
```

```
[34]: wiki_tables = pd.concat(wiki_table, ignore_index=True)
```

```
[35]: wiki_tables
```

```
[35]:
```

	Airline	Image	IATA	ICAO	\
0	Alaska Airlines	NaN	AS	ASA	
1	Allegiant Air	NaN	G4	AAY	
2	American Airlines	NaN	AA	AAL	
3	Avelo Airlines	NaN	XP	VXP	
4	Breeze Airways	NaN	MX	MXY	
..	
137	Lifestar	NaN	NaN	NaN	
138	Life Lion	NaN	NaN	NaN	
139	Comco	NaN	NaN	NaN	
140	Janet	NaN	NaN	WWW	
141	Justice Prisoner and Alien Transportation System	NaN	NaN	JUD	

	Callsign	Primary hubs,	Secondary hubs	Founded	\
0	ALASKA	Seattle/Tacoma	AnchoragePortland (OR)San Franci...	1932.0	
1	ALLEGiant	Las VegasCincinnati	Fort Walton BeachIndianapol...	1997.0	
2	AMERICAN	Dallas/Fort Worth	CharlotteChicago-O'HareLos An...	1926.0	
3	AVELO		BurbankNew HavenOrlando	1987.0	
4	MOXY	Charleston	HartfordNew OrleansNorfolkProvoTampa	2018.0	
..	
137	NaN		NaN	NaN	
138	NaN		NaN	NaN	
139	NaN		NaN	2002.0	
140	JANET		NaN	1972.0	
141	JUSTICE		NaN	1980.0	


```

Notes \
0    Founded as McGee Airways and commenced operati...
1    Founded as WestJet Express and commenced opera...
2    Founded as American Airways and commenced oper...
3    First did business as Casino Express Airlines ...
4                                         NaN
..                                     ...
137                                     NaN
138                                     NaN
139                                     NaN
140                                     NaN
141                                     Commenced operations in 1995.

```

```

Primary Hubs, Secondary Hubs
0                                         NaN
1                                         NaN
2                                         NaN
3                                         NaN
4                                         NaN
..                                     ...
137                                     NaN
138                                     NaN
139                                     NaN
140                                     Las Vegas
141                                     Oklahoma City

```

```
[142 rows x 9 columns]
```

1.2.2 c. You should then get all the information gathered so far in one place.

First we got only that column from wiki pedia table that we need to merge.

```
[36]: wiki_df = wiki_tables[['IATA', "Founded"]]
      wiki_df
```

```

[36]:   IATA  Founded
0     AS   1932.0
1     G4   1997.0
2     AA   1926.0
3     XP   1987.0
4     MX   2018.0
..    ...     ...
137  NaN     NaN
138  NaN     NaN
139  NaN   2002.0
140  NaN   1972.0

```

```
141 NaN 1980.0
```

```
[142 rows x 2 columns]
```

```
[37]: # Now we gather all the information that we got from wiki pedia link and the
↪data that we have.
df = final_df.merge(wiki_df, left_on='Airline', right_on = "IATA")
```

```
[38]: df
```

```
[38]:
```

	id	Airline	Flight	AirportFrom	AirportTo	DayOfWeek	Time	Length	\
0	4	AA	2466	SFO	DFW	3	20	195	
1	231	AA	526	SFO	DFW	3	360	215	
2	234	AA	552	SFO	MIA	3	360	315	
3	905	AA	810	SFO	ORD	3	385	255	
4	1739	AA	24	SFO	JFK	3	425	325	
...	
434919	497838	9E	4292	LWB	JFK	3	890	110	
434920	516333	9E	4292	LWB	JFK	4	890	110	
434921	534123	9E	4292	LWB	JFK	5	890	110	
434922	69058	9E	3752	ABR	MSP	7	410	76	
434923	189396	9E	3752	ABR	MSP	7	410	76	

	Delay	ident	...	iata_code	airport_ref	airport_ident	length_ft	\
0	1	KSFO	...	SFO	3878	KSFO	7500.0	
1	0	KSFO	...	SFO	3878	KSFO	7500.0	
2	1	KSFO	...	SFO	3878	KSFO	7500.0	
3	0	KSFO	...	SFO	3878	KSFO	7500.0	
4	1	KSFO	...	SFO	3878	KSFO	7500.0	
...	
434919	1	KLWB	...	LWB	20390	KLWB	7004.0	
434920	0	KLWB	...	LWB	20390	KLWB	7004.0	
434921	0	KLWB	...	LWB	20390	KLWB	7004.0	
434922	1	KABR	...	ABR	3358	KABR	6901.0	
434923	0	KABR	...	ABR	3358	KABR	6901.0	

	width_ft	surface	lighted	closed	IATA	Founded
0	200.0	ASP	1	0	AA	1926.0
1	200.0	ASP	1	0	AA	1926.0
2	200.0	ASP	1	0	AA	1926.0
3	200.0	ASP	1	0	AA	1926.0
4	200.0	ASP	1	0	AA	1926.0
...
434919	150.0	ASP	1	0	9E	1985.0
434920	150.0	ASP	1	0	9E	1985.0
434921	150.0	ASP	1	0	9E	1985.0
434922	100.0	CON	1	0	9E	1985.0

```
434923      100.0      CON      1      0      9E      1985.0
```

```
[434924 rows x 26 columns]
```

- 1.2.3 d. The total passenger traffic may also contribute to flight delays. The term hub refers to busy commercial airports. Large hubs are airports that account for at least 1 percent of the total passenger enplanements in the United States. Airports that account for 0.25 percent to 1 percent of total passenger enplanements are considered medium hubs. Pull passenger traffic data from the Wikipedia page given below using web scrapping and collate it in a table.

https://en.wikipedia.org/wiki/List_of_the_busiest_airports_in_the_United_States

```
[39]: # Now lets use the web scrapping to import the data frome the wikipedia.
url2 = "https://en.wikipedia.org/wiki/
↳List_of_the_busiest_airports_in_the_United_States"
table = pd.read_html(url2)
```

```
[40]: table
```

```
[40]: [   Rank(2021)                Airports (large hubs) IATACode \
0         1  Hartsfield-Jackson Atlanta International Airport    ATL
1         2           Los Angeles International Airport    LAX
2         3      Chicago O'Hare International Airport    ORD
3         4  Dallas/Fort Worth International Airport    DFW
4         5      Denver International Airport    DEN
5         6  John F. Kennedy International Airport    JFK
6         7  San Francisco International Airport    SFO
7         8  Seattle-Tacoma International Airport    SEA
8         9      Orlando International Airport    MCO
9        10      Harry Reid International Airport    LAS
10       11  Charlotte-Douglas International Airport    CLT
11       12  Newark Liberty International Airport    EWR
12       13  Phoenix Sky Harbor International Airport    PHX
13       14  George Bush Intercontinental Airport    IAH
14       15      Miami International Airport    MIA
15       16  Boston Logan International Airport    BOS
16       17  Minneapolis-Saint Paul International Airport    MSP
17       18      Detroit Metropolitan Airport    DTW
18       19  Fort Lauderdale-Hollywood International Airport    FLL
19       20      Philadelphia International Airport    PHL
20       21      New York LaGuardia Airport    LGA
21       22  Baltimore/Washington International Airport    BWI
22       23      Salt Lake City International Airport    SLC
23       24      San Diego International Airport    SAN
24       25  Washington Dulles International Airport    IAD
```

25	26	Ronald Reagan Washington National Airport	DCA
26	27	Tampa International Airport	TPA
27	28	Chicago Midway International Airport	MDW

	Major cities served	State	2021	2020[3]	2019[4]	\
0	Atlanta	GA	53485010	20559866	53505795	
1	Los Angeles	CA	42880266	18593421	35778573	
2	Chicago	IL	40872527	16243216	33592945	
3	Dallas/Fort Worth	TX	35764976	14606034	40871223	
4	Denver	CO	33575410	14055777	42939104	
5	New York	NY	31036875	12952869	24199688	
6	San Francisco	CA	27701838	10467728	24562271	
7	Seattle	WA	24962342	10584059	24728361	
8	Orlando	FL	24550287	10531436	22433552	
9	Las Vegas	NV	24411096	8786007	21421031	
10	Charlotte	NC	24181195	9462411	25001762	
11	Newark	NJ	23140821	8682558	21905309	
12	Phoenix	AZ	22411342	8269819	31036655	
13	Houston	TX	21897049	7985474	23160763	
14	Miami	FL	21292994	8015744	17950989	
15	Boston	MA	20679409	7069720	19192917	
16	Minneapolis	MN	19152347	7745057	27779230	
17	Detroit	MI	18122696	6822324	18143040	
18	Fort Lauderdale	FL	17936817	6035452	20699377	
19	Philadelphia	PA	15989906	5753239	12840841	
20	New York	NY	15392968	5753239	16006389	
21	Baltimore & Washington, D.C.	MD	9253561	5451355	13284687	
22	Salt Lake City	UT	12829197	4966775	10978756	
23	San Diego	CA	12641360	4637856	12648692	
24	Washington, D.C.	VA	11857307	4147116	15393601	
25	Arlington	VA	11576617	4236603	10081781	
26	Tampa	FL	10921049	4013995	8935654	
27	Chicago	IL	10064875	3862658	11884117	

	2018[5]	2017[6]	2016[7]	2015[8]	2014[9]	2013[10]	2012[11]	\
0	51865797	50251964	50501858	49340732	46604273	45308407	45798928	
1	32821799	31816933	31283579	31589839	30804567	29038128	28022904	
2	31362941	29809097	28267394	26280043	26000591	25496885	25799841	
3	39873927	38593028	37589899	36305668	33843426	32317835	32171795	
4	42624050	41232432	39636042	36351272	34314197	32425892	31326268	
5	22281949	22011251	21511880	21913166	21537725	21346601	20033816	
6	23202480	21565448	20283541	18759938	17278608	16884524	17159427	
7	23795012	23364393	22833267	21857693	20620248	19946179	19959651	
8	21622580	21185458	20896265	21351504	20344867	19525109	19560870	
9	21021640	20709225	20875813	20986349	19471466	19420089	18987488	
10	24024908	22639124	21887110	20148980	17888080	16690295	16121123	
11	21157398	19603731	20062072	20595881	19772087	18952840	19039000	

12	30620769	29533154	29239151	27782369	26244928	25036358	24520981
13	22797602	21571198	19923009	18684818	17773405	17546506	17055993
14	17612331	15817043	14263270	13061632	12031860	11538140	11445103
15	18361942	18409704	18123844	17634273	16972678	16280835	15943878
16	27790717	26900048	25707101	24190560	22770783	21704626	21284236
17	17436837	17036092	16847135	16255520	15775941	15683523	15599879
18	20006521	18759742	17759044	16290362	15507561	14810153	14293695
19	12226730	11615954	11143738	10634538	10139065	9668048	9579840
20	15292670	14271243	14564419	15101349	14792339	14727945	14589337
21	13371816	12976554	12340972	11738845	11022200	11132731	11186444
22	10368514	9548580	9194994	9150458	8531561	8267752	8218487
23	12174224	11139933	10340164	9985763	9333152	8878772	8686621
24	15058501	14614802	14762593	14319924	13535372	13372269	12818717
25	10678018	10912074	11044387	10830850	10311996	9915646	9436387
26	8017347	6902771	6338517	5715205	5396958	5050989	4797102
27	11621623	11024306	10596942	10363974	10415948	10570993	10816216

2011[12]

0	44414121
1	27518358
2	25667499
3	31892301
4	30528737
5	19022535
6	17250415
7	19872617
8	19750306
9	18342158
10	15971676
11	19306660
12	23664832
13	16814092
14	11332466
15	15895653
16	20056568
17	15716865
18	14180730
19	9701756
20	14883180
21	11067319
22	8174194
23	8465683
24	11989227
25	9134576
26	4673047
27	11044383 ,

Rank(2020)

Airports (medium hubs) IATACode \

0	29	Daniel K. Inouye International Airport	HNL
1	30	Portland International Airport	PDX
2	31	Nashville International Airport	BNA
3	32	Austin-Bergstrom International Airport	AUS
4	33	Dallas Love Field	DAL
5	34	St. Louis Lambert International Airport	STL
6	35	Norman Y. Mineta San Jose International Airport	SJC
7	36	William P. Hobby Airport	HOU
8	37	Raleigh-Durham International Airport	RDU
9	38	Louis Armstrong New Orleans International Airport	MSY
10	39	Oakland International Airport	OAK
11	40	Sacramento International Airport	SMF
12	41	Kansas City International Airport	MCI
13	42	John Wayne Airport	SNA
14	43	Fort Myers International Airport	RSW
15	44	San Antonio International Airport	SAT
16	45	Cleveland Hopkins International Airport	CLE
17	46	Indianapolis International Airport	IND
18	47	Pittsburgh International Airport	PIT
19	48	Luis Munoz Marin International Airport	SJU
20	49	Cincinnati/Northern Kentucky International Air...	CVG
21	50	John Glenn Columbus International Airport	CMH
22	51	Kahului Airport	OGG
23	52	Jacksonville International Airport	JAX
24	53	Palm Beach International Airport	PBI
25	54	General Mitchell International Airport	MKE
26	55	Bradley International Airport	BDL
27	56	Hollywood Burbank Airport	BUR
28	57	Ontario International Airport	ONT
29	58	Ted Stevens Anchorage International Airport	ANC
30	59	Albuquerque International Sunport	ABQ
31	60	Omaha Eppley Airfield	OMA
32	61	Buffalo Niagara International Airport	BUF
33	62	Charleston International Airport	CHS
34	63	Memphis International Airport	MEM
35	64	Richmond International Airport	RIC

	City served	State	2020	2019	2018	2017 \
0	Honolulu	HI	9893930	8408457	8134848.0	7876769.0
1	Portland	OR	9790489	11595454	11367176.0	11506310.0
2	Nashville	TN	8498877	9797408	9940866.0	9435473.0
3	Austin	TX	3141505	8683711	7921797.0	6973115.0
4	Dallas	TX	8069178	7069614	6937061.0	6741870.0
5	St Louis	MO	7750391	9988678	9578505.0	9743989.0
6	San Jose	CA	7676765	7946986	7822274.0	7372805.0
7	Houston	TX	7057139	5144467	4719568.0	4461304.0
8	Raleigh	NC	6907342	6454413	6031630.0	5460526.0

9	New Orleans	LA	6858606	6717105	6565482.0	6005527.0
10	Oakland	CA	6540851	4590117	4033412.0	4163587.0
11	Sacramento	CA	6448496	6919429	6416822.0	5851004.0
12	Kansas City	MO	5748186	7828885	7140616.0	6225148.0
13	Santa Ana	CA	5150294	6560230	6798321.0	6530308.0
14	Fort Myers	FL	5039616	5759419	5935131.0	5744918.0
15	San Antonio	TX	5021156	4894541	4836580.0	4562740.0
16	Cleveland	OH	4884126	4709183	4695040.0	4376432.0
17	Indianapolis	IN	4684958	5022980	4844427.0	4521611.0
18	Pittsburgh	PA	4675836	5153276	5317149.0	5195047.0
19	San Juan	Puerto Rico	4543406	4715947	4670033.0	4327431.0
20	Cincinnati	OH/KY	4396395	4413457	4269258.0	3926158.0
21	Columbus	OH	4155596	4172067	4054572.0	3765007.0
22	Kahului	HI	3776732	3460429	3263042.0	3166532.0
23	Jacksonville	FL	3472501	3479923	3118540.0	2759067.0
24	Palm Beach	FL	3453385	3374073	3548817.0	3452544.0
25	Milwaukee	WI	3358946	2723002	2499171.0	2247645.0
26	Hartford	CT	3318301	2713843	2642901.0	2556188.0
27	Burbank	CA	2985033	3323614	3330734.0	3214976.0
28	Ontario	CA	2717141	3791807	3572133.0	3442189.0
29	Anchorage	AK	2988720	2654702	NaN	NaN
30	Albuquerque	NM	2638245	2455274	2457087.0	2303223.0
31	Omaha	NB	2450981	2318442	NaN	NaN
32	Buffalo	NY	2448241	2057750	NaN	NaN
33	Charleston	SC	2375944	2375660	NaN	NaN
34	Memphis	TN	2307944	2375868	NaN	NaN
35	Richmond	VA	2187142	2142156	NaN	NaN

	2016	2015[7]	2014[1]
0	7554596.0	7040921.0	4522341.0
1	11470854.0	11242375.0	10057794.0
2	9071154.0	8340234.0	7878760.0
3	6095545.0	5797547.0	5219982.0
4	6285181.0	5937944.0	5800726.0
5	9656340.0	9656340.0	9463000.0
6	6793076.0	6239231.0	6108758.0
7	4350650.0	4231134.0	4025959.0
8	4969366.0	4816440.0	4384616.0
9	5569705.0	5329696.0	4870569.0
10	4343354.0	4218785.0	4150828.0
11	5401714.0	4954717.0	4673869.0
12	5321603.0	4885690.0	4621003.0
13	5934639.0	5506672.0	5069257.0
14	5391557.0	5135127.0	4982722.0
15	4205739.0	4083476.0	3686315.0
16	4216766.0	3889567.0	3605908.0
17	4179994.0	4091389.0	4046856.0

18	5217242.0	4945175.0	4584147.0
19	3986114.0	3890677.0	3827860.0
20	3269979.0	3036697.0	2874684.0
21	3567864.0	3312496.0	3115501.0
22	3100624.0	3113485.0	2926242.0
23	2799587.0	2716465.0	2589198.0
24	3383271.0	3229876.0	3228607.0
25	2127387.0	2089801.0	2037346.0
26	2563524.0	2525876.0	2381826.0
27	2982194.0	2926047.0	2913380.0
28	3352813.0	3220753.0	3019338.0
29	NaN	NaN	NaN
30	2127387.0	2046155.0	2020354.0
31	NaN	NaN	NaN
32	NaN	NaN	NaN
33	NaN	NaN	NaN
34	NaN	NaN	NaN
35	NaN	NaN	NaN ,

	Rank	Rank change	Airport name \
	Rank	Rank change	Airport name
0	1	NaN	Hartsfield-Jackson Atlanta International Airport
1	2	2.0	Dallas/Fort Worth International Airport
2	3	2.0	Denver International Airport
3	4	1.0	O'Hare International Airport
4	5	3.0	Los Angeles International Airport
5	6	5.0	Charlotte Douglas International Airport
6	7	2.0	Harry Reid International Airport
7	8	5.0	Phoenix Sky Harbor International Airport
8	9	1.0	Orlando International Airport
9	10	2.0	Seattle-Tacoma International Airport
10	11	3.0	Miami International Airport
11	12	3.0	George Bush Intercontinental Airport
12	13	7.0	John F. Kennedy International Airport
13	14	5.0	Fort Lauderdale-Hollywood International Airport
14	15	8.0	San Francisco International Airport
15	16	4.0	Newark Liberty International Airport
16	17	NaN	Minneapolis-Saint Paul International Airport
17	18	NaN	Detroit Metropolitan Airport
18	19	3.0	General Edward Lawrence Logan International Ai...
19	20	3.0	Salt Lake City International Airport
20	21	1.0	Philadelphia International Airport
21	22	NaN	Baltimore/Washington International Airport
22	23	4.0	Tampa International Airport
23	24	NaN	San Diego International Airport
24	25	4.0	Chicago Midway International Airport
25	26	1.0	Washington Dulles International Airport
26	27	4.0	Nashville International Airport

27	28	7.0	LaGuardia Airport
28	29	4.0	Dallas Love Field
29	30	4.0	Ronald Reagan Washington National Airport
30	31	1.0	Portland International Airport
31	32	4.0	Daniel K. Inouye International Airport
32	33	NaN	William P. Hobby Airport
33	34	2.0	Austin-Bergstrom International Airport
34	35	1.0	St. Louis Lambert International Airport

		Location	IATA Code	Traffic	Aircraft	
\		Location	IATA Code	Passengers	% chg. 2019/20	Movements
0		College Park, Georgia	ATL	42918685	61.2	NaN
1		Irving, Texas	DFW	39364990	47.6	NaN
2		Denver, Colorado	DEN	33741129	51.1	NaN
3		Chicago, Illinois	ORD	30860251	63.5	NaN
4		Los Angeles, California	LAX	28779527	67.3	NaN
5		Charlotte, North Carolina	CLT	27205082	45.8	NaN
6		Paradise, Nevada	LAS	22201479	56.9	NaN
7		Phoenix, Arizona	PHX	21978708	52.5	NaN
8		Orlando, Florida	MCO	21617803	57.3	NaN
9		SeaTac, Washington	SEA	20061507	61.3	NaN
10		Miami, Florida	MIA	18663858	59.4	NaN
11		Houston, Texas	IAH	18213571	59.8	NaN
12		Queens, New York	JFK	16630642	73.4	NaN
13		Fort Lauderdale, Florida	FLL	16484132	55.1	NaN
14		San Mateo County, California	SFO	16409625	71.5	NaN
15		Newark, New Jersey	EWK	15892892	65.7	NaN
16		Minneapolis, Minnesota	MSP	14851289	59.8	NaN
17		Romulus, Michigan	DTW	14105007	61.6	NaN
18		Boston, Massachusetts	BOS	12618128	70.3	NaN
19		Salt Lake City, Utah	SLC	12559026	53.2	NaN
20		Philadelphia, Pennsylvania	PHL	11865006	64.1	NaN
21		Linthicum Heights, Maryland	BWI	11204511	58.5	NaN
22		Tampa, Florida	TPA	10238151	54.5	NaN
23		San Diego, California	SAN	8991533	64.3	NaN
24		Chicago, Illinois	MDW	8853948	57.5	NaN
25		Dulles, Virginia	IAD	8333460	66.4	NaN
26		Nashville, Tennessee	BNA	8284570	54.7	NaN
27		Queens, New York	LGA	8245192	73.5	NaN
28		Dallas, Texas	DAL	7684653	54.2	NaN
29		Arlington, Virginia	DCA	7574966	68.4	NaN
30		Portland, Oregon	PDX	7084543	64.4	NaN
31		Honolulu, Hawaii	HNL	6656825	69.6	NaN
32		Houston, Texas	HOU	6476309	55.2	NaN
33		Austin, Texas	AUS	6472579	62.7	NaN
34		St Louis, Missouri	STL	6302402	60.3	NaN

	% chg.2019/20
0	0.0
1	NaN
2	NaN
3	NaN
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5	NaN
6	NaN
7	NaN
8	NaN
9	NaN
10	NaN
11	NaN
12	NaN
13	NaN
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15	NaN
16	NaN
17	NaN
18	NaN
19	NaN
20	NaN
21	NaN
22	NaN
23	NaN
24	NaN
25	NaN
26	NaN
27	NaN
28	NaN
29	NaN
30	NaN
31	NaN
32	NaN
33	NaN
34	NaN ,

Location of 35 busiest airports in the United States

	.mw-parser-output .locmap .od{position:absolut...			Airport name \
	Rank	Rank change		Airport name
0	1	NaN	Hartsfield-Jackson Atlanta International Airport	
1	2	NaN	Los Angeles International Airport[13]	
2	3	NaN	O'Hare International Airport	
3	4	NaN	Dallas/Fort Worth International Airport	
4	5	NaN	Denver International Airport	

5	6	NaN	John F. Kennedy International Airport[14]
6	7	NaN	San Francisco International Airport
7	8	NaN	Seattle-Tacoma International Airport[15]
8	9	NaN	Harry Reid International Airport[16]
9	10	NaN	Orlando International Airport
10	11	NaN	Charlotte Douglas International Airport
11	12	NaN	Newark Liberty International Airport[17]
12	13	1.0	Phoenix Sky Harbor International Airport[18]
13	14	1.0	Miami International Airport
14	15	NaN	George Bush Intercontinental Airport[19]
15	16	NaN	General Edward Lawrence Logan International Ai...
16	17	NaN	Minneapolis-Saint Paul International Airport[21]
17	18	1.0	Detroit Metropolitan Airport[22]
18	19	1.0	Fort Lauderdale-Hollywood International Airpor...
19	20	NaN	Philadelphia International Airport
20	21	NaN	LaGuardia Airport[24]
21	22	NaN	Baltimore/Washington International Airport
22	23	NaN	Salt Lake City International Airport[25]
23	24	NaN	San Diego International Airport[26]
24	25	NaN	Washington Dulles International Airport
25	26	NaN	Ronald Reagan Washington National Airport
26	27	1.0	Tampa International Airport[27]
27	28	1.0	Daniel K. Inouye International Airport[28]
28	29	2.0	Chicago Midway International Airport
29	30	NaN	Portland International Airport[29]
30	31	1.0	Nashville International Airport[30]
31	32	1.0	Austin-Bergstrom International Airport
32	33	2.0	Dallas Love Field[31]
33	34	NaN	St. Louis Lambert International Airport[32]
34	35	NaN	Norman Y. Mineta San Jose International Airpor...

		Location	IATA Code	Traffic	Aircraft	
\		Location	IATA Code	Passengers	% chg.2018/19	Movements
0		College Park, Georgia	ATL	110531300	2.3	904301.0
1		Los Angeles, California	LAX	88068013	0.6	691257.0
2		Chicago, Illinois	ORD	84649115	1.7	919704.0
3		Irving, Texas	DFW	75066956	8.6	720007.0
4		Denver, Colorado	DEN	69015703	7.0	640098.0
5		Queens, New York	JFK	62551072	1.5	456060.0
6		San Mateo County, California	SFO	57488023	0.5	458496.0
7		SeaTac, Washington	SEA	51829239	4.0	450487.0
8		Paradise, Nevada	LAS	51537638	3.7	552962.0
9		Orlando, Florida	MCO	50613072	6.1	357689.0
10		Charlotte, North Carolina	CLT	50168783	8.0	578263.0
11		Newark, New Jersey	EWK	46336452	1.0	446320.0
12		Phoenix, Arizona	PHX	46288337	3.0	438891.0

13	Miami, Florida	MIA	45924466	2.0	416773.0
14	Houston, Texas	IAH	45264059	3.3	478070.0
15	Boston, Massachusetts	BOS	42522411	3.9	427176.0
16	Minneapolis, Minnesota	MSP	39555035	4.0	406076.0
17	Romulus, Michigan	DTW	36769279	4.3	396909.0
18	Fort Lauderdale, Florida	FLL	36747622	2.2	331447.0
19	Philadelphia, Pennsylvania	PHL	33018886	4.2	390321.0
20	Queens, New York	LGA	31084894	3.3	373356.0
21	Linthicum Heights, Maryland	BWI	26993896	0.6	262597.0
22	Salt Lake City, Utah	SLC	26808014	4.9	344715.0
23	San Diego, California	SAN	25216947	4.0	231354.0
24	Dulles, Virginia	IAD	24817677	3.1	285042.0
25	Arlington, Virginia	DCA	23945527	1.8	292682.0
26	Tampa, Florida	TPA	22497953	5.7	217360.0
27	Honolulu, Hawaii	HNL	21870691	4.2	326832.0
28	Chicago, Illinois	MDW	20844860	5.4	232084.0
29	Portland, Oregon	PDX	19891365	0.0	238384.0
30	Nashville, Tennessee	BNA	18273434	14.2	NaN
31	Austin, Texas	AUS	17343729	9.6	209726.0
32	Dallas, Texas	DAL	16780158	3.4	231879.0
33	St Louis, Missouri	STL	15878527	1.6	193925.0
34	San Jose, California	SJC	15650444	9.3	207111.0

	% chg.2018/19
0	1.0
1	2.3
2	1.8
3	7.9
4	6.1
5	0.1
6	2.5
7	2.8
8	2.4
9	2.9
10	5.1
11	1.6
12	1.1
13	0.2
14	2.4
15	0.7
16	0.3
17	0.8
18	0.6
19	2.8
20	0.4
21	1.5

22	2.2
23	2.8
24	3.9
25	0.4
26	5.0
27	10.7
28	4.6
29	1.9
30	NaN
31	0.2
32	0.3
33	0.2
34	19.4 ,

	Rank	Rank	change	Airport name \
	Rank	Rank	change	Airport name
0	1	NaN	Hartsfield-Jackson Atlanta International Airpo...	
1	2	NaN	Los Angeles International Airport[35]	
2	3	NaN	O'Hare International Airport[36]	
3	4	NaN	Dallas/Fort Worth International Airport[37]	
4	5	NaN	Denver International Airport[38]	
5	6	NaN	John F. Kennedy International Airport[39]	
6	7	NaN	San Francisco International Airport[40]	
7	8	1.0	Seattle-Tacoma International Airport[41]	
8	9	1.0	Harry Reid International Airport[42]	
9	10	2.0	Orlando International Airport[43]	
10	11	1.0	Charlotte Douglas International Airport[44]	
11	12	1.0	Newark Liberty International Airport[45]	
12	13	1.0	Miami International Airport[46]	
13	14	1.0	Phoenix Sky Harbor International Airport[47]	
14	15	NaN	George Bush Intercontinental Airport[48]	
15	16	NaN	General Edward Lawrence Logan International Ai...	
16	17	NaN	Minneapolis-Saint Paul International Airport[50]	
17	18	1.0	Fort Lauderdale-Hollywood International Airpor...	
18	19	1.0	Detroit Metropolitan Airport[52]	
19	20	NaN	Philadelphia International Airport[53]	
20	21	NaN	LaGuardia Airport[54]	
21	22	NaN	Baltimore/Washington International Airport[55]	
22	23	NaN	Salt Lake City International Airport[56]	
23	24	2.0	San Diego International Airport[57]	
24	25	NaN	Washington Dulles International Airport[58]	
25	26	2.0	Ronald Reagan Washington National Airport[59]	
26	27	NaN	Chicago Midway International Airport[60]	
27	28	NaN	Tampa International Airport[61]	
28	29	NaN	Daniel K. Inouye International Airport	
29	30	NaN	Portland International Airport[62]	
30	31	NaN	Dallas Love Field[63]	
31	32	1.0	Nashville International Airport[64]	

32	33	1.0	Austin-Bergstrom International Airport[65]
33	34	2.0	St. Louis Lambert International Airport[66]
34	35	NaN	Norman Y. Mineta San Jose International Airpor...

	Location	IATA Code	Traffic	
	Location	IATA Code	Passengers	% chg.2017/18
0	College Park, Georgia	ATL	107394029	3.3
1	Los Angeles, California	LAX	87534384	3.5
2	Chicago, Illinois	ORD	83245472	4.3
3	Irving, Texas	DFW	69112607	3.0
4	Denver, Colorado	DEN	64494613	5.1
5	Queens, New York	JFK	61909148	3.9
6	South San Francisco, California	SFO	57793313	3.5
7	SeaTac, Washington	SEA	49849520	6.2
8	Las Vegas, Nevada	LAS	49716584	2.5
9	Orlando, Florida	MCO	47696627	6.9
10	Charlotte, North Carolina	CLT	46444380	1.2
11	Newark, New Jersey	EWR	46065175	6.6
12	Miami, Florida	MIA	45044312	2.2
13	Phoenix, Arizona	PHX	44943686	2.3
14	Houston, Texas	IAH	43807539	7.6
15	Boston, Massachusetts	BOS	40941925	6.6
16	Minneapolis, Minnesota	MSP	38037381	0.0
17	Fort Lauderdale, Florida	FLL	35963370	10.6
18	Romulus, Michigan	DTW	35236676	1.5
19	Philadelphia, Pennsylvania	PHL	31691956	7.1
20	Queens, New York	LGA	30094074	1.8
21	Linthicum Heights, Maryland	BWI	27145831	2.9
22	Salt Lake City, Utah	SLC	25554244	5.6
23	San Diego, California	SAN	24238300	9.3
24	Dulles, Virginia	IAD	24060709	5.1
25	Arlington, Virginia	DCA	23464618	1.8
26	Chicago, Illinois	MDW	22027737	1.9
27	Tampa, Florida	TPA	21289390	8.5
28	Honolulu, Hawaii	HNL	20990932	1.1
29	Portland, Oregon	PDX	19882788	4.2
30	Dallas, Texas	DAL	16229151	3.2
31	Nashville, Tennessee	BNA	15996194	13.2
32	Austin, Texas	AUS	15819912	13.9
33	St Louis, Missouri	STL	15632586	5.9
34	San Jose, California	SJC	14319292	14.7

Aircraft		
	Movements	% chg.2017/18
0	895682	01.7
1	707833	01.1
2	903747	04.2

3	667213	02.0
4	603403	03.6
5	455529	01.6
6	470164	02.1
7	438391	05.4
8	539857	00.6
9	347672	05.1
10	550013	00.4
11	458674	04.6
12	416032	00.7
13	434252	00.8
14	466738	03.6
15	424024	05.6
16	407476	02.1
17	329662	05.4
18	393681	00.4
19	379665	02.6
20	372025	00.8
21	266569	01.9
22	337276	03.1
23	225058	07.5
24	274281	03.6
25	293827	00.2
26	243322	03.2
27	206938	05.9
28	295233	5.30
29	233993	02.2
30	231110	01.6
31	216966	05.2
32	210080	05.2
33	-	-
34	173389	011.3 ,

	Rank	Airport name \
	Rank	Airport name
0	1	Hartsfield-Jackson Atlanta International Airport
1	2	Los Angeles International Airport
2	3	O'Hare International Airport
3	4	Dallas/Fort Worth International Airport
4	5	John F. Kennedy International Airport
5	6	Denver International Airport
6	7	San Francisco International Airport
7	8	Harry Reid International Airport
8	9	Seattle-Tacoma International Airport
9	10	Miami International Airport
10	11	Charlotte Douglas International Airport
11	12	Phoenix Sky Harbor International Airport
12	13	Orlando International Airport

13	14	George Bush Intercontinental Airport
14	15	Newark Liberty International Airport
15	16	Minneapolis-Saint Paul International Airport
16	17	General Edward Lawrence Logan International Ai...
17	18	Detroit Metropolitan Airport
18	19	Philadelphia International Airport
19	20	LaGuardia Airport
20	21	Fort Lauderdale-Hollywood International Airport
21	22	Baltimore/Washington International Airport
22	23	Ronald Reagan Washington National Airport
23	24	Salt Lake City International Airport
24	25	Chicago Midway International Airport
25	26	Washington Dulles International Airport
26	27	San Diego International Airport
27	28	Honolulu International Airport
28	29	Tampa International Airport
29	30	Portland International Airport
30	31	Dallas Love Field
31	32	St. Louis Lambert International Airport
32	33	Nashville International Airport
33	34	William P. Hobby Airport
34	35	Austin-Bergstrom International Airport
35	36	Oakland International Airport

	Location	IATA Code	Traffic	
	Location	IATA Code	Passengers	% chg.2015/16
0	College Park, Georgia	ATL	104171935	02.6
1	Los Angeles, California	LAX	80921527	08.0
2	Chicago, Illinois	ORD	77960588	01.3
3	Irving, Texas	DFW	65670697	00.2
4	Queens, New York	JFK	59105513	03.9
5	Denver, Colorado	DEN	58266515	07.9
6	South San Francisco, California	SFO	53099282	06.1
7	Las Vegas, Nevada	LAS	47496614	04.5
8	SeaTac, Washington	SEA	45736700	08.0
9	Miami, Florida	MIA	44584603	00.5
10	Charlotte, North Carolina	CLT	44422022	01.0
11	Phoenix, Arizona	PHX	43302381	01.6
12	Orlando, Florida	MCO	41923399	08.0
13	Houston, Texas	IAH	41622594	03.3
14	Newark, New Jersey	EWR	40563285	08.2
15	Minneapolis, Minnesota	MSP	37413728	02.3
16	Boston, Massachusetts	BOS	36356917	08.5
17	Romulus, Michigan	DTW	34401254	02.9
18	Philadelphia, Pennsylvania	PHL	30155090	04.1
19	Queens, New York	LGA	29786769	04.7
20	Fort Lauderdale, Florida	FLL	29205002	08.4

21	Linthicum Heights, Maryland	BWI	25122651	05.4
22	Arlington, Virginia	DCA	23568586	02.4
23	Salt Lake City, Utah	SLC	23157445	04.5
24	Chicago, Illinois	MDW	22677859	02.1
25	Dulles, Virginia	IAD	21817340	01.5
26	San Diego, California	SAN	20725801	03.2
27	Honolulu, Hawaii	HNL	19878659	- 00.0
28	Tampa, Florida	TPA	18931922	00.6
29	Portland, Oregon	PDX	18352767	08.9
30	Dallas, Texas	DAL	15562738	07.3
31	St Louis, Missouri	STL	13959126	09.5
32	Nashville, Tennessee	BNA	12979803	011.2
33	Houston, Texas	HOU	12909075	06.1
34	Austin, Texas	AUS	12436849	04.5
35	Oakland, California	OAK	12070967	07.7

	Aircraft	
	Movements	% chg. 2015/16
0	898356	01.8
1	697138	06.3
2	867635	00.9
3	672748	01.3
4	452415	03.0
5	565503	04.5
6	450388	04.8
7	541428	02.1
8	412170	08.1
9	414234	00.3
10	545742	00.3
11	440643	00.1
12	316981	02.9
13	470780	06.4
14	435907	05.3
15	412872	02.0
16	372930	02.5
17	393427	03.7
18	394022	04.2
19	369987	02.7
20	290239	04.4
21	248585	00.9
22	295038	00.8
23	320137	02.7
24	253046	00.2
25	265743	01.5
26	197132	01.5
27	316154	01.1
28	-	-

29	227709	04.4
30	224193	03.7
31	190560	02.5
32	194758	05.6
33	200741	00.1
34	192032	00.4
35	222771	03.3 ,

	Rank	Airport name \
0	1	John F. Kennedy International Airport
1	2	Miami International Airport
2	3	Los Angeles International Airport
3	4	George Bush Intercontinental Airport
4	5	Newark Liberty International Airport
5	6	Dallas/Fort Worth International Airport
6	7	Hartsfield-Jackson Atlanta International Airport
7	8	O'Hare International Airport
8	9	Fort Lauderdale-Hollywood International Airport
9	10	Washington Dulles International Airport
10	11	San Francisco International Airport
11	12	General Edward Lawrence Logan International Ai...
12	13	Charlotte Douglas International Airport
13	14	Denver International Airport
14	15	Orlando International Airport
15	16	Seattle-Tacoma International Airport
16	17	Phoenix Sky Harbor International Airport
17	18	Philadelphia International Airport
18	19	Detroit Metropolitan Wayne County Airport
19	20	Harry Reid International Airport
20	22	Minneapolis-Saint Paul International Airport
21	38	Daniel K. Inouye International Airport

	Location	IATA Code	2021[68]	2020[69]	2019[70]
0	Queens, New York	JFK	12466165	8219317	33432159
1	Miami, Florida	MIA	11592445	6565834	20735658
2	Los Angeles, California	LAX	7862532	6246602	25210140
3	Houston, Texas	IAH	6458473	3491935	10764589
4	Newark, New Jersey	EWR	6250880	3688541	14087622
5	Irving, Texas	DFW	5852397	3268822	9103438
6	College Park, Georgia	ATL	5474264	3347184	12268779
7	Chicago, Illinois	ORD	5148494	3481860	13412885
8	Fort Lauderdale, Florida	FLL	4016553	2839383	8524251
9	Dulles, Virginia	IAD	3230027	1917510	7990292
10	South San Francisco, California	SFO	3139041	3210024	14357960
11	Boston, Massachusetts	BOS	2046561	1574712	7534504
12	Charlotte, North Carolina	CLT	1989704	1069001	3405907
13	Denver, Colorado	DEN	1856124	934563	3037012
14	Orlando, Florida	MCO	1837706	1525177	6957048

15	SeaTac, Washington	SEA	1393603	1273179	5392147
16	Phoenix, Arizona	PHX	1223856	750138	1958468
17	Philadelphia, Pennsylvania	PHL	988733	682030	3847253
18	Romulus, Michigan	DTW	966375	873744	3717775
19	Paradise, Nevada	LAS	738257	711614	3462627
20	Minneapolis, Minnesota	MSP	673759	835721	3144386
21	Honolulu, Hawaii	HNL	167502	1052697	5207875 ,

Rank		Airport name \
Rank		Airport name
0	1	Memphis International Airport
1	2	Ted Stevens Anchorage International Airport
2	3	Louisville International Airport
3	4	O'Hare International Airport
4	5	Miami International Airport
5	6	Los Angeles International Airport
6	7	Cincinnati/Northern Kentucky International Air...
7	8	Indianapolis International Airport
8	9	Dallas/Fort Worth International Airport
9	10	Ontario International Airport

	Location	IATA code	Cargo	
	Location	IATA code	Ibs.	% chg.2017/16
0	Memphis, Tennessee	MEM	23949525780	00.35%
1	Anchorage, Alaska	ANC	17337337377	02.79%
2	Louisville, Kentucky	SDF	13403682652	04.68%
3	Chicago, Illinois	ORD	10373559593	010.84%
4	Miami, Florida	MIA	7963988407	00.82%
5	Los Angeles, California	LAX	7197930264	03.85%
6	Hebron, Kentucky	CVG	5700282994	033.32%
7	Indianapolis, Indiana	IND	5138500318	0-3.58%
8	Irving, Texas	DFW	4155362297	07.65%
9	Ontario, California	ONT	3522510318	015.81% ,

```
.mw-parser-output .navbar{display:inline;font-size:88%;font-
weight:normal}.mw-parser-output .navbar-collapse{float:left;text-align:left}.mw-
parser-output .navbar-boxtext{word-spacing:0}.mw-parser-output .navbar
ul{display:inline-block;white-space:nowrap;line-height:inherit}.mw-parser-output
.navbar-brackets::before{margin-right:-0.125em;content:"[ "}.mw-parser-output
.navbar-brackets::after{margin-left:-0.125em;content:" ]"}.mw-parser-output
.navbar li{word-spacing:-0.125em}.mw-parser-output .navbar a>span,.mw-parser-
output .navbar a>abbr{text-decoration:inherit}.mw-parser-output .navbar-mini
abbr{font-variant:small-caps;border-bottom:none;text-
decoration:none;cursor:inherit}.mw-parser-output .navbar-ct-full{font-
size:114%;margin:0 7em}.mw-parser-output .navbar-ct-mini{font-size:114%;margin:0
4em}vteMajor airports in the United States \
0 Atlanta (Hartsfield-Jackson - ATL) Baltimore (...
1 Statistics
```

```

.mw-parser-output .navbar{display:inline;font-size:88%;font-
weight:normal}.mw-parser-output .navbar-collapse{float:left;text-align:left}.mw-
parser-output .navbar-boxtext{word-spacing:0}.mw-parser-output .navbar
ul{display:inline-block;white-space:nowrap;line-height:inherit}.mw-parser-output
.navbar-brackets::before{margin-right:-0.125em;content:"[ "}.mw-parser-output
.navbar-brackets::after{margin-left:-0.125em;content:" ]"}.mw-parser-output
.navbar li{word-spacing:-0.125em}.mw-parser-output .navbar a>span,.mw-parser-
output .navbar a>abbr{text-decoration:inherit}.mw-parser-output .navbar-mini
abbr{font-variant:small-caps;border-bottom:none;text-
decoration:none;cursor:inherit}.mw-parser-output .navbar-ct-full{font-
size:114%;margin:0 7em}.mw-parser-output .navbar-ct-mini{font-size:114%;margin:0
4em}vteMajor airports in the United States.1

```

```

0 Atlanta (Hartsfield-Jackson - ATL) Baltimore (...
1 Statistics

```

```

,
vteList of the busiest airports in North America \
0 Sovereign states
1 Dependencies andother territories

```

```

vteList of the busiest airports in North America.1
0 Antigua and Barbuda Bahamas Barbados Belize Ca...
1 Anguilla Aruba Bermuda Bonaire British Virgin ... ,
vteLists of the busiest airports by continent \
0 Africa Asia Europe North America Oceania South...

```

```

vteLists of the busiest airports by continent.1
0 Africa Asia Europe North America Oceania South... ,
vteAviation statistics \
0 Airports worldwide
1 Busiest airports by continent and country
2 Africa
3 Asia
4 Europe
5 North America
6 Oceania
7 South America
8 By region
9 Airlines
10 Routes

```

```

vteAviation statistics.1
0 Busiest airports by continent By aircraft move...
1 Africa Morocco South Africa Asia China (exclud...
2 Morocco South Africa
3 China (excluding Hong Kong and Macau) India In...
4 Austria Belgium Bulgaria Croatia France German...
5 Canada Dominican Republic Mexico United States...

```

```

6                Australia New Zealand
7  Argentina Brazil Chile Colombia Ecuador Paragu...
8  Baltic Caribbean Central America Latin America...
9  World's largest airlines Airline holding compa...
10 Busiest passenger air routes General aviation ... ,
    0                                1
0        Africa                                Morocco South Africa
1        Asia  China (excluding Hong Kong and Macau) India In...
2        Europe  Austria Belgium Bulgaria Croatia France German...
3  North America  Canada Dominican Republic Mexico United States...
4        Oceania                                Australia New Zealand
5  South America  Argentina Brazil Chile Colombia Ecuador Paragu...
6        By region  Baltic Caribbean Central America Latin America...]
```

```
[41]: table[0] = table[0].drop(['2021', '2013[10]', '2012[11]', '2011[12]'], axis=1)
```

```
[42]: table[0].head()
```

```
[42]:
```

	Rank(2021)		Airports (large hubs)	IATACode	\
0	1	Hartsfield-Jackson Atlanta International Airport	ATL		
1	2	Los Angeles International Airport	LAX		
2	3	Chicago O'Hare International Airport	ORD		
3	4	Dallas/Fort Worth International Airport	DFW		
4	5	Denver International Airport	DEN		

	Major cities served	State	2020[3]	2019[4]	2018[5]	2017[6]	2016[7]	\
0	Atlanta	GA	20559866	53505795	51865797	50251964	50501858	
1	Los Angeles	CA	18593421	35778573	32821799	31816933	31283579	
2	Chicago	IL	16243216	33592945	31362941	29809097	28267394	
3	Dallas/Fort Worth	TX	14606034	40871223	39873927	38593028	37589899	
4	Denver	CO	14055777	42939104	42624050	41232432	39636042	

	2015[8]	2014[9]
0	49340732	46604273
1	31589839	30804567
2	26280043	26000591
3	36305668	33843426
4	36351272	34314197

```
[43]: table[0]['traffic_Chg19_20'] = table[0]['2020[3]'] - table[0]['2019[4]']
```

```
[44]: table[0]['traffic_Chg18_19'] = table[0]['2019[4]'] - table[0]['2018[5]']
table[0]['hubs'] = str('large_hub')
```

```
[45]: table[0] = table[0][['IATACode', 'traffic_Chg19_20', 'traffic_Chg18_19',
↪ 'hubs']]
table[0]
```

```
[45]: IATACode  traffic_Chg19_20  traffic_Chg18_19      hubs
0      ATL      -32945929          1639998  large_hub
1      LAX      -17185152          2956774  large_hub
2      ORD      -17349729          2230004  large_hub
3      DFW      -26265189           997296  large_hub
4      DEN      -28883327           315054  large_hub
5      JFK      -11246819          1917739  large_hub
6      SFO      -14094543          1359791  large_hub
7      SEA      -14144302           933349  large_hub
8      MCO      -11902116           810972  large_hub
9      LAS      -12635024           399391  large_hub
10     CLT      -15539351           976854  large_hub
11     EWR      -13222751           747911  large_hub
12     PHX      -22766836           415886  large_hub
13     IAH      -15175289           363161  large_hub
14     MIA      -9935245            338658  large_hub
15     BOS      -12123197           830975  large_hub
16     MSP      -20034173           -11487  large_hub
17     DTW      -11320716           706203  large_hub
18     FLL      -14663925           692856  large_hub
19     PHL      -7087602            614111  large_hub
20     LGA      -10253150           713719  large_hub
21     BWI      -7833332            -87129  large_hub
22     SLC      -6011981            610242  large_hub
23     SAN      -8010836            474468  large_hub
24     IAD      -11246485           335100  large_hub
25     DCA      -5845178           -596237  large_hub
26     TPA      -4921659            918307  large_hub
27     MDW      -8021459            262494  large_hub
```

```
[46]: table[1].head()
```

```
[46]: Rank(2020)      Airports (medium hubs) IATACode City served \
0      29  Daniel K. Inouye International Airport      HNL      Honolulu
1      30      Portland International Airport          PDX      Portland
2      31      Nashville International Airport          BNA      Nashville
3      32  Austin-Bergstrom International Airport      AUS      Austin
4      33      Dallas Love Field                        DAL      Dallas
```

```
State      2020      2019      2018      2017      2016      2015[7] \
0  HI  9893930  8408457  8134848.0  7876769.0  7554596.0  7040921.0
1  OR  9790489  11595454  11367176.0  11506310.0  11470854.0  11242375.0
2  TN  8498877  9797408  9940866.0  9435473.0  9071154.0  8340234.0
3  TX  3141505  8683711  7921797.0  6973115.0  6095545.0  5797547.0
4  TX  8069178  7069614  6937061.0  6741870.0  6285181.0  5937944.0
```

```
2014[1]
```

```

0    4522341.0
1    10057794.0
2     7878760.0
3     5219982.0
4     5800726.0

```

```

[47]: table[1]['traffic_Chg19_20'] = table[1]['2020'] - table[1]['2019']
      table[1]['traffic_Chg18_19'] = table[1]['2019'] - table[1]['2018']
      table[1]['hubs'] = str('Medium_hub')

```

```

[48]: table[1] = table[1][['IATACode', 'traffic_Chg19_20', 'traffic_Chg18_19', 'hubs']]
      table[1]

```

```

[48]:
   IATACode  traffic_Chg19_20  traffic_Chg18_19  hubs
0      HNL         1485473         273609.0  Medium_hub
1      PDX        -1804965         228278.0  Medium_hub
2      BNA        -1298531        -143458.0  Medium_hub
3      AUS        -5542206         761914.0  Medium_hub
4      DAL         999564         132553.0  Medium_hub
5      STL        -2238287         410173.0  Medium_hub
6      SJC        -270221         124712.0  Medium_hub
7      HOU         1912672         424899.0  Medium_hub
8      RDU         452929         422783.0  Medium_hub
9      MSY         141501         151623.0  Medium_hub
10     OAK         1950734         556705.0  Medium_hub
11     SMF        -470933         502607.0  Medium_hub
12     MCI        -2080699         688269.0  Medium_hub
13     SNA        -1409936        -238091.0  Medium_hub
14     RSW        -719803        -175712.0  Medium_hub
15     SAT         126615          57961.0  Medium_hub
16     CLE         174943          14143.0  Medium_hub
17     IND        -338022         178553.0  Medium_hub
18     PIT        -477440        -163873.0  Medium_hub
19     SJU        -172541          45914.0  Medium_hub
20     CVG        -17062         144199.0  Medium_hub
21     CMH        -16471         117495.0  Medium_hub
22     OGG         316303         197387.0  Medium_hub
23     JAX         -7422         361383.0  Medium_hub
24     PBI         79312        -174744.0  Medium_hub
25     MKE         635944         223831.0  Medium_hub
26     BDL         604458          70942.0  Medium_hub
27     BUR        -338581         -7120.0  Medium_hub
28     ONT       -1074666         219674.0  Medium_hub
29     ANC         334018             NaN  Medium_hub
30     ABQ         182971        -1813.0  Medium_hub
31     OMA         132539             NaN  Medium_hub
32     BUF         390491             NaN  Medium_hub

```

33	CHS	284	NaN	Medium_hub
34	MEM	-67924	NaN	Medium_hub
35	RIC	44986	NaN	Medium_hub

```
[49]: # Lets first merge all wikipedia table.
wiki_data = [table[0],table[1]]
wiki_data = pd.concat(wiki_data, ignore_index=True)
wiki_data
```

```
[49]: IATACode  traffic_Chg19_20  traffic_Chg18_19  hubs
0      ATL      -32945929      1639998.0  large_hub
1      LAX      -17185152      2956774.0  large_hub
2      ORD      -17349729      2230004.0  large_hub
3      DFW      -26265189      997296.0   large_hub
4      DEN      -28883327      315054.0   large_hub
..      ...      ...      ...      ...
59     OMA      132539      NaN   Medium_hub
60     BUF      390491      NaN   Medium_hub
61     CHS      284      NaN   Medium_hub
62     MEM      -67924      NaN   Medium_hub
63     RIC      44986      NaN   Medium_hub
```

[64 rows x 4 columns]

```
[50]: # Now we gather all the information that we got from wiki pedia link and the
↪data that we have.
final_df = df.merge(wiki_data, left_on = 'iata_code', right_on = "IATACode")
```

```
[51]: final_df
```

```
[51]:      id Airline  Flight AirportFrom AirportTo  DayOfWeek  Time  Length  \
0      4      AA    2466           SFO         DFW          3    20    195
1     231      AA     526           SFO         DFW          3   360    215
2     234      AA     552           SFO         MIA          3   360    315
3     905      AA     810           SFO         ORD          3   385    255
4    1739      AA      24           SFO         JFK          3   425    325
...    ...    ...    ...    ...    ...    ...    ...    ...
363125 506267      9E    4052          DAL         MEM          4   370     90
363126 512858      9E    3704          DAL         MEM          4   705     92
363127 518247      9E    4060          DAL         MEM          4   990     90
363128 524678      9E    4052          DAL         MEM          5   370     90
363129 530841      9E    3704          DAL         MEM          5   705     92

      Delay ident  ... width_ft surface  lighted  closed  IATA Founded  \
0      1  KSFO  ...   200.0     ASP          1        0   AA   1926.0
1      0  KSFO  ...   200.0     ASP          1        0   AA   1926.0
2      1  KSFO  ...   200.0     ASP          1        0   AA   1926.0
```


3	0	KSFO	...	200.0	ASP	1	0	AA	1926.0
4	1	KSFO	...	200.0	ASP	1	0	AA	1926.0
...
363125	0	KDAL	...	150.0	CON	1	0	9E	1985.0
363126	1	KDAL	...	150.0	CON	1	0	9E	1985.0
363127	0	KDAL	...	150.0	CON	1	0	9E	1985.0
363128	1	KDAL	...	150.0	CON	1	0	9E	1985.0
363129	0	KDAL	...	150.0	CON	1	0	9E	1985.0

	IATACode	traffic_Chg19_20	traffic_Chg18_19	hubs
0	SFO	-14094543	1359791.0	large_hub
1	SFO	-14094543	1359791.0	large_hub
2	SFO	-14094543	1359791.0	large_hub
3	SFO	-14094543	1359791.0	large_hub
4	SFO	-14094543	1359791.0	large_hub
...
363125	DAL	999564	132553.0	Medium_hub
363126	DAL	999564	132553.0	Medium_hub
363127	DAL	999564	132553.0	Medium_hub
363128	DAL	999564	132553.0	Medium_hub
363129	DAL	999564	132553.0	Medium_hub

[363130 rows x 30 columns]

1.2.4 2. You should then examine the missing values in each field, perform missing value treatment, and justify your actions.

```
[52]: # Now we have the final data first we remove some column that is not useable.
final_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 363130 entries, 0 to 363129
Data columns (total 30 columns):
#   Column              Non-Null Count  Dtype
---  -
0   id                  363130 non-null  int64
1   Airline             363130 non-null  object
2   Flight              363130 non-null  int64
3   AirportFrom         363130 non-null  object
4   AirportTo           363130 non-null  object
5   DayOfWeek           363130 non-null  int64
6   Time                363130 non-null  int64
7   Length              363130 non-null  int64
8   Delay               363130 non-null  int64
9   ident               363130 non-null  object
10  type                363130 non-null  object
```

```

11  name                363130 non-null object
12  latitude_deg        363130 non-null float64
13  longitude_deg       363130 non-null float64
14  elevation_ft        363130 non-null float64
15  scheduled_service   363130 non-null object
16  iata_code           363130 non-null object
17  airport_ref         363130 non-null int64
18  airport_ident       363130 non-null object
19  length_ft           363130 non-null float64
20  width_ft            363130 non-null float64
21  surface             363130 non-null object
22  lighted             363130 non-null int64
23  closed              363130 non-null int64
24  IATA                363130 non-null object
25  Founded             363130 non-null float64
26  IATACode            363130 non-null object
27  traffic_Chg19_20    363130 non-null int64
28  traffic_Chg18_19    351555 non-null float64
29  hubs                363130 non-null object
dtypes: float64(7), int64(10), object(13)
memory usage: 85.9+ MB

```

```

[53]: final_df.
      ↪drop(['id', 'AirportFrom', 'airport_ident', 'iata_code', 'AirportTo', 'surface', '
      ↪'ident',
                                     'IATA', 'IATACode', 'name'], axis=1, inplace=True)

```

```

[54]: final_df.info()

```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 363130 entries, 0 to 363129
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Airline               363130 non-null object
1   Flight               363130 non-null int64
2   DayOfWeek            363130 non-null int64
3   Time                 363130 non-null int64
4   Length               363130 non-null int64
5   Delay                363130 non-null int64
6   type                 363130 non-null object
7   latitude_deg         363130 non-null float64
8   longitude_deg        363130 non-null float64
9   elevation_ft         363130 non-null float64
10  scheduled_service    363130 non-null object
11  airport_ref          363130 non-null int64
12  length_ft            363130 non-null float64
13  width_ft             363130 non-null float64

```

```

14 lighted          363130 non-null int64
15 closed          363130 non-null int64
16 Founded          363130 non-null float64
17 traffic_Chg19_20 363130 non-null int64
18 traffic_Chg18_19 351555 non-null float64
19 hubs            363130 non-null object
dtypes: float64(7), int64(9), object(4)
memory usage: 58.2+ MB

```

```
[55]: # Now lets check the null value and treat them.
final_df.isnull().sum()
```

```

[55]: Airline          0
Flight              0
DayOfWeek           0
Time               0
Length             0
Delay              0
type               0
latitude_deg       0
longitude_deg      0
elevation_ft       0
scheduled_service  0
airport_ref        0
length_ft          0
width_ft           0
lighted            0
closed             0
Founded            0
traffic_Chg19_20   0
traffic_Chg18_19   11575
hubs               0
dtype: int64

```

```
[62]: final_df
```

```

[62]:   Airline  Flight  DayOfWeek  Time  Length  Delay  type \
0      AA    2466         3    20    195     1  large_airport
1      AA     526         3   360    215     0  large_airport
2      AA     552         3   360    315     1  large_airport
3      AA     810         3   385    255     0  large_airport
4      AA      24         3   425    325     1  large_airport
...     ...     ...     ...   ...   ...     ...
363125  9E    4052         4   370     90     0  medium_airport
363126  9E    3704         4   705     92     1  medium_airport
363127  9E    4060         4   990     90     0  medium_airport
363128  9E    4052         5   370     90     1  medium_airport

```

363129	9E	3704	5	705	92	0	medium_airport
--------	----	------	---	-----	----	---	----------------

	latitude_deg	longitude_deg	elevation_ft	scheduled_service	\
0	37.618999	-122.375000	13.0	yes	
1	37.618999	-122.375000	13.0	yes	
2	37.618999	-122.375000	13.0	yes	
3	37.618999	-122.375000	13.0	yes	
4	37.618999	-122.375000	13.0	yes	
...	
363125	32.847099	-96.851799	487.0	yes	
363126	32.847099	-96.851799	487.0	yes	
363127	32.847099	-96.851799	487.0	yes	
363128	32.847099	-96.851799	487.0	yes	
363129	32.847099	-96.851799	487.0	yes	

	airport_ref	length_ft	width_ft	lighted	closed	Founded	\
0	3878	7500.0	200.0	1	0	1926.0	
1	3878	7500.0	200.0	1	0	1926.0	
2	3878	7500.0	200.0	1	0	1926.0	
3	3878	7500.0	200.0	1	0	1926.0	
4	3878	7500.0	200.0	1	0	1926.0	
...	
363125	3479	7752.0	150.0	1	0	1985.0	
363126	3479	7752.0	150.0	1	0	1985.0	
363127	3479	7752.0	150.0	1	0	1985.0	
363128	3479	7752.0	150.0	1	0	1985.0	
363129	3479	7752.0	150.0	1	0	1985.0	

	traffic_Chg19_20	traffic_Chg18_19	hubs
0	-14094543	1359791.0	large_hub
1	-14094543	1359791.0	large_hub
2	-14094543	1359791.0	large_hub
3	-14094543	1359791.0	large_hub
4	-14094543	1359791.0	large_hub
...
363125	999564	132553.0	Medium_hub
363126	999564	132553.0	Medium_hub
363127	999564	132553.0	Medium_hub
363128	999564	132553.0	Medium_hub
363129	999564	132553.0	Medium_hub

[351555 rows x 20 columns]

Only one traffic_Chg18_19 column contain the null value which is 3.18% of total data.

So we will drop that rows of null value because we have plenty of data.

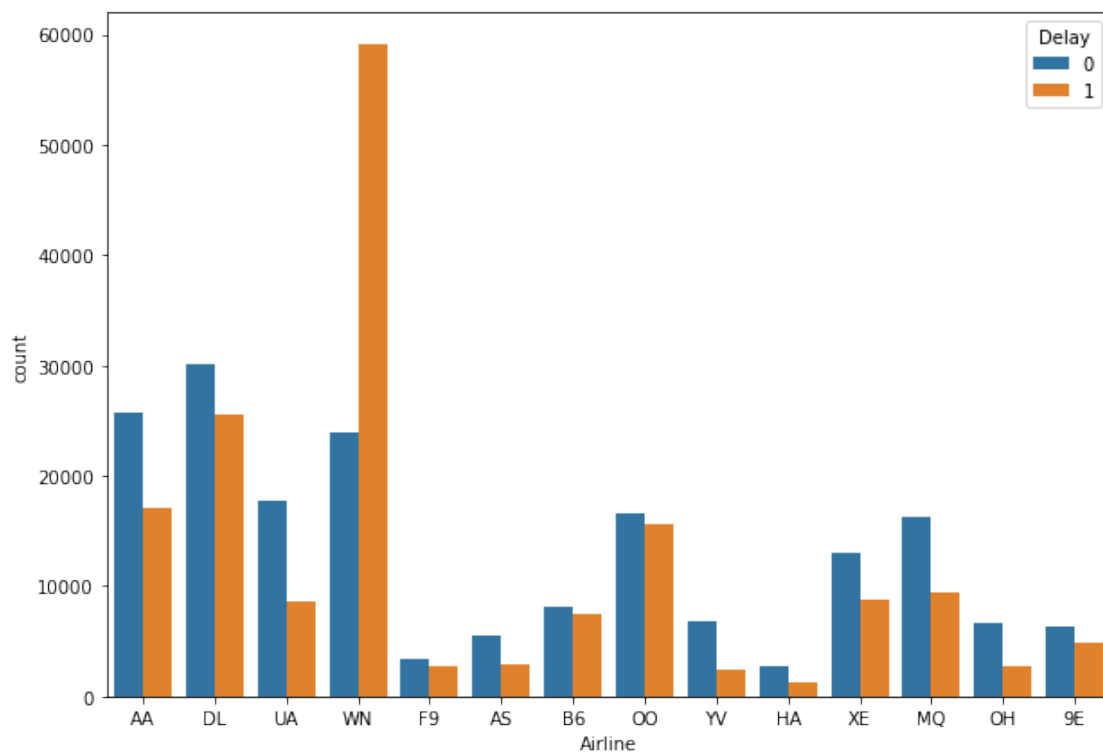
```
[58]: final_df.dropna(axis=0,inplace=True)
```

1.3 3. Perform data visualization and share your insights on the following points:

1.3.1 a. According to the data provided, approximately 70% of Southwest Airlines flights are delayed. Visualize it to compare it with the data of other airlines.

```
[60]: plt.figure(figsize=(10,7))
sns.countplot(final_df['Airline'], hue= final_df['Delay'])
```

```
[60]: <AxesSubplot:xlabel='Airline', ylabel='count'>
```

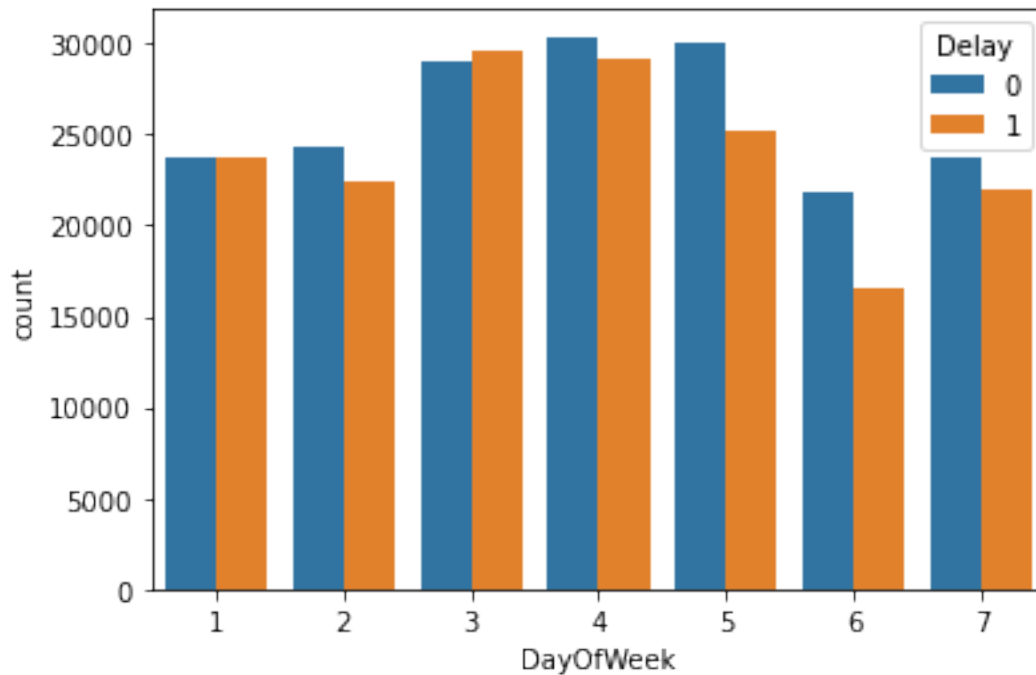


Airline code WN represent the southwest airlines. The graph clear show that 70% of flight of south west airline is delayed

1.3.2 b. Flights were delayed on various weekdays. Which day of the week is the safest for travel?

```
[61]: sns.countplot(final_df['DayOfWeek'], hue= final_df['Delay'])
```

```
[61]: <AxesSubplot:xlabel='DayOfWeek', ylabel='count'>
```



We can from the above graph that on 6th day of the week we have least delayed flight.

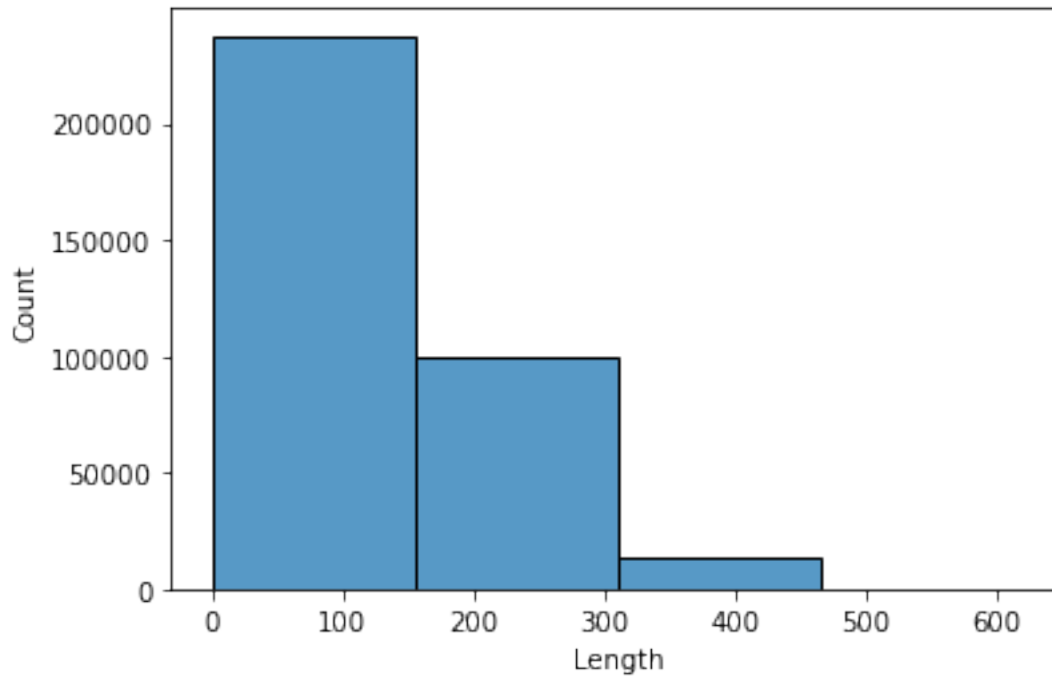
1.3.3 c. Which airlines should be recommended for short-, medium-, and long-distance travel?

```
[69]: final_df['Length'].max()
```

```
[69]: 620
```

```
[78]: sns.histplot(final_df['Length'], bins = 4)
```

```
[78]: <AxesSubplot:xlabel='Length', ylabel='Count'>
```



```
[80]: final_df['Airline'][final_df['Length']<180].value_counts()
```

```
[80]: WN      70485
      DL      38483
      OO      30767
      MQ      24954
      AA      24370
      XE      20807
      UA      15534
      9E      11145
      B6       9766
      YV       9152
      OH       8989
      AS       5386
      F9       5105
      HA       3034
      Name: Airline, dtype: int64
```

The above airlines are recommended for Short distance as flight last anywhere from 30 minutes to 3 hours.

```
[86]: final_df['Airline'][(final_df['Length']>180) & (final_df['Length']<360)].
      ↪value_counts()
```

```
[86]: DL      16180
      AA      15959
      WN      11333
      UA       9380
      B6       4847
      AS       2354
      OO       1473
      F9       1075
      XE        911
      HA        751
      MQ        512
      OH        423
      YV        185
      9E         47
      Name: Airline, dtype: int64
```

The above airlines are recommended for Medium distance as flight last anywhere from 3 hours to 6 hours.

```
[87]: final_df['Airline'][final_df['Length']>360].value_counts()
```

```
[87]: UA      1304
      AA      1081
      DL       842
      B6       822
      AS       496
      HA       252
      WN        52
      Name: Airline, dtype: int64
```

The above airlines are recommended for Long distance as flights are extend beyond 6 hours.

d. Do you notice any patterns in the departure times of long-duration flights?

```
[114]: final_df[final_df['Length']>360].describe().T
```

```
[114]:
```

	count	mean	std	min	\
Flight	4849.0	5.563933e+02	7.537248e+02	1.000000e+00	
DayOfWeek	4849.0	4.004125e+00	1.925699e+00	1.000000e+00	
Time	4849.0	8.292021e+02	2.868094e+02	1.000000e+02	
Length	4849.0	3.959827e+02	4.084421e+01	3.610000e+02	
Delay	4849.0	4.188492e-01	4.934215e-01	0.000000e+00	
latitude_deg	4849.0	3.873399e+01	6.170430e+00	1.843940e+01	
longitude_deg	4849.0	-8.772557e+01	2.456466e+01	-1.579242e+02	
elevation_ft	4849.0	2.393364e+02	7.504192e+02	8.000000e+00	
airport_ref	4849.0	3.726874e+03	4.652847e+02	3.384000e+03	
length_ft	4849.0	1.004975e+04	2.111061e+03	4.892000e+03	
width_ft	4849.0	1.718602e+02	2.509397e+01	1.000000e+02	

lighted	4849.0	9.509177e-01	2.160621e-01	0.000000e+00
closed	4849.0	0.000000e+00	0.000000e+00	0.000000e+00
Founded	4849.0	1.939067e+03	2.703800e+01	1.924000e+03
traffic_Chg19_20	4849.0	-1.167185e+07	5.257852e+06	-3.294593e+07
traffic_Chg18_19	4849.0	1.297978e+06	7.706599e+05	-8.712900e+04

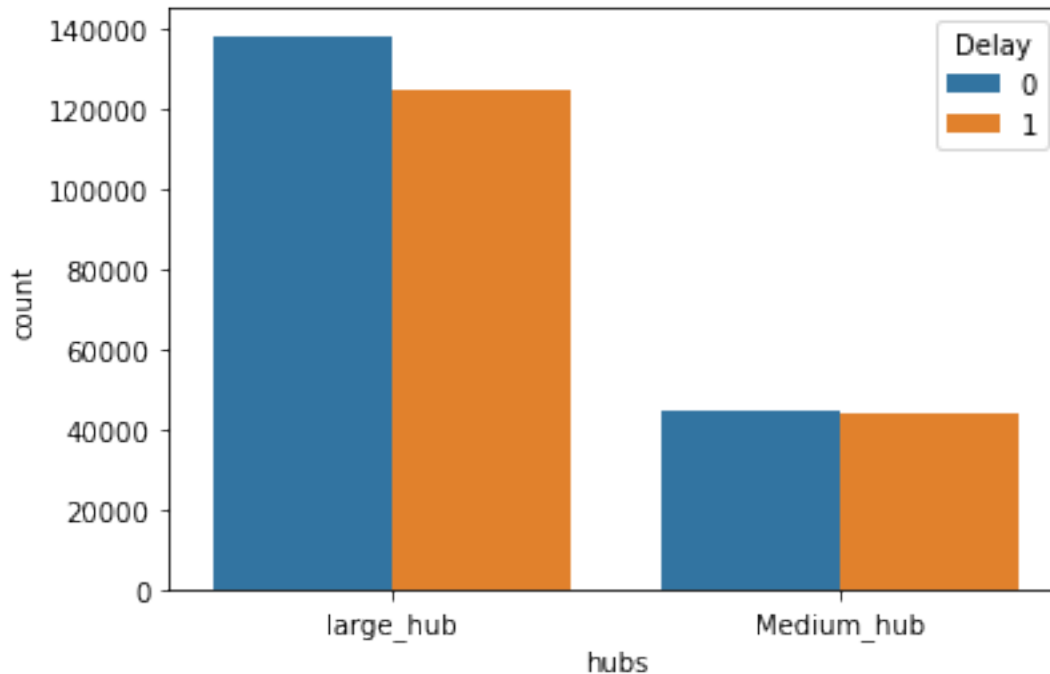
		25%	50%	75%	max
Flight	5.900000e+01	2.010000e+02	8.460000e+02	3.760000e+03	
DayOfWeek	2.000000e+00	4.000000e+00	6.000000e+00	7.000000e+00	
Time	5.500000e+02	8.850000e+02	1.080000e+03	1.435000e+03	
Length	3.750000e+02	3.850000e+02	4.000000e+02	6.200000e+02	
Delay	0.000000e+00	0.000000e+00	1.000000e+00	1.000000e+00	
latitude_deg	3.894450e+01	4.063945e+01	4.078575e+01	4.744916e+01	
longitude_deg	-9.703800e+01	-7.377932e+01	-7.377932e+01	-6.600180e+01	
elevation_ft	1.300000e+01	1.800000e+01	1.250000e+02	5.431000e+03	
airport_ref	3.602000e+03	3.622000e+03	3.632000e+03	6.384000e+03	
length_ft	7.861000e+03	1.100000e+04	1.207900e+04	1.207900e+04	
width_ft	1.500000e+02	1.500000e+02	2.000000e+02	2.000000e+02	
lighted	1.000000e+00	1.000000e+00	1.000000e+00	1.000000e+00	
closed	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	
Founded	1.926000e+03	1.926000e+03	1.932000e+03	1.998000e+03	
traffic_Chg19_20	-1.212320e+07	-1.124682e+07	-1.124682e+07	1.485473e+06	
traffic_Chg18_19	7.479110e+05	9.972960e+05	1.917739e+06	2.956774e+06	

```
[ ]:
```

1.3.4 4. How many flights were delayed at large hubs compared to medium hubs? Use appropriate visualization to represent your findings

```
[105]: sns.countplot(final_df['hubs'], hue = final_df['Delay'])
```

```
[105]: <AxesSubplot:xlabel='hubs', ylabel='count'>
```



From the large hubs its clear approx 120000 flight is delayed but from the medium hubs approx 40000 is delayed.

1.3.5 5. Use hypothesis testing strategies to discover:

a. If the airport's altitude has anything to do with flight delays for incoming and departing flights

```
[115]: from scipy.stats import chi2_contingency
table = [final_df['latitude_deg'], final_df['Delay']]
stat, p, dof, expected = chi2_contingency(table)
print('stat=%.3f, p=%.3f' % (stat, p))
if p > 0.05:
    print('Probably independent')
else:
    print('Probably dependent')
```

stat=186602.569, p=1.000

Probably independent

So its clear from the above hypothesis testing that altitude is nothing to do with the flight delay

b. If the number of runways at an airport affects flight delays

```
[117]: from scipy.stats import chi2_contingency
table = [final_df['airport_ref'], final_df['Delay']]
stat, p, dof, expected = chi2_contingency(table)
print('stat=%.3f, p=%.3f' % (stat, p))
if p > 0.05:
    print('Probably independent')
else:
    print('Probably dependent')
```

stat=192200.911, p=1.000

Probably independent

So its clear from the above hypothesis testing that no of runway is nothing to do with the flight delay

c. If the duration of a flight (length) affects flight delays

```
[118]: from scipy.stats import spearmanr
data1 = final_df['Length']
data2 = final_df['Delay']
stat, p = spearmanr(data1, data2)
print('stat=%.3f, p=%.3f' % (stat, p))
if p > 0.05:
    print('Probably independent')
else:
    print('Probably dependent')
```

stat=-0.002, p=0.179

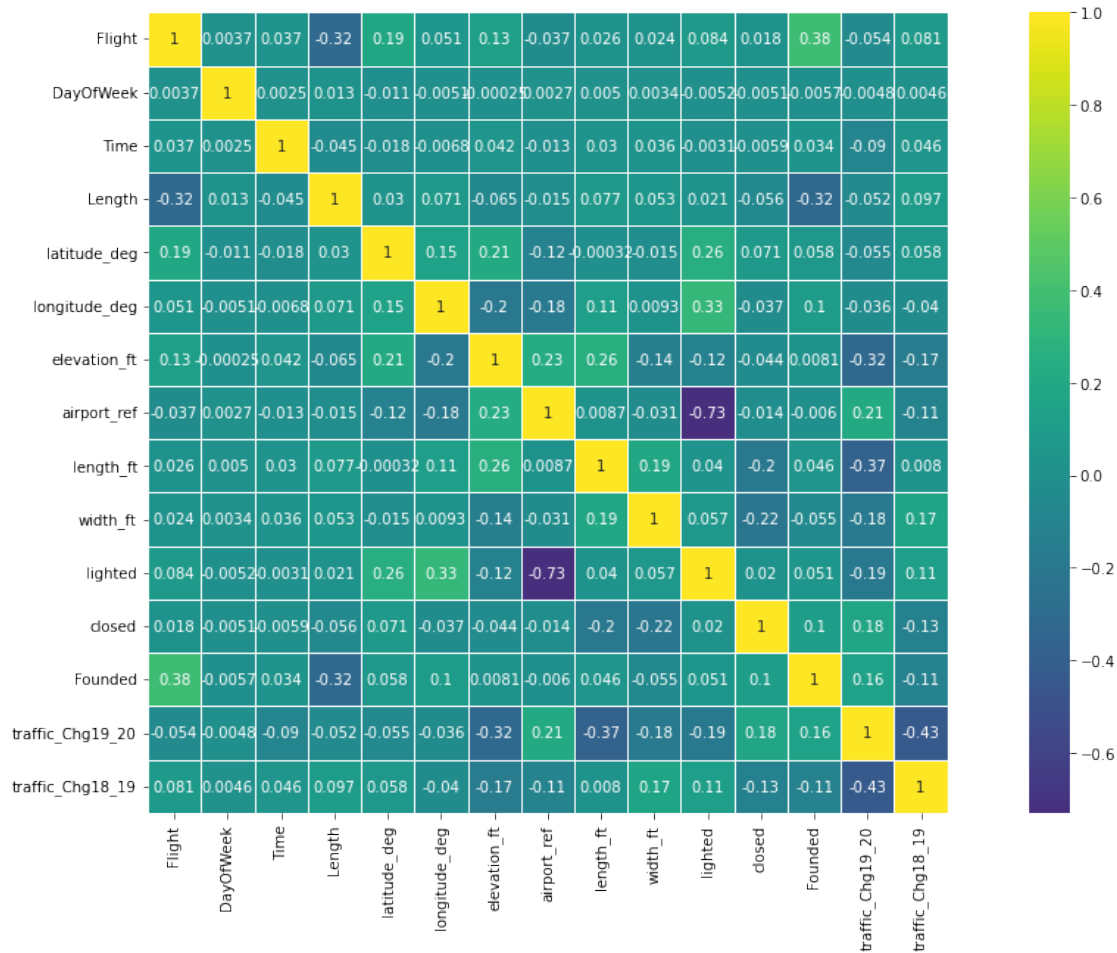
Probably independent

Both the variable are independent so that length of the flight is not affecting directly the delay.

1.3.6 6. Find the correlation matrix between the flight delay predictors, create a heatmap to visualize this, and share your findings

```
[121]: predictor = final_df.drop(['Delay'], axis=1)
corr = predictor.corr()
```

```
[126]: plt.figure(figsize=(20,10))
sns.heatmap(corr, center=0, linewidths=.5, square = True , annot = True ,
            cmap='viridis')
plt.show()
```



2 Project Task: Week 1

2.1 Machine learning

2.1.1 1. Use OneHotEncoder and OrdinalEncoder to deal with categorical variables

[127]: *# Before applying the one hot encoding or the label encoding first we check*
↪ all feature data type.

```
final_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 351555 entries, 0 to 363129
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Airline                351555 non-null object
```

```

1  Flight          351555 non-null  int64
2  DayOfWeek       351555 non-null  int64
3  Time            351555 non-null  int64
4  Length          351555 non-null  int64
5  Delay           351555 non-null  int64
6  type            351555 non-null  object
7  latitude_deg    351555 non-null  float64
8  longitude_deg   351555 non-null  float64
9  elevation_ft    351555 non-null  float64
10 scheduled_service 351555 non-null  object
11 airport_ref     351555 non-null  int64
12 length_ft       351555 non-null  float64
13 width_ft        351555 non-null  float64
14 lighted         351555 non-null  int64
15 closed          351555 non-null  int64
16 Founded         351555 non-null  float64
17 traffic_Chg19_20 351555 non-null  int64
18 traffic_Chg18_19 351555 non-null  float64
19 hubs            351555 non-null  object
dtypes: float64(7), int64(9), object(4)
memory usage: 56.3+ MB

```

```
[128]: final_df['Airline'].value_counts()
```

```

[128]: WN      82903
      DL      55724
      AA      42841
      OO      32315
      UA      26303
      MQ      25698
      XE      21733
      B6      15497
      9E      11192
      OH       9440
      YV       9337
      AS       8355
      F9       6180
      HA       4037
      Name: Airline, dtype: int64

```

```
[129]: final_df['type'].value_counts()
```

```

[129]: large_airport    334982
      medium_airport    16573
      Name: type, dtype: int64

```

```
[130]: final_df['scheduled_service'].value_counts()
```

```
[130]: yes      351555
      Name: scheduled_service, dtype: int64
```

```
[131]: final_df['hubs'].value_counts()
```

```
[131]: large_hub      262540
      Medium_hub      89015
      Name: hubs, dtype: int64
```

The scheduled_service column thought has same value so it will not help in prediction so lets remove it and other three object column we will change through label encoder.

```
[132]: final_df = final_df.drop(['scheduled_service'], axis=1)
```

```
[133]: # Now using the ordinal encoder.
      from sklearn.preprocessing import LabelEncoder
      le = LabelEncoder()
```

```
[134]: final_df['Airline'] = le.fit_transform(final_df['Airline'])
      final_df['type'] = le.fit_transform(final_df['type'])
      final_df['hubs'] = le.fit_transform(final_df['hubs'])
```

```
[135]: final_df.head()
```

```
[135]:
```

	Airline	Flight	DayOfWeek	Time	Length	Delay	type	latitude_deg	\
0	1	2466	3	20	195	1	0	37.618999	
1	1	526	3	360	215	0	0	37.618999	
2	1	552	3	360	315	1	0	37.618999	
3	1	810	3	385	255	0	0	37.618999	
4	1	24	3	425	325	1	0	37.618999	

	longitude_deg	elevation_ft	airport_ref	length_ft	width_ft	lighted	\
0	-122.375	13.0	3878	7500.0	200.0	1	
1	-122.375	13.0	3878	7500.0	200.0	1	
2	-122.375	13.0	3878	7500.0	200.0	1	
3	-122.375	13.0	3878	7500.0	200.0	1	
4	-122.375	13.0	3878	7500.0	200.0	1	

	closed	Founded	traffic_Chg19_20	traffic_Chg18_19	hubs
0	0	1926.0	-14094543	1359791.0	1
1	0	1926.0	-14094543	1359791.0	1
2	0	1926.0	-14094543	1359791.0	1
3	0	1926.0	-14094543	1359791.0	1
4	0	1926.0	-14094543	1359791.0	1

2.1.2 2. Perform the following model building steps:

a. Apply logistic regression (use stochastic gradient descent optimizer) and decision tree models

b. Use the stratified five-fold method to build and validate the models Note: Make sure you use standardization effectively, ensuring no data leakage and leverage pipelines to have a cleaner code ##### c. Use RandomizedSearchCV for hyperparameter tuning, and use k-fold for crossvalidation ##### d. Keep a few data points (10%) for prediction purposes to evaluate how you would make the final prediction, and do not use this data for testing or validation Note: The final prediction will be based on the voting (majority class by 5 models created using the stratified 5-fold method) ##### g. Compare the results of logistic regression and decision tree classifier

```
[136]: # Lets first separate the features and the target.
x = final_df.drop(['Delay'], axis= 1)
y = final_df["Delay"]
```

```
[137]: from sklearn import preprocessing
scaler = preprocessing.MinMaxScaler()
x = scaler.fit_transform(x)
```

```
[138]: # First Split the data into the training and testing set before performing the
      ↪ further operation.
from sklearn.model_selection import train_test_split
```

```
[139]: x_train, x_test, y_train, y_test = train_test_split(x, y, train_size=0.9,
      ↪ random_state=10)
```

LogisticRegression

```
[140]: # lets apply the logistic regression with the randomsearchcv hypermeter tuning.
from sklearn.linear_model import LogisticRegression
lr = LogisticRegression()
```

```
[141]: from sklearn.model_selection import RandomizedSearchCV
```

```
[142]: params = {"penalty": ["l1","l2"],
               'solver': ['newton-cg', 'liblinear']}

# Cross Validation
folds = 5

rscv = RandomizedSearchCV(estimator = lr,
                          param_distributions = params,
                          scoring = "accuracy",
                          verbose = 1,
                          cv= folds)
```

```
rscv.fit(x_train, y_train)
```

Fitting 5 folds for each of 4 candidates, totalling 20 fits

```
[142]: RandomizedSearchCV(cv=5, estimator=LogisticRegression(),  
                        param_distributions={'penalty': ['l1', 'l2'],  
                                           'solver': ['newton-cg', 'liblinear']},  
                        scoring='accuracy', verbose=1)
```

```
[143]: print(rscv.best_params_)  
       print(rscv.best_score_)
```

```
{'solver': 'liblinear', 'penalty': 'l1'}  
0.5921478987792488
```

```
[144]: lr = LogisticRegression(penalty= 'l2', solver= 'newton-cg')  
       lr.fit(x_train,y_train).score(x_train,y_train)
```

```
[144]: 0.5923280414919136
```

```
[145]: lr.score(x_test, y_test)
```

```
[145]: 0.593013994766185
```

DecisionTreeClassifier

```
[146]: from sklearn.tree import DecisionTreeClassifier  
  
dt = DecisionTreeClassifier()  
  
params = {'criterion': ["gini", "entropy"],  
          'min_samples_leaf' : [2,3,4,5,6,7,8,9],  
          "max_depth": [2,3,4,5,6,7,8,9]}  
  
rscv = RandomizedSearchCV(estimator = dt,  
                          param_distributions= params,  
                          scoring = "accuracy",  
                          cv= 5,  
                          verbose=1)  
  
rscv.fit(x_train, y_train)
```

Fitting 5 folds for each of 10 candidates, totalling 50 fits

```
[146]: RandomizedSearchCV(cv=5, estimator=DecisionTreeClassifier(),  
                        param_distributions={'criterion': ['gini', 'entropy'],  
                                           'max_depth': [2, 3, 4, 5, 6, 7, 8, 9],  
                                           'min_samples_leaf': [2, 3, 4, 5, 6, 7,  
                                                                8, 9]}),
```



```
scoring='accuracy', verbose=1)
```

```
[147]: print(rscv.best_params_)  
       print(rscv.best_score_)
```

```
{'min_samples_leaf': 6, 'max_depth': 9, 'criterion': 'entropy'}  
0.6469110137916109
```

```
[148]: dtc = DecisionTreeClassifier(max_depth= 9, criterion=□  
       ↪ 'entropy', min_samples_leaf= 6)  
       dtc.fit(x_train, y_train).score(x_train, y_train)
```

```
[148]: 0.6539464410443775
```

```
[149]: dtc.score(x_test, y_test)
```

```
[149]: 0.649049948799636
```

After seeing the result its clear decision tree has good accuracy.

2.1.3 3. Use the stratified five-fold method to build and validate the models using the XGB classifier, compare all methods, and share your findings

```
[150]: from xgboost import XGBClassifier  
  
       # Create the parameter grid: gbm_param_grid  
       gbm_param_grid = {  
           'n_estimators': range(8, 20),  
           'max_depth': range(6, 10),  
           'learning_rate': [.4, .45, .5, .55, .6],  
           'colsample_bytree': [.6, .7, .8, .9, 1]  
       }  
  
       # Instantiate the regressor: gbm  
       gbm = XGBClassifier()  
  
       # Perform random search: grid_mse  
       xgb_random = RandomizedSearchCV(param_distributions=gbm_param_grid,  
                                       estimator = gbm, scoring = "accuracy",  
                                       verbose = 1, n_iter = 50, cv = 3)  
  
       # Fit randomized_mse to the data  
       xgb_random.fit(x_train, y_train)  
  
       # Print the best parameters and lowest RMSE  
       print("Best parameters found: ", xgb_random.best_params_)
```

```
print("Best accuracy found: ", xgb_random.best_score_)
```

Fitting 3 folds for each of 50 candidates, totalling 150 fits

Best parameters found: {'n_estimators': 16, 'max_depth': 8, 'learning_rate': 0.5, 'colsample_bytree': 0.9}

Best accuracy found: 0.6608301551327947

```
[151]: xgb = XGBClassifier(n_estimators=14, max_depth=9, learning_rate=0.45,
    ↪ colsample_bytree=0.9)
xgb.fit(x_train,y_train).score(x_train,y_train)
```

```
[151]: 0.6860830786443699
```

```
[152]: # Now lets compare the all method.
print(lr.score(x_test, y_test))
print(dtc.score(x_test, y_test))
print(xgb.score(x_test, y_test))
```

0.593013994766185

0.649049948799636

0.6630447149846399

After comparing the accuracy of the different model the best result we getting from the XGBclassifier.