

## Importing necessary libraries

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

## PROJECT TASK: WEEK 1

### 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.

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

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

Out[3]:

	id	ident	type	name	latitude_deg	longitude_deg	elevation_ft	continent	iso_country	iso_region	mun...
0	6523	00A	heliport	Total Rf Heliport	40.070801	-74.933601	11.0	NaN	US	US-PA	Be
1	323361	00AA	small_airport	Aero B Ranch Airport	38.704022	-101.473911	3435.0	NaN	US	US-KS	
2	6524	00AK	small_airport	Lowell Field	59.947733	-151.692524	450.0	NaN	US	US-AK	Anch
3	6525	00AL	small_airport	Epps Airpark	34.864799	-86.770302	820.0	NaN	US	US-AL	
4	6526	00AR	closed	Newport Hospital & Clinic Heliport	35.608700	-91.254898	237.0	NaN	US	US-AR	N

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

Out[4]:

(73805, 18)

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

Out[5]:

	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

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

Out[6]:

(518556, 9)

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

Out[7]:

	id	airport_ref	airport_ident	length_ft	width_ft	surface	lighted	closed	le_ident	le_latitude_deg	le_longitude_deg
0	269408	6523	00A	80.0	80.0	ASPH-G	1	0	H1	NaN	NaN
1	255155	6524	00AK	2500.0	70.0	GRVL	0	0	N	NaN	NaN
2	254165	6525	00AL	2300.0	200.0	TURF	0	0	1	NaN	NaN
3	270932	6526	00AR	40.0	40.0	GRASS	0	0	H1	NaN	NaN
4	322128	322127	00AS	1450.0	60.0	Turf	0	0	1	NaN	NaN

In [8]:

runways.shape

Out[8]:

(43977, 20)

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

In [9]:

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

In [10]:

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\_displaced\_threshold\_ft'], axis = 1, inplace=True)

In [11]:

runways.head()

Out[11]:

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

In [12]:

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
```

```
In [13]: airport.drop(['continent', 'iso_country', 'iso_region','municipality', 'gps_code','local_code', 'home_li
                'wikipedia_link', 'keywords'], axis=1, inplace=True)
airport
```

Out[13]:

	id	ident	type	name	latitude_deg	longitude_deg	elevation_ft	scheduled_service	iata_code
0	6523	00A	heliport	Total Rf Heliport	40.070801	-74.933601	11.0	no	NaN
1	323361	00AA	small_airport	Aero B Ranch Airport	38.704022	-101.473911	3435.0	no	NaN
2	6524	00AK	small_airport	Lowell Field	59.947733	-151.692524	450.0	no	NaN
3	6525	00AL	small_airport	Epps Airpark	34.864799	-86.770302	820.0	no	NaN
4	6526	00AR	closed	Newport Hospital & Clinic Heliport	35.608700	-91.254898	237.0	no	NaN
...	...	...	...	...	...	...	...	...	...
73800	46378	ZZ-0001	heliport	Sealand Helipad	51.894444	1.482500	40.0	no	NaN
73801	307326	ZZ-0002	small_airport	Glorioso Islands Airstrip	-11.584278	47.296389	11.0	no	NaN
73802	346788	ZZ-0003	small_airport	Fainting Goat Airport	32.110587	-97.356312	690.0	no	NaN
73803	342102	ZZZW	closed	Scandium City Heliport	69.355287	-138.939310	4.0	no	ZYW
73804	313629	ZZZZ	small_airport	Satsuma Iejima Airport	30.784722	130.270556	338.0	no	NaN

73805 rows × 9 columns

merge the runways and airport data.

```
In [14]: airport_runway = pd.merge(airport, runways, left_on = "ident", right_on = "airport_ident")

In [15]: airport_runway.drop(['id_x', 'id_y'], axis=1, inplace=True)

In [16]: airport_runway
```

Out[16]:

	ident	type	name	latitude_deg	longitude_deg	elevation_ft	scheduled_service	iata_code	airpor
0	00A	heliport	Total Rf Heliport	40.070801	-74.933601	11.0	no	NaN	
1	00AK	small_airport	Lowell Field	59.947733	-151.692524	450.0	no	NaN	
2	00AL	small_airport	Epps Airpark	34.864799	-86.770302	820.0	no	NaN	
3	00AR	closed	Newport Hospital & Clinic Heliport	35.608700	-91.254898	237.0	no	NaN	
4	00AS	small_airport	Fulton Airport	34.942803	-97.818019	1100.0	no	NaN	32
...	...	...	...	...	...	...	...	...	
43972	ZYTX	large_airport	Shenyang Taoxian International Airport	41.639801	123.483002	198.0	yes	SHE	2
43973	ZYYJ	medium_airport	Yanji Chaoyangchuan Airport	42.882801	129.451004	624.0	yes	YNJ	2
43974	ZYYK	medium_airport	Yingkou Lanqi Airport	40.542524	122.358600	NaN	yes	YKH	31
43975	ZZ-0003	small_airport	Fainting Goat Airport	32.110587	-97.356312	690.0	no	NaN	34
43976	ZZZZ	small_airport	Satsuma Iqjima Airport	30.784722	130.270556	338.0	no	NaN	31

43977 rows × 15 columns



merge the final column airline.

In [17]:

```
final_df = pd.merge(airlines,airport_runway,how = "inner", left_on = "AirportFrom", right_on = "iata_code")
```

In [18]:

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

final\_df

Out[18]:

	id	Airline	Flight	AirportFrom	AirportTo	DayOfWeek	Time	Length	Delay	ident	...	elevation_ft	sched
0	1	CO	269	SFO	IAH	3	15	205	1	KSFO	...	13.0	
4	4	AA	2466	SFO	DFW	3	20	195	1	KSFO	...	13.0	
8	9	DL	2606	SFO	MSP	3	35	216	1	KSFO	...	13.0	
12	129	DL	1580	SFO	DTW	3	345	270	0	KSFO	...	13.0	
16	150	UA	756	SFO	DEN	3	348	158	0	KSFO	...	13.0	
...	...	...	...	...	...	...	...	...	...	...	...	...	
2160266	451344	CO	2	GUM	HNL	1	400	430	1	PGUM	...	298.0	
2160268	469866	CO	2	GUM	HNL	2	400	430	1	PGUM	...	298.0	
2160270	488365	CO	2	GUM	HNL	3	400	430	0	PGUM	...	298.0	
2160272	506855	CO	2	GUM	HNL	4	400	430	1	PGUM	...	298.0	
2160274	525138	CO	2	GUM	HNL	5	400	430	1	PGUM	...	298.0	

518525 rows × 24 columns



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 19860. 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](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.

```
In [19]: url = "https://en.wikipedia.org/wiki/List_of_airlines_of_the_United_States"  
         tables = pd.read_html(url)
```

```
In [20]: print(tables)
```

	0					1			
0	NaN	This article does not cite any sources. Please...					Airline	Image	IATA ICAO
	Callsign \								
0	Alaska Airlines	NaN	AS	ASA		ALASKA			
1	Allegiant Air	NaN	G4	AAY		ALLEGIAN			
2	American Airlines	NaN	AA	AAL		AMERICAN			
3	Avelo Airlines	NaN	XP	VXP		AVELO			
4	Breeze Airways	NaN	MX	MX		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/TacomaAnchoragePortland (OR)San Franci...		1932
1	Las VegasCincinnatiFort Walton BeachIndianapol...		1997
2	Dallas/Fort WorthCharlotteChicago-O'HareLos An...		1926
3	BurbankNew HavenOrlando		1987
4	CharlestonHartfordNew OrleansNorfolkProvoTampa		2018
5	AtlantaBostonDetroitLos AngelesMinneapolis/St....		1924
6	MiamiNew York-JFK		2010
7	DenverAtlantaChicago-O'HareCincinnatiCleveland...		1994
8	HonoluluKahului		1929
9	New York-JFKBostonLos AngelesFort LauderdaleOr...		1998
10	Dallas-LoveAtlantaBaltimoreChicago-MidwayDenve...		1967
11	Atlantic CityDetroitLas VegasFort LauderdaleCh...		1980
12	Minneapolis/St. PaulDallas/Fort WorthLas Vegas		1982
13	Chicago-O'HareDenverGuamHouston-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... ,

	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	AppletonChicago-O'HareColumbiaMilwaukeeWashing...		1965
1	HyannisBillingsBostonNantucketSt. LouisSan Jua...		1988
2	DenverNewarkWashington-Dulles		1989
3	Smyrna (TN)		1982
4	Melbourne/OrlandoNewarkPortland (Maine)		2006
5	Minneapolis/St. PaulAtlanta CincinnatiDetroitN...		1985
6	Dallas/Fort WorthChicago-O'Hare Miami		1984
7	Chicago-O'HareDenver		2004
8	Seattle/TacomaPortland (OR)		1981
9	As American Eagle:Phoenix-Sky HarborDallas/For...		1980
10	CharlottePhiladelphiaWashington-National		1961

11	CharlottePhiladelphiaWashington-National	1979
12	As American Eagle:IndianapolisColumbus (OH)Kan...	1998
13	Fort LauderdaleOrlandoTampa	2011
14	As Delta Connection:AtlantaBoiseColorado 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 IC

A0	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	VTG	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	Operator of Taos Air flights from 2022.
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 IA

TA	ICAO	Callsign \				
0		Air Charter Bahamas	NaN	NaN	NaN	NaN
1		Air Flight Charters	NaN	NaN	FLL	NaN
2		Airshare	NaN	NaN	XSR	AIRSHARE
3		Berry Aviation	NaN	NaN	BYA	BERRY
4		Bighorn Airways	NaN	NaN	BHR	BIGHORN AIR
5		Charter Air Transport	NaN	VC	SRY	STINGRAY
6		Choice Airways	NaN	NaN	CSX	CHOICE AIR
7		ExcelAire	NaN	NaN	XLS	EXCELAIRE
8		Global Crossing Airlines	NaN	G6	GXA	GEMINI
9		Great Lakes Air	NaN	NaN	NaN	NaN
10		Gryphon Airlines	NaN	Y3	VOS	NaN
11		IAero Airways	NaN	WQ	SWQ	SWIFTFLIGHT
12		IBC Airways	NaN	II	CSQ	CHASQUI
13	L-3 Flight International Aviation		NaN	NaN	RTD	RIPTIDE
14	Liberty Jet Management		NaN	NaN	LRT	LIBERTY JET
15	NetJets		NaN	1I	EJA	EXECJET
16	Omni Air International		NaN	X9	OAE	OMNI-EXPRESS
17	Omni Air Transport		NaN	NaN	DRL	DRILLER
18	Pacific Coast Jet		NaN	NaN	PXT	PACK COAST
19	Pentastar Aviation		NaN	NaN	DCX	TANGO
20	Phoenix Air		NaN	NaN	PHA	GRAY BIRD
21	PlaneSense		NaN	NaN	CNS	CHRONOS
22	Presidential Airways		NaN	NaN	PRD	PRESIDENTIAL
23	Sierra Pacific Airlines		NaN	SI	SPA	SIERRA PACIFIC
24	Skymax		NaN	NaN	SMX	SKYMAX
25	Stampede Aviation		NaN	NaN	NaN	NaN
26	Superior Air Charter		NaN	NaN	RSP	REDSTRIPE
27	Superior Aviation		NaN	SO	HKA	SPEND AIR
28	Talkeetna Air Taxi		NaN	NaN	NaN	NaN
29	Tropic Ocean Airways		NaN	NaN	NaN	NaN
30	World Atlantic Airlines		NaN	K8	WAL	WORLD ATLANTIC
31	XOJET Aviation LLC		NaN	NaN	XOJ	XOJET

	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 CityLas VegasMiami	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	Healy/Denali NP	2011.0
26	NaN	2006.0
27	Lansing	1979.0
28	Talkeetna	1947.0
29	Fort Lauderdale	2009.0
30	Miami	2002.0
31	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
15	Founded as Executive Jets Aviation.
16	NaN
17	NaN
18	NaN
19	Founded as Chrysler Air Transportation.
20	NaN
21	NaN
22	NaN
23	Commenced operations in 1971.
24	Commenced operations in 2013.
25	NaN
26	NaN
27	NaN
28	Founded as Talkeetna Flying Service.
29	NaN
30	Founded as Caribbean Sun Airlines and commence...
31	NaN

	Callsign \	21 Air	NaN	2I	CSB	CARGO SOUTH
0						
1		ABX Air	NaN	GB	ABX	ABEX
2		Air Cargo Carriers	NaN	2Q	SNC	NIGHT CARGO
3		AirNet Express	NaN	NaN	USC	STAR CHECK
4	Air Transport International		NaN	8C	ATN	AIR TRANSPORT
5	Alaska Central Express		NaN	KO	AER	ACE AIR
6	Aloha Air Cargo		NaN	KH	AAH	ALOHA
7	Alpine Air Express		NaN	5A	AIP	ALPINE AIR
8	Amazon Air		NaN	AFW	KAFW	AMAZON AIR
9	Ameriflight		NaN	A8	AMF	AMFLIGHT
10	Amerijet International		NaN	M6	AJT	AMERIJET
11	Ameristar Jet Charter		NaN	7Z	AJI	AMERISTAR
12	Asia Pacific Airlines		NaN	P9	MGE	MAGELLAN
13	Atlas Air		NaN	5Y	GTI	GIANT
14	Bemidji Airlines		NaN	CH	BMJ	BEMIDJI
15	Castle Aviation		NaN	NaN	CSJ	CASTLE
16	Corporate Air		NaN	NaN	CPT	AIRSPUR
17	CSA Air		NaN	NaN	IRO	IRON AIR
18	Empire Airlines		NaN	EM	CFS	EMPIRE
19	Everts Air Cargo		NaN	5V	VTS	EVERTS
20	FedEx Express		NaN	FX	FDX	FEDEX
21	Freight Runners Express		NaN	NaN	FRG	FREIGHT RUNNERS

22	IFL Group	NaN	IF	IFL	EIFFEL
23	Kalitta Air	NaN	K4	CKS	CONNIE
24	Kalitta Charters	NaN	CB	KFS	KALITTA
25	Lynden Air Cargo	NaN	L2	LYC	LYNDEN
26	Martinaire	NaN	NaN	MRA	MARTEX
27	Merlin Airways	NaN	NaN	MEI	AVALON
28	Mountain Air Cargo	NaN	C2	MTN	MOUNTAIN
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	Trans Executive Airlines	NaN	KH	MUI	RHOADES EXPRESS
38	UPS Airlines	NaN	5X	UPS	UPS
39	USA Jet Airlines	NaN	UJ	JUS	JET USA
40	West Air	NaN	NaN	PCM	PAC VALLEY
41	Western Global Airlines	NaN	KD	WGN	WESTERN GLOBAL
42	Wiggins Airways	NaN	WG	WIG	WIGGINS AIRWAYS

	Primary Hubs, Secondary Hubs	Founded \
0	Miami	2014.0
1	Wilmington (OH)Cincinnati	1980.0
2	MilwaukeeCincinnati	1986.0
3	Columbus-Rickenbacker	1974.0
4	Wilmington (OH)Cincinnati	1978.0
5	Anchorage	1996.0
6	Honolulu	1946.0
7	ProvoBillingsSioux Falls	1971.0
8	Fort Worth/AllianceCincinnatiLeipzig/HalleSan ...	2015.0
9	Dallas/Fort Worth	1968.0
10	MiamiPort of Spain	1974.0
11	Dallas-AddisonEl PasoWillow Run	2000.0
12	GuamHonolulu	1998.0
13	New York-JFKAnchorageCincinnatiHoustonHuntsvil...	1992.0
14	BemidjiMinneapolis/St. Paul	1946.0
15	Akron/Canton	1986.0
16	Billings	1981.0
17	Iron Mountain	1998.0
18	Coeur d' AleneSpokane	1977.0
19	FairbanksAnchorage	1995.0
20	MemphisAnchorageCologne/BonnDubaiFort WorthGre...	1971.0
21	Milwaukee	1985.0
22	WaterfordMiami	1983.0
23	YpsilantiAnchorageBahrainCincinnatiHong KongNe...	1967.0
24	Ypsilanti	NaN
25	Anchorage	1995.0
26	Addison	1978.0
27	BillingsMiamiSan Juan	1983.0
28	Kinston	1974.0
29	Orlando/Sanford	1985.0
30	AnchorageMiami	1956.0
31	AnchorageCincinnatiHong KongHonoluluLos Angele...	1993.0
32	Waterford	1961.0
33	AnchorageAniakBethelEmmonakKotzebueNomeSt. Mar...	1953.0
34	Miami	1969.0
35	NaN	1981.0
36	Miami	2018.0
37	Honolulu	1982.0
38	LouisvilleChicago/RockfordCologne/BonnColumbia...	1988.0
39	YpsilantiLaredo	1994.0
40	Las VegasOaklandOntarioSacramentoSan Diego	1988.0
41	Miami Liege, Belgium; AnchorageFort Myers, FL	2013.0
42	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.

	Notes
0	NaN
1	NaN
2	Commenced operations in 1995. , vteLists of airlines \
0	By airline codes
1	By continent
2	By country
3	vteExpand for full list
4	A Abkhazia Afghanistan Akrotiri and Dhekelia Å...
5	A
6	B
7	C
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9	E
10	F
11	G
12	H
13	I
14	J
15	K
16	L
17	M



0	A	Abkhazia Afghanistan Akrotiri and Dhekelia Åla...		
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		Uganda Ukraine United Arab Emirates United K...		
22		Ukraine United Arab Emirates United Kin...		
23		Vanuatu Vatican City Venezuela Vietnam British...		
24		Wallis and Futuna		
25		Yemen		
26		Zambia Zimbabwe ,	0	
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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	
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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 S
	tates \			
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	
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0  Affiliated Air Wisconsin CommutAir Endeavor Air Envoy Air...
1  Independent Advanced Air Air Flamenco Air Sunshine Bering ...,
of the Americas \
0  List of active airlines of the AmericasNorth A...
1      List of active airlines of the Americas
2          North America
3              Sovereign states
4              Dependencies andother territories
5              South America
6              Sovereign states
7              Dependencies andother territories
8  List of defunct airlines of the AmericasNorth ...
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15              Dependencies andother territories

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3          Antigua and Barbuda The Bahamas Barbados Beliz...
4          Anguilla San Andrés, Providencia and Santa Cat...
5          Sovereign states Argentina Bolivia Brazil Chil...
6          Argentina Bolivia Brazil Chile Colombia Ecuado...
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13         Dependencies andother territories

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5          Bouvet Island Falkland Islands French Guiana S...
6  List of defunct airlines of the AmericasNorth ...
7      List of defunct airlines of the Americas
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9          Antigua and Barbuda The Bahamas Barbados Beliz...
10         Anguilla San Andrés, Providencia and Santa Cat...
11         Sovereign states Argentina Bolivia Brazil Chil...
12         Argentina Bolivia Brazil Chile Colombia Ecuado...
13         Bouvet Island Falkland Islands French Guiana S... ,
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1          Anguilla San Andrés, Providencia and Santa Cat... ,
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1 Bouvet Island Falkland Islands French Guiana S... , List of defunct airlines of the Americas \
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1 Sovereign states
2 Dependencies andother territories
3 South America
4 Sovereign states
5 Dependencies andother territories

List of defunct airlines of the Americas.1
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2 Anguilla San Andrés, Providencia and Santa Cat...
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5 Bouvet Island Falkland Islands French Guiana S... , 0 \
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1 Bouvet Island Falkland Islands French Guiana S... , 0
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0 Authority control: National Israel United States]
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```
In [22]: tables[1]
```

Out[22]:

	Airline	Image	IATA	ICAO	Callsign	Primary hubs, Secondary hubs	Founded	Notes
0	Alaska Airlines	NaN	AS	ASA	ALASKA	Seattle/TacomaAnchoragePortland (OR)San Franci...	1932	Founded as McGee Airways and commenced operati...
1	Allegiant Air	NaN	G4	AAY	ALLEGiant	Las VegasCincinnatiFort Walton BeachIndianapol...	1997	Founded as WestJet Express and commenced opera...
2	American Airlines	NaN	AA	AAL	AMERICAN	Dallas/Fort WorthCharlotteChicago-O'HareLos An...	1926	Founded as American Airways and commenced oper...
3	Avelo Airlines	NaN	XP	VXP	AVELO	BurbankNew HavenOrlando	1987	First did business as Casino Express Airlines ...
4	Breeze Airways	NaN	MX	MXV	MOXY	CharlestonHartfordNew OrleansNorfolkProvoTampa	2018	NaN
5	Delta Air Lines	NaN	DL	DAL	DELTA	AtlantaBostonDetroitLos AngelesMinneapolis/St....	1924	Founded as Huff Daland Dusters and commenced o...
6	Eastern Airlines	NaN	2D	EAL	EASTERN	MiamiNew York-JFK	2010	NaN
7	Frontier Airlines	NaN	F9	FFT	FRONTIER FLIGHT	DenverAtlantaChicago-O'HareCincinnatiCleveland...	1994	NaN
8	Hawaiian Airlines	NaN	HA	HAL	HAWAIIAN	HonoluluKahului	1929	Founded as Inter-Island Airways in early 1929 ...
9	JetBlue	NaN	B6	JBU	JETBLUE	New York-JFKBostonLos AngelesFort LauderdaleOr...	1998	Founded as New Air and commenced operations in...
10	Southwest Airlines	NaN	WN	SWA	SOUTHWEST	Dallas-LoveAtlantaBaltimoreChicago-MidwayDenve...	1967	Founded as Air Southwest and commenced operati...
11	Spirit Airlines	NaN	NK	NKS	SPIRIT WINGS	Atlantic CityDetroitLas VegasFort LauderdaleCh...	1980	Founded as Charter One.
12	Sun Country Airlines	NaN	SY	SCX	SUN COUNTRY	Minneapolis/St. PaulDallas/Fort WorthLas Vegas	1982	Commenced operations in 1983.Operates some Ama...
13	United Airlines	NaN	UA	UAL	UNITED	Chicago-O'HareDenverGuamHouston-Intercontinent...	1926	Founded as Varney Air Lines and commenced oper...

In [24]:

```
tables[7]
```

Out[24]:

	Airline	Image	IATA	ICAO	Callsign	Primary Hubs, Secondary Hubs	Founded	Notes
0			Comco	NaN	NaN	NaN	2002	NaN
1			Janet	NaN	WWW	JANET	Las Vegas	NaN
2	Justice Prisoner and Alien Transportation System		NaN	NaN	JUD	JUSTICE	Oklahoma City	1980 Commenced operations in 1995.

In [25]:

```
# Lets first merge all wikipedia table.  
wiki_table = [tables[1],tables[2],tables[3],tables[4],tables[5],tables[6],tables[7]]
```

In [26]:

```
wiki_tables = pd.concat(wiki_table, ignore_index=True)
```

In [27]:

```
wiki_tables
```



Out[27]:

	Airline	Image	IATA	ICAO	Callsign	Primary hubs, Secondary hubs	Founded	Notes	Primary Hubs, Secondary Hubs
0	Alaska Airlines	NaN	AS	ASA	ALASKA	Seattle/TacomaAnchoragePortland (OR)San Franci...	1932.0	Founded as McGee Airways and commenced operati...	NaN
1	Allegiant Air	NaN	G4	AAY	ALLEGiant	Las VegasCincinnatiFort Walton BeachIndianapol...	1997.0	Founded as WestJet Express and commenced opera...	NaN
2	American Airlines	NaN	AA	AAL	AMERICAN	WorthCharlotteChicago-O'HareLos An...	1926.0	Founded as American Airways and commenced oper...	NaN
3	Avelo Airlines	NaN	XP	VXP	AVELO	BurbankNew HavenOrlando	1987.0	First did business as Casino Express Airlines ...	NaN
4	Breeze Airways	NaN	MX	MXV	MOXY	CharlestonHartfordNew OrleansNorfolkProvoTampa	2018.0	NaN	NaN
...	...	...	...	...	...	...	...	...	...
135	Lifestar	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
136	Life Lion	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
137	Comco	NaN	NaN	NaN	NaN	NaN	2002.0	NaN	NaN
138	Janet	NaN	NaN	WWW	JANET	NaN	1972.0	NaN	Las Vegas
139	Justice Prisoner and Alien Transportation System	NaN	NaN	JUD	JUSTICE	NaN	1980.0	Commenced operations in 1995.	Oklahoma City

140 rows × 9 columns

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

```
In [28]: wiki_df = wiki_tables[['IATA', "Founded"]]
wiki_df
```

Out[28]:

	IATA	Founded
0	AS	1932.0
1	G4	1997.0
2	AA	1926.0
3	XP	1987.0
4	MX	2018.0
...	...	...
135	NaN	NaN
136	NaN	NaN
137	NaN	2002.0
138	NaN	1972.0
139	NaN	1980.0

140 rows × 2 columns

In [29]:

# 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")

In [30]:

df

Out[30]:

	id	Airline	Flight	AirportFrom	AirportTo	DayOfWeek	Time	Length	Delay	ident	...	iata_code	airport_re
0	4	AA	2466	SFO	DFW	3	20	195	1	KSFO	...	SFO	3878
1	231	AA	526	SFO	DFW	3	360	215	0	KSFO	...	SFO	3878
2	234	AA	552	SFO	MIA	3	360	315	1	KSFO	...	SFO	3878
3	905	AA	810	SFO	ORD	3	385	255	0	KSFO	...	SFO	3878
4	1739	AA	24	SFO	JFK	3	425	325	1	KSFO	...	SFO	3878
...	...	...	...	...	...	...	...	...	...	...	...	...	...
434919	497838	9E	4292	LWB	JFK	3	890	110	1	KLWB	...	LWB	20390
434920	516333	9E	4292	LWB	JFK	4	890	110	0	KLWB	...	LWB	20390
434921	534123	9E	4292	LWB	JFK	5	890	110	0	KLWB	...	LWB	20390
434922	69058	9E	3752	ABR	MSP	7	410	76	1	KABR	...	ABR	3358
434923	189396	9E	3752	ABR	MSP	7	410	76	0	KABR	...	ABR	3358

434924 rows × 26 columns

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 scraping and collate it in a table.

[https://en.wikipedia.org/wiki/List\\_of\\_the\\_busiest\\_airports\\_in\\_the\\_United\\_States](https://en.wikipedia.org/wiki/List_of_the_busiest_airports_in_the_United_States)

In [31]:

# Now Lets use the web scrapping to import the data frome the wikipedia.  
ur12 = "https://en.wikipedia.org/wiki/List\_of\_the\_busiest\_airports\_in\_the\_United\_States"  
table = pd.read\_html(ur12)

In [32]:

table

Out[32]:		Rank(2021)	Airports (large hubs)		IATACode	\		
	0	1	Hartsfield-Jackson Atlanta International Airport	ATL				
	1	2	Dallas/Fort Worth International Airport	DFW				
	2	3	Denver International Airport	DEN				
	3	4	O'Hare International Airport	ORD				
	4	5	Los Angeles International Airport	LAX				
	5	6	Charlotte Douglas International Airport	CLT				
	6	7	Orlando International Airport	MCO				
	7	8	Harry Reid International Airport	LAS				
	8	9	Phoenix Sky Harbor International Airport	PHX				
	9	10	Miami International Airport	MIA				
	10	11	Seattle-Tacoma International Airport	SEA				
	11	12	George Bush Intercontinental Airport	IAH				
	12	13	John F. Kennedy International Airport	JFK				
	13	14	Newark Liberty International Airport	EWR				
	14	15	Fort Lauderdale-Hollywood International Airport	FLL				
	15	16	Minneapolis-Saint Paul International Airport	MSP				
	16	17	San Francisco International Airport	SFO				
	17	18	Detroit Metropolitan Airport	DTW				
	18	19	Logan International Airport	BOS				
	19	20	Salt Lake City International Airport	SLC				
	20	21	Philadelphia International Airport	PHL				
	21	22	Baltimore/Washington International Airport	BWI				
	22	23	Tampa International Airport	TPA				
	23	24	San Diego International Airport	SAN				
	24	25	LaGuardia Airport	LGA				
	25	26	Midway International Airport	MDW				
	26	27	Nashville International Airport	BNA				
	27	28	Washington Dulles International Airport	IAD				
	28	29	Ronald Reagan Washington National Airport	DCA				
	29	30	Austin-Bergstrom International Airport	AUS				
		Major cities served	State	2021[3]	2020[4]	2019[5]	\	
	0	Atlanta	GA	36676010	20559866	53505795		
	1	Dallas & Fort Worth	TX	30005266	18593421	35778573		
	2	Denver	CO	28645527	16243216	33592945		
	3	Chicago	IL	26350976	14606034	40871223		
	4	Los Angeles	CA	23663410	14055777	42939104		
	5	Charlotte	NC	20900875	12952869	24199688		
	6	Orlando	FL	19618838	10467728	24562271		
	7	Las Vegas	NV	19160342	10584059	24728361		
	8	Phoenix	AZ	18940287	10531436	22433552		
	9	Miami	FL	17500096	8786007	21421031		
	10	Seattle	WA	17430195	9462411	25001762		
	11	Houston	TX	16242821	8682558	21905309		
	12	New York City	NY	15273342	8269819	31036655		
	13	Newark & New York City	NJ	14514049	7985474	23160763		
	14	Fort Lauderdale & Hollywood	FL	13598994	8015744	17950989		
	15	Minneapolis & Saint Paul	MN	12211409	7069720	19192917		
	16	San Francisco	CA	11725347	7745057	27779230		
	17	Detroit	MI	11517696	6822324	18143040		
	18	Boston	MA	10909817	6035452	20699377		
	19	Salt Lake City	UT	10795906	5753239	12840841		
	20	Philadelphia	PA	9820222	5753239	16006389		
	21	Baltimore & Washington, D.C.	MD	9253561	5451355	13284687		
	22	Tampa	FL	8847197	4966775	10978756		
	23	San Diego	CA	7836360	4637856	12648692		
	24	New York City	NY	7827307	4147116	15393601		
	25	Chicago	IL	7680617	4236603	10081781		
	26	Nashville	TN	7594049	4013995	8935654		
	27	Washington, D.C.	VA	7227875	3862658	11884117		
	28	Washington, D.C.	VA	6731737	3573489	11595454		
	29	Austin	TX	6666215	3141505	8683711		
		2018[6]	2017[7]	2016[8]	2015[9]	2014[10]	2013[11]	2012[12]
	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	
28	11367176	11506310	11470854	11242375	10057794	9838034	9462231	
29	7921797	6973115	6095545	5797547	5219982	4900959	4606252	,
	Rank(2021)				Airports (medium hubs)	IATACode	\	
0	31				Dallas Love Field	DAL		
1	32			Daniel K. Inouye International Airport	HNL			
2	33			Portland International Airport	PDX			
3	34			William P. Hobby Airport	HOU			
4	35			Southwest Florida International Airport	RSW			
5	36			St. Louis Lambert International Airport	STL			
6	37			Sacramento International Airport	SMF			
7	38			Luis Muñoz Marín International Airport	SJU			
8	39			Raleigh-Durham International Airport	RDU			
9	40	Louis Armstrong New Orleans International Airport			MSY			
10	41			Oakland International Airport	OAK			
11	42			John Wayne Airport	SNA			
12	43			Kansas City International Airport	MCI			
13	44			San Antonio International Airport	SAT			
14	45	Norman Y. Mineta San José International Airport			SJC			
15	46	Cleveland Hopkins International Airport			CLE			
16	47	Indianapolis International Airport			IND			
17	48	Pittsburgh International Airport			PIT			
18	49	Cincinnati/Northern Kentucky International Air...			CVG			
19	50			Kahului Airport	OGG			
20	51	John Glenn Columbus International Airport			CMH			
21	52	Palm Beach International Airport			PBI			
22	53	Jacksonville International Airport			JAX			
23	54	Bradley International Airport			BDL			
24	55	Milwaukee Mitchell International Airport			MKE			
25	56	Ontario International Airport			ONT			
26	57	Ted Stevens Anchorage International Airport			ANC			
27	58	Charleston International Airport			CHS			
28	59	Hollywood Burbank Airport			BUR			
29	60	Eppley Airfield			OMA			
30	61	Boise Airport			BOI			
31	62	Memphis International Airport			MEM			
32	63	Reno-Tahoe International Airport			RNO			
33	64	Albuquerque International Sunport			ABQ			
34	65	Norfolk International Airport			ORF			

		City served	State	2021[3]	2020[4]	2019[5]	2018[6]	\
0		Dallas	TX	6487563	3669930	8408457	8134848	
1		Honolulu	HI	5830928	3126391	9988678	9578505	
2		Portland	OR	5759879	3455877	9797408	9940866	
3		Houston	TX	5560780	3127178	7069614	6937061	
4		Fort Myers	FL	5080805	2947139	5144467	4719568	
5		St. Louis	MO	5070471	3041765	7946986	7822274	
6		Sacramento	CA	4760275	2710342	6454413	6031630	
7		San Juan	PR	4738725	2362851	4590117	4033412	
8		Raleigh	NC	4311049	2337496	6919429	6416822	
9		New Orleans	LA	4017147	2632606	6717105	6565482	
10		Oakland	CA	4011953	2271294	6560230	6798321	
11		Orange County	CA	3807205	1824836	5153276	5317149	
12		Kansas City	MO	3795290	2167616	5759419	5935131	
13		San Antonio	TX	3677643	1919958	5022980	4844427	
14		San Jose	CA	3619690	2283186	7828885	7140616	
15		Cleveland	OH	3552402	1990156	4894541	4836580	
16		Indianapolis	IN	3487100	1989126	4709183	4695040	
17		Pittsburgh	PA	3069259	1742406	4715947	4670033	
18		Cincinnati & Covington	OH/KY	3050597	1729395	4413457	4269258	
19		Kahului	HI	2933315	1135141	3791807	3572133	
20		Columbus	OH	2825259	1577596	4172067	4054572	
21		West Palm Beach	FL	2567897	1518732	3460429	3263042	
22		Jacksonville	FL	2425685	1367501	3479923	3118540	

23	Hartford	CT	2273259	1150033	3323614	3330734
24	Milwaukee	WI	2231010	1263385	3374073	3548817
25	Ontario	CA	2201528	1237946	2723002	2499171
26	Anchorage	AK	2184959	1157301	2713843	2642901
27	Charleston	SC	2015277	944660	2375868	2192893
28	Burbank	CA	1942417	1056838	2988720	2680240
29	Omaha	NE	1829912	1036245	2455274	2457087
30	Boise	ID	1809000	991241	2057750	1943181
31	Memphis	TN	1793073	1015981	2318442	2213083
32	Reno	NV	1781785	976937	2162250	2048916
33	Albuquerque	NM	1688646	868922	2641450	2647269
34	Norfolk	VA	1658024	884882	1990864	1846031

	2017[7]	2016[8]	2015[9]	2014[10]	2013[11]	2012[12]
0	7876769	7554596	7040921	4522341	4023779	3902628
1	9743989	9656340	9656340	9463000	9466995	9225848
2	9435473	9071154	8340234	7878760	7452603	7142620
3	6741870	6285181	5937944	5800726	5377050	5043737
4	4461304	4350650	4231134	4025959	3788870	3634152
5	7372805	6793076	6239231	6108758	6216104	6208750
6	5460526	4969366	4816440	4384616	4255145	4357899
7	4163587	4343354	4218785	4150828	4103197	4204478
8	5851004	5401714	4954717	4673869	4482016	4490374
9	6005527	5569705	5329696	4870569	4576539	4293624
10	6530308	5934639	5506672	5069257	4770716	4926683
11	5195047	5217242	4945175	4584147	4540628	4381172
12	5744918	5391557	5135127	4982722	4836221	4866850
13	4521611	4179994	4091389	4046856	4005874	4036625
14	6225148	5321603	4885690	4621003	4315839	4077654
15	4562740	4205739	4083476	3686315	4375448	4346941
16	4376432	4216766	3889567	3605908	3535015	3586422
17	4327431	3986114	3890677	3827860	3812460	3892338
18	3926158	3269979	3036697	2874684	2776377	2937850
19	3442189	3352813	3220753	3019338	2955304	2861278
20	3765007	3567864	3312496	3115501	3063822	3095575
21	3166532	3100624	3113485	2926242	2844507	2796359
22	2759067	2799587	2716465	2589198	2549070	2579023
23	3214976	2982194	2926047	2913380	2681181	2647610
24	3452544	3383271	3229876	3228607	3214811	3710384
25	2247645	2127387	2089801	2037346	1970538	2142393
26	2556188	2563524	2525876	2381826	2325030	2249717
27	1945699	1811695	1669988	1539326	1441415	1283970
28	2402106	2077892	1973897	1928491	1918011	2027203
29	2303223	2127387	2046155	2020354	1975339	2018738
30	1777642	1633507	1487777	1378352	1313741	1307505
31	2102739	2016089	1873716	1800268	2301003	3359668
32	1953028	1771864	1669876	1611572	1671926	1685333
33	2412328	2341719	2323883	2354184	2477783	2630574
34	1694329	1602631	1515200	1488114	1560754	1651440

	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
4	NaN
5	NaN
6	NaN
7	NaN
8	NaN
9	NaN
10	NaN
11	NaN
12	NaN
13	NaN
14	NaN
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

0	Rank	Rank change	Location	Airport name \
1	Rank	Rank change	Location	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...	

0	Location	IATA Code	Traffic	Aircraft \
1	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	EWB	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	Rank	Airport name \
Rank	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	EWK	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

	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	EWK	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
Rank	Airport name \				
Rank	Airport name				
0	1	Memphis International Airport			
1	2	Ted Stevens Anchorage International Airport			
2	3	Louisville Muhammad Ali 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 .navavbar ul{display:inline-block;white-space:nowrap;line-height:inherit}.mw-parser-output .navbar-bracket s::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) Austin (Aus...  
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 .navavbar ul{display:inline-block;white-space:nowrap;line-height:inherit}.mw-parser-output .navbar-bracket s::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) Austin (Aus...  
1 Statistics

,  
vteList of the busiest airports in North America \

0	Sovereign states
1	Dependencies and other 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...
---	---

Airports worldwide \

0	Busiest airports by continent and country
1	Africa
2	Asia
3	Europe
4	North America
5	Oceania
6	South America
7	By region
8	Airlines
9	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...]
```

```
In [34]: table[0]['traffic_Chg19_20'] = table[0]['2020[4]'] - table[0]['2019[5]']
```

```
In [35]: table[0]['traffic_Chg18_19'] = table[0]['2019[5]'] - table[0]['2018[6]']
table[0]['hubs'] = str('large_hub')
```

```
In [36]: table[0] = table[0][['IATACode', 'traffic_Chg19_20', 'traffic_Chg18_19', 'hubs']]
table[0]
```

Out[36]:

	IATACode	traffic_Chg19_20	traffic_Chg18_19	hubs
0	ATL	-32945929	1639998	large_hub
1	DFW	-17185152	2956774	large_hub
2	DEN	-17349729	2230004	large_hub
3	ORD	-26265189	997296	large_hub
4	LAX	-28883327	315054	large_hub
5	CLT	-11246819	1917739	large_hub
6	MCO	-14094543	1359791	large_hub
7	LAS	-14144302	933349	large_hub
8	PHX	-11902116	810972	large_hub
9	MIA	-12635024	399391	large_hub
10	SEA	-15539351	976854	large_hub
11	IAH	-13222751	747911	large_hub
12	JFK	-22766836	415886	large_hub
13	EWR	-15175289	363161	large_hub
14	FLL	-9935245	338658	large_hub
15	MSP	-12123197	830975	large_hub
16	SFO	-20034173	-11487	large_hub
17	DTW	-11320716	706203	large_hub
18	BOS	-14663925	692856	large_hub
19	SLC	-7087602	614111	large_hub
20	PHL	-10253150	713719	large_hub
21	BWI	-7833332	-87129	large_hub
22	TPA	-6011981	610242	large_hub
23	SAN	-8010836	474468	large_hub
24	LGA	-11246485	335100	large_hub
25	MDW	-5845178	-596237	large_hub
26	BNA	-4921659	918307	large_hub
27	IAD	-8021459	262494	large_hub
28	DCA	-8021965	228278	large_hub
29	AUS	-5542206	761914	large_hub

In [40]: `table[1].head()`

Out[40]:

	Rank(2021)	Airports (medium hubs)	IATACode	City served	State	2021[3]	2020[4]	2019[5]	2018[6]	2017[7]	2016[8]	2015[9]	:
0	31	Dallas Love Field	DAL	Dallas	TX	6487563	3669930	8408457	8134848	7876769	7554596	7040921	
1	32	Daniel K. Inouye International Airport	HNL	Honolulu	HI	5830928	3126391	9988678	9578505	9743989	9656340	9656340	
2	33	Portland International Airport	PDX	Portland	OR	5759879	3455877	9797408	9940866	9435473	9071154	8340234	
3	34	William P. Hobby Airport	HOU	Houston	TX	5560780	3127178	7069614	6937061	6741870	6285181	5937944	
4	35	Southwest Florida International Airport	RSW	Fort Myers	FL	5080805	2947139	5144467	4719568	4461304	4350650	4231134	



```
In [42]: table[1]['traffic_Chg19_20'] = table[1]['2020[4]'] - table[1]['2019[5]']
table[1]['traffic_Chg18_19'] = table[1]['2019[5]'] - table[1]['2018[6]']
table[1]['hubs'] = str('Medium_hub')
```

```
In [43]: table[1] = table[1][['IATACode', 'traffic_Chg19_20', 'traffic_Chg18_19', 'hubs']]
table[1]
```

Out[43]:

	IATACode	traffic_Chg19_20	traffic_Chg18_19	hubs
0	DAL	-4738527	273609	Medium_hub
1	HNL	-6862287	410173	Medium_hub
2	PDX	-6341531	-143458	Medium_hub
3	HOU	-3942436	132553	Medium_hub
4	RSW	-2197328	424899	Medium_hub
5	STL	-4905221	124712	Medium_hub
6	SMF	-3744071	422783	Medium_hub
7	SJU	-2227266	556705	Medium_hub
8	RDU	-4581933	502607	Medium_hub
9	MSY	-4084499	151623	Medium_hub
10	OAK	-4288936	-238091	Medium_hub
11	SNA	-3328440	-163873	Medium_hub
12	MCI	-3591803	-175712	Medium_hub
13	SAT	-3103022	178553	Medium_hub
14	SJC	-5545699	688269	Medium_hub
15	CLE	-2904385	57961	Medium_hub
16	IND	-2720057	14143	Medium_hub
17	PIT	-2973541	45914	Medium_hub
18	CVG	-2684062	144199	Medium_hub
19	OGG	-2656666	219674	Medium_hub
20	CMH	-2594471	117495	Medium_hub
21	PBI	-1941697	197387	Medium_hub
22	JAX	-2112422	361383	Medium_hub
23	BDL	-2173581	-7120	Medium_hub
24	MKE	-2110688	-174744	Medium_hub
25	ONT	-1485056	223831	Medium_hub
26	ANC	-1556542	70942	Medium_hub
27	CHS	-1431208	182975	Medium_hub
28	BUR	-1931882	308480	Medium_hub
29	OMA	-1419029	-1813	Medium_hub
30	BOI	-1066509	114569	Medium_hub
31	MEM	-1302461	105359	Medium_hub
32	RNO	-1185313	113334	Medium_hub
33	ABQ	-1772528	-5819	Medium_hub
34	ORF	-1105982	144833	Medium_hub

```
In [44]: # Lets first merge all wikipedia table.
wiki_data = [table[0],table[1]]
wiki_data = pd.concat(wiki_data, ignore_index=True)
wiki_data
```



Out[44]:

	IATACode	traffic_Chg19_20	traffic_Chg18_19	hubs
0	ATL	-32945929	1639998	large_hub
1	DFW	-17185152	2956774	large_hub
2	DEN	-17349729	2230004	large_hub
3	ORD	-26265189	997296	large_hub
4	LAX	-28883327	315054	large_hub
...	...	...	...	...
60	BOI	-1066509	114569	Medium_hub
61	MEM	-1302461	105359	Medium_hub
62	RNO	-1185313	113334	Medium_hub
63	ABQ	-1772528	-5819	Medium_hub
64	ORF	-1105982	144833	Medium_hub

65 rows × 4 columns

In [45]:

```
# 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")
```

In [46]:

```
final_df
```

Out[46]:

	id	Airline	Flight	AirportFrom	AirportTo	DayOfWeek	Time	Length	Delay	ident	...	width_ft	surface	lic
0	4	AA	2466	SFO	DFW	3	20	195	1	KSFO	...	200.0	ASP	
1	231	AA	526	SFO	DFW	3	360	215	0	KSFO	...	200.0	ASP	
2	234	AA	552	SFO	MIA	3	360	315	1	KSFO	...	200.0	ASP	
3	905	AA	810	SFO	ORD	3	385	255	0	KSFO	...	200.0	ASP	
4	1739	AA	24	SFO	JFK	3	425	325	1	KSFO	...	200.0	ASP	
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
364272	506267	9E	4052	DAL	MEM	4	370	90	0	KDAL	...	150.0	CON	
364273	512858	9E	3704	DAL	MEM	4	705	92	1	KDAL	...	150.0	CON	
364274	518247	9E	4060	DAL	MEM	4	990	90	0	KDAL	...	150.0	CON	
364275	524678	9E	4052	DAL	MEM	5	370	90	1	KDAL	...	150.0	CON	
364276	530841	9E	3704	DAL	MEM	5	705	92	0	KDAL	...	150.0	CON	

364277 rows × 30 columns



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

In [47]:

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

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 364277 entries, 0 to 364276
Data columns (total 30 columns):
#   Column                Non-Null Count  Dtype
---  -
0   id                    364277 non-null  int64
1   Airline               364277 non-null  object
2   Flight               364277 non-null  int64
3   AirportFrom          364277 non-null  object
4   AirportTo            364277 non-null  object
5   DayOfWeek            364277 non-null  int64
6   Time                 364277 non-null  int64
7   Length               364277 non-null  int64
8   Delay                364277 non-null  int64
9   ident                364277 non-null  object
10  type                 364277 non-null  object
11  name                 364277 non-null  object
12  latitude_deg         364277 non-null  float64
13  longitude_deg        364277 non-null  float64
14  elevation_ft         364277 non-null  float64
15  scheduled_service    364277 non-null  object
16  iata_code            364277 non-null  object
17  airport_ref          364277 non-null  int64
18  airport_ident        364277 non-null  object
19  length_ft           364277 non-null  float64
20  width_ft            364277 non-null  float64
21  surface              364277 non-null  object
22  lighted              364277 non-null  int64
23  closed              364277 non-null  int64
24  IATA                 364277 non-null  object
25  Founded              364277 non-null  float64
26  IATACode             364277 non-null  object
27  traffic_Chg19_20     364277 non-null  int64
28  traffic_Chg18_19     364277 non-null  int64
29  hubs                364277 non-null  object
dtypes: float64(6), int64(11), object(13)
memory usage: 86.2+ MB

```

```
In [48]: final_df.drop(['id','AirportFrom','airport_ident','iata_code','AirportTo','surface', 'ident','IATA', 'IA'
```

```
In [49]: final_df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 364277 entries, 0 to 364276
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Airline               364277 non-null  object
1   Flight               364277 non-null  int64
2   DayOfWeek            364277 non-null  int64
3   Time                 364277 non-null  int64
4   Length               364277 non-null  int64
5   Delay                364277 non-null  int64
6   type                 364277 non-null  object
7   latitude_deg         364277 non-null  float64
8   longitude_deg        364277 non-null  float64
9   elevation_ft         364277 non-null  float64
10  scheduled_service    364277 non-null  object
11  airport_ref          364277 non-null  int64
12  length_ft           364277 non-null  float64
13  width_ft            364277 non-null  float64
14  lighted              364277 non-null  int64
15  closed              364277 non-null  int64
16  Founded              364277 non-null  float64
17  traffic_Chg19_20     364277 non-null  int64
18  traffic_Chg18_19     364277 non-null  int64
19  hubs                364277 non-null  object
dtypes: float64(6), int64(10), object(4)
memory usage: 58.4+ MB

```

```
In [50]: # Now Lets check the null value and treat them.
final_df.isnull().sum()
```

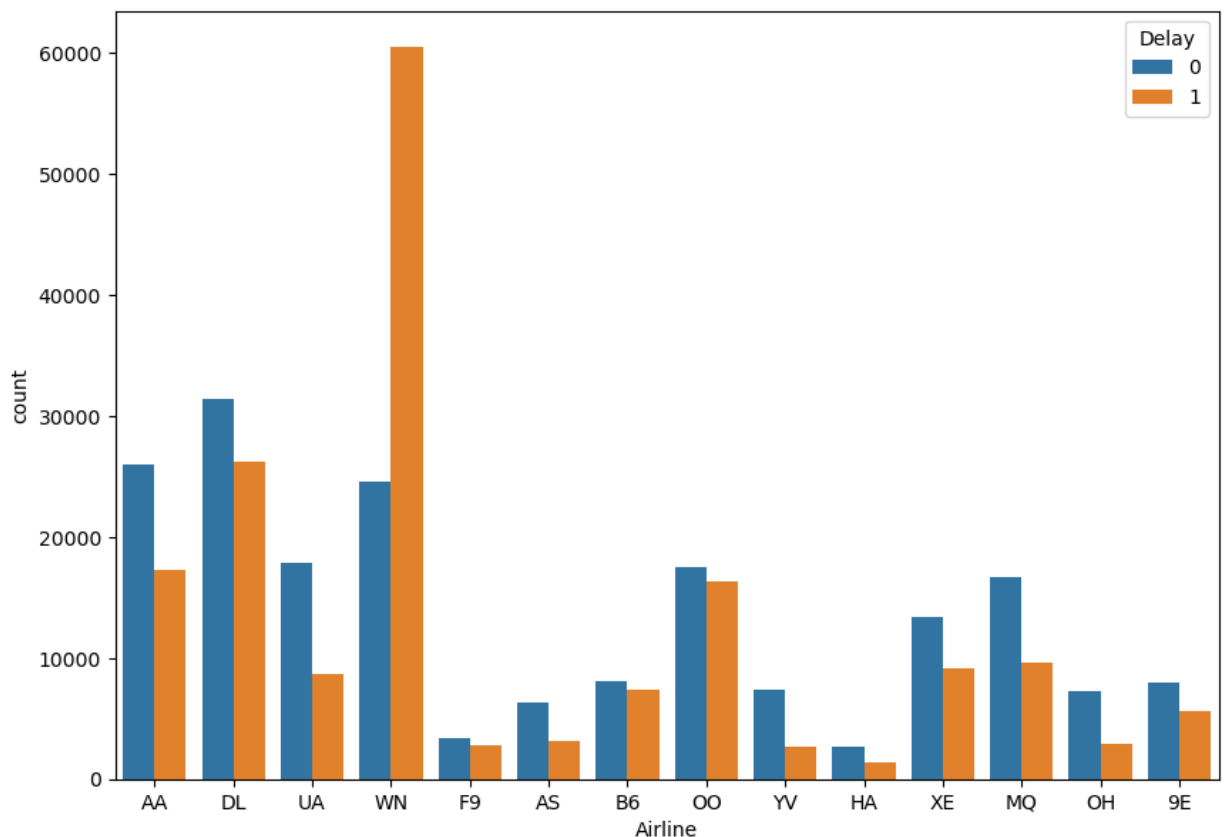
```
Out[50]: Airline      0
Flight      0
DayOfWeek   0
Time        0
Length      0
Delay       0
type        0
latitude_deg 0
longitude_deg 0
elevation_ft 0
scheduled_service 0
airport_ref 0
length_ft   0
width_ft    0
lighted     0
closed      0
Founded     0
traffic_Chg19_20 0
traffic_Chg18_19 0
hubs        0
dtype: int64
```

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

**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.**

```
In [51]: plt.figure(figsize=(10,7))
sns.countplot(final_df['Airline'], hue= final_df['Delay'])
```

```
Out[51]: <AxesSubplot:xlabel='Airline', ylabel='count'>
```

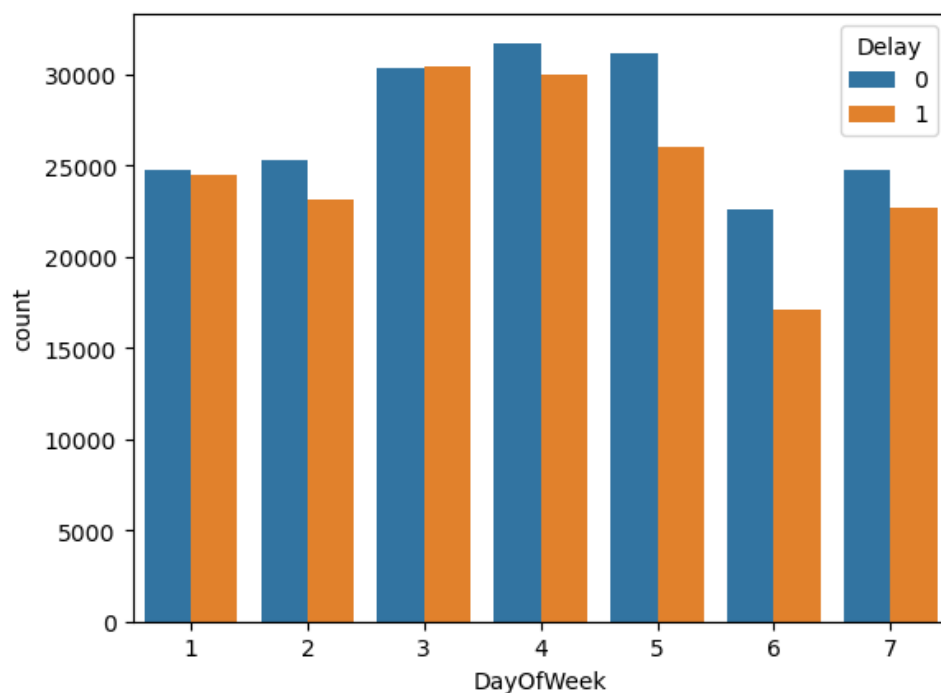


Airline code **WN** represent the **southwest airlines**. The graph clear show that **70% of flight of south west airline is delayed**

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

```
In [52]: sns.countplot(final_df['DayOfWeek'], hue= final_df['Delay'])
```

```
Out[52]: <AxesSubplot:xlabel='DayOfWeek', ylabel='count'>
```



on 6th day of the week we have least delayed flight.

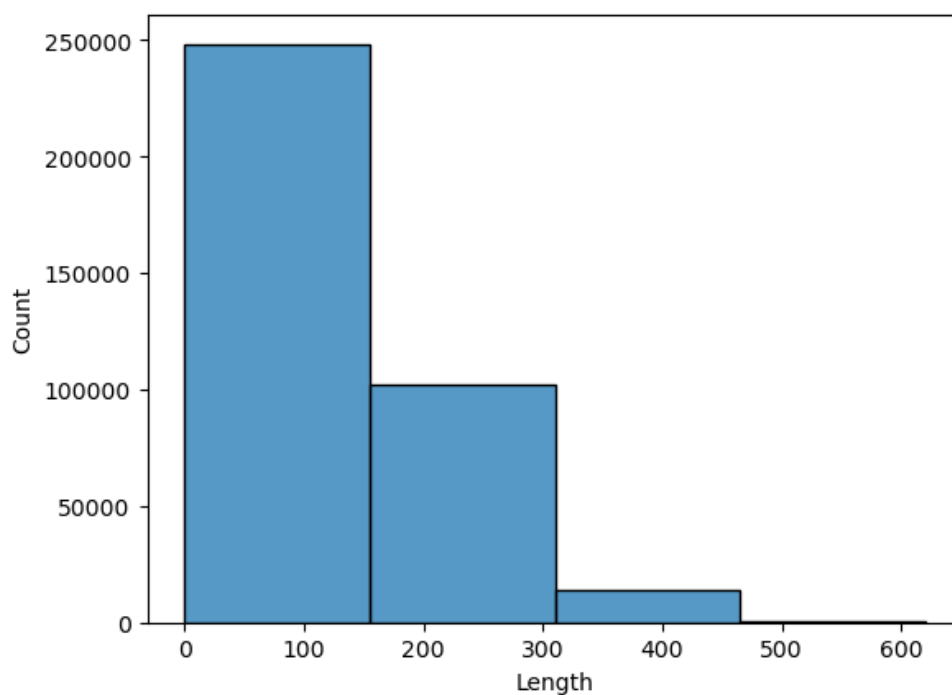
**c. Which airlines should be recommended for short-, medium-, and long-distance travel?**

```
In [53]: final_df['Length'].max()
```

```
Out[53]: 620
```

```
In [54]: sns.histplot(final_df['Length'], bins = 4)
```

```
Out[54]: <AxesSubplot:xlabel='Length', ylabel='Count'>
```



```
In [55]: final_df['Airline'][final_df['Length'] < 180].value_counts()
```

```
Out[55]: WN      72531
         DL      40149
         OO      32192
         MQ      25472
         AA      24637
         XE      21576
         UA      15761
         9E      13526
         B6       9766
         YV       9756
         OH       9750
         AS       5996
         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.**

```
In [56]: final_df['Airline'][(final_df['Length']>180) & (final_df['Length']<360)].value_counts()
```

```
Out[56]: DL      16510
         AA      16112
         WN      11427
         UA       9385
         B6       4847
         AS       2822
         OO       1576
         F9       1075
         XE        975
         HA        751
         MQ        604
         OH        433
         YV        246
         9E         47
         Name: Airline, dtype: int64
```

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

```
In [57]: final_df['Airline'][final_df['Length']>360].value_counts()
```

```
Out[57]: UA      1304
         AA      1081
         DL       842
         B6       822
         AS       540
         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?**

```
In [58]: final_df[final_df['Length']>360].describe().T
```

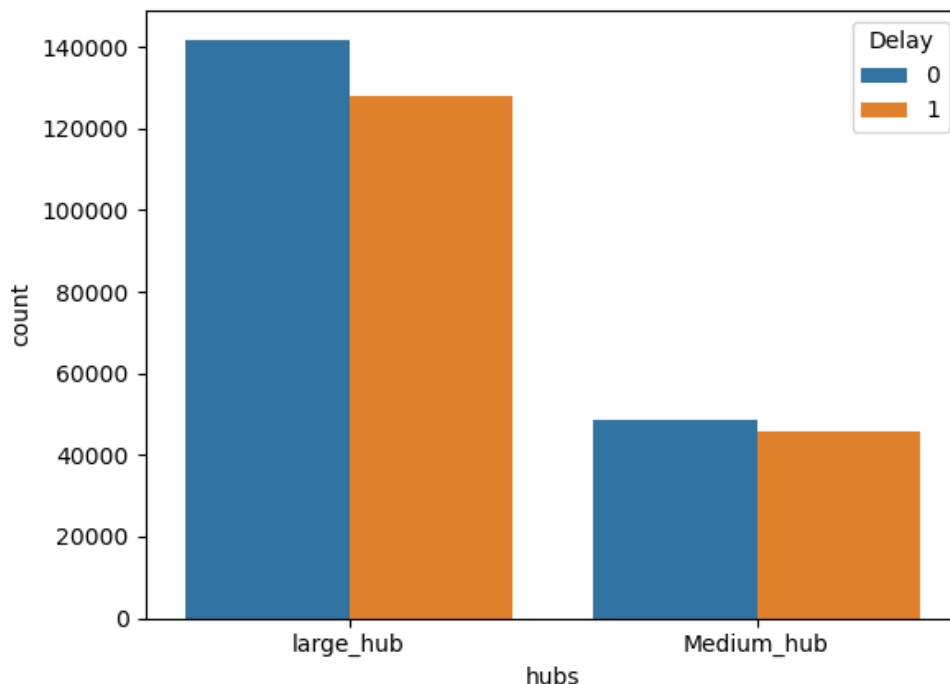
Out[58]:

	count	mean	std	min	25%	50%	75%	
<b>Flight</b>	4893.0	5.592187e+02	7.509137e+02	1.000000e+00	5.900000e+01	2.090000e+02	8.490000e+02	3.76000
<b>DayOfWeek</b>	4893.0	4.003270e+00	1.925851e+00	1.000000e+00	2.000000e+00	4.000000e+00	6.000000e+00	7.00000
<b>Time</b>	4893.0	8.287658e+02	2.857339e+02	1.000000e+02	5.500000e+02	8.850000e+02	1.080000e+03	1.43500
<b>Length</b>	4893.0	3.957466e+02	4.073634e+01	3.610000e+02	3.740000e+02	3.850000e+02	4.000000e+02	6.20000
<b>Delay</b>	4893.0	4.161046e-01	4.929617e-01	0.000000e+00	0.000000e+00	0.000000e+00	1.000000e+00	1.00000
<b>latitude_deg</b>	4893.0	3.893579e+01	6.497714e+00	1.843940e+01	3.894450e+01	4.063945e+01	4.193851e+01	6.11744
<b>longitude_deg</b>	4893.0	-8.828553e+01	2.515070e+01	-1.579242e+02	-9.703800e+01	-7.377932e+01	-7.377932e+01	-6.60018
<b>elevation_ft</b>	4893.0	2.385510e+02	7.470823e+02	8.000000e+00	1.300000e+01	1.800000e+01	1.250000e+02	5.43100
<b>airport_ref</b>	4893.0	3.741812e+03	4.890171e+02	3.384000e+03	3.602000e+03	3.622000e+03	3.670000e+03	6.38400
<b>length_ft</b>	4893.0	1.005470e+04	2.102188e+03	4.892000e+03	7.861000e+03	1.100000e+04	1.207900e+04	1.20790
<b>width_ft</b>	4893.0	1.716636e+02	2.506597e+01	1.000000e+02	1.500000e+02	1.500000e+02	2.000000e+02	2.00000
<b>lighted</b>	4893.0	9.513591e-01	2.151382e-01	0.000000e+00	1.000000e+00	1.000000e+00	1.000000e+00	1.00000
<b>closed</b>	4893.0	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.00000
<b>Founded</b>	4893.0	1.939004e+03	2.692440e+01	1.924000e+03	1.926000e+03	1.926000e+03	1.932000e+03	1.99800
<b>traffic_Chg19_20</b>	4893.0	-1.763080e+07	6.656975e+06	-3.294593e+07	-2.276684e+07	-1.718515e+07	-1.414430e+07	-1.55654
<b>traffic_Chg18_19</b>	4893.0	5.823171e+05	4.670072e+05	-1.434580e+05	4.101730e+05	4.158860e+05	6.928560e+05	2.95677

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

```
In [59]: sns.countplot(final_df['hubs'], hue = final_df['Delay'])
```

```
Out[59]: <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.

#### 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

```
In [60]: 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=194730.438, 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

```
In [61]: 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=200241.469, 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

```
In [62]: 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')
```

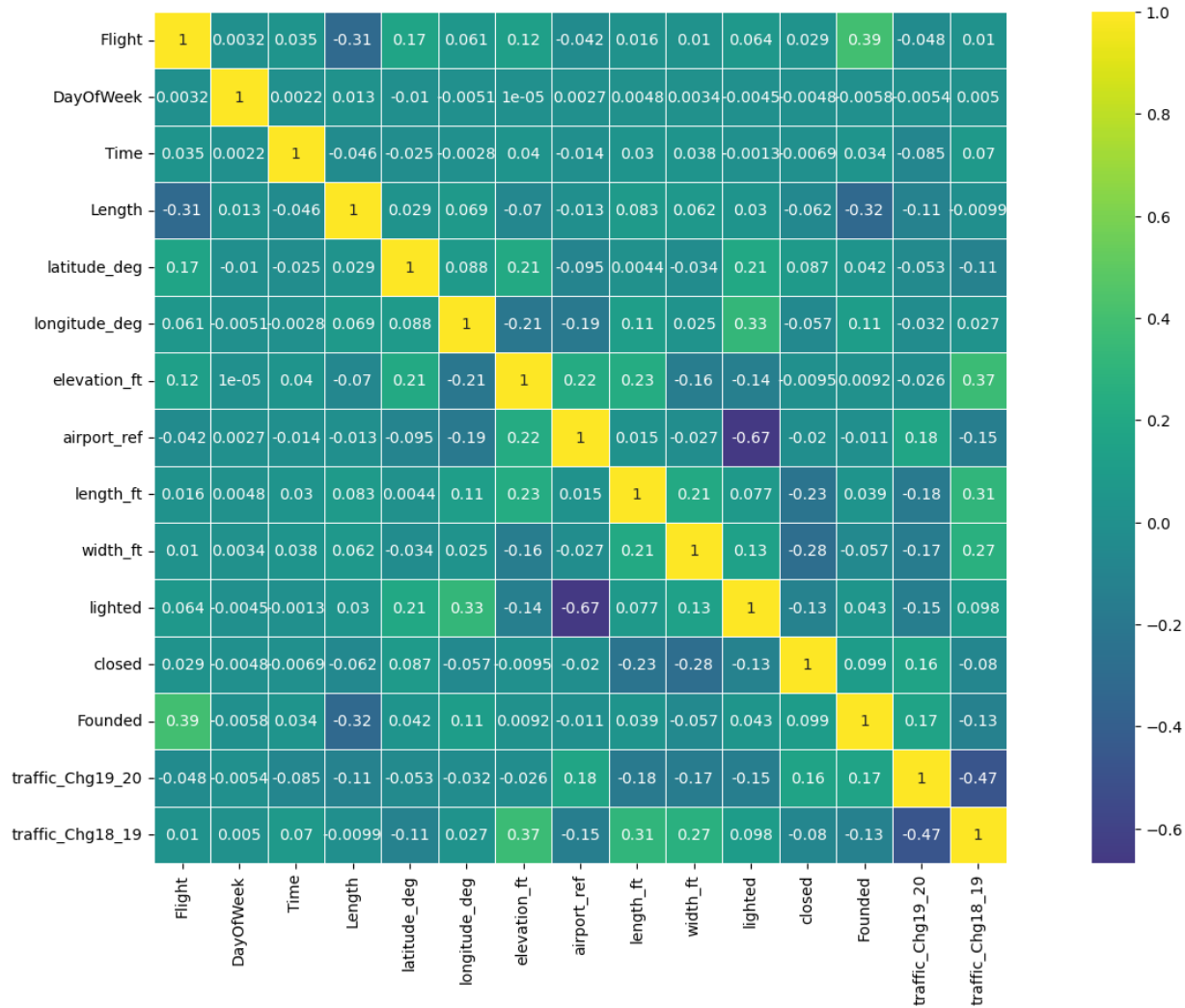
```
stat=-0.002, p=0.203
Probably independent
```

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

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

```
In [63]: predictor = final_df.drop(['Delay'], axis=1)
corr = predictor.corr()
```

```
In [64]: plt.figure(figsize=(20,10))
sns.heatmap(corr, center=0, linewidths=.5, square = True , annot = True , cmap='viridis')
plt.show()
```



# Machine learning

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

```
In [65]: # Before applying the one hot encodding or the label encoding first we check all feature data type.
final_df.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 364277 entries, 0 to 364276
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Airline                364277 non-null object
1   Flight                 364277 non-null int64
2   DayOfWeek              364277 non-null int64
3   Time                   364277 non-null int64
4   Length                 364277 non-null int64
5   Delay                  364277 non-null int64
6   type                   364277 non-null object
7   latitude_deg           364277 non-null float64
8   longitude_deg          364277 non-null float64
9   elevation_ft           364277 non-null float64
10  scheduled_service      364277 non-null object
11  airport_ref            364277 non-null int64
12  length_ft              364277 non-null float64
13  width_ft               364277 non-null float64
14  lighted                364277 non-null int64
15  closed                 364277 non-null int64
16  Founded                364277 non-null float64
17  traffic_Chg19_20       364277 non-null int64
18  traffic_Chg18_19       364277 non-null int64
19  hubs                   364277 non-null object
dtypes: float64(6), int64(10), object(4)
memory usage: 58.4+ MB
```



```
In [66]: final_df['Airline'].value_counts()
```

```
Out[66]: WN      85067
DL       57720
AA       43261
OO       33843
UA       26535
MQ       26308
XE       22566
B6       15497
9E       13573
OH       10211
YV       10002
AS        9477
F9        6180
HA        4037
Name: Airline, dtype: int64
```

```
In [67]: final_df['type'].value_counts()
```

```
Out[67]: large_airport    342705
medium_airport    21572
Name: type, dtype: int64
```

```
In [68]: final_df['scheduled_service'].value_counts()
```

```
Out[68]: yes    364277
Name: scheduled_service, dtype: int64
```

```
In [69]: final_df['hubs'].value_counts()
```

```
Out[69]: large_hub    269953
Medium_hub    94324
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.

```
In [70]: final_df = final_df.drop(['scheduled_service'], axis=1)
```

```
In [71]: # Now using the ordinal encoder.
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
```

```
In [72]: 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'])
```

```
In [73]: final_df.head()
```

```
Out[73]:
```

	Airline	Flight	DayOfWeek	Time	Length	Delay	type	latitude_deg	longitude_deg	elevation_ft	airport_ref	length_ft
0	1	2466	3	20	195	1	0	37.618999	-122.375	13.0	3878	7500.0
1	1	526	3	360	215	0	0	37.618999	-122.375	13.0	3878	7500.0
2	1	552	3	360	315	1	0	37.618999	-122.375	13.0	3878	7500.0
3	1	810	3	385	255	0	0	37.618999	-122.375	13.0	3878	7500.0
4	1	24	3	425	325	1	0	37.618999	-122.375	13.0	3878	7500.0

## 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)

**e. Compare the results of logistic regression and decision tree classifier**

```
In [74]: # Lets first seperate the features and the target.
x = final_df.drop(['Delay'], axis=1)
y = final_df["Delay"]
```

```
In [75]: from sklearn import preprocessing
scaler = preprocessing.MinMaxScaler()
x = scaler.fit_transform(x)
```

```
In [76]: # First Split the data into the training and testing set before performing the further operation.
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, train_size=0.9, random_state=10)
```

### Logistic Regression

```
In [77]: # Lets apply the logistic regression with the randomsearchcv hypermeter tuning.
from sklearn.linear_model import LogisticRegression
lr = LogisticRegression()
```

```
In [78]: from sklearn.model_selection import RandomizedSearchCV
```

```
In [79]: 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

```
Out[79]: RandomizedSearchCV(cv=5, estimator=LogisticRegression(),
                          param_distributions={'penalty': ['l1', 'l2'],
                                              'solver': ['newton-cg', 'liblinear']}),
                          scoring='accuracy', verbose=1)
```

```
In [80]: print(rscv.best_params_)
print(rscv.best_score_)

{'solver': 'newton-cg', 'penalty': 'l2'}
0.5950940921237212
```

```
In [81]: lr = LogisticRegression(penalty='l2', solver='newton-cg')
lr.fit(x_train,y_train).score(x_train,y_train)
```

```
Out[81]: 0.5951215346089206
```

```
In [82]: lr.score(x_test, y_test)
```

```
Out[82]: 0.595640715932799
```

### Decision Tree Classifier

```
In [83]: 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

```
Out[83]: 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)
```

```
In [84]: print(rscv.best_params_)
         print(rscv.best_score_)
```

```
{'min_samples_leaf': 7, 'max_depth': 9, 'criterion': 'gini'}
0.6491830132746778
```

```
In [85]: dtc = DecisionTreeClassifier(max_depth= 9, criterion= 'gini',min_samples_leaf= 7)
         dtc.fit(x_train, y_train).score(x_train, y_train)
```

```
Out[85]: 0.6556280482783232
```

```
In [86]: dtc.score(x_test, y_test)
```

```
Out[86]: 0.651065114746898
```

decision tree has good accuracy.

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

```
In [87]: 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)
```

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

```

Out[87]: RandomizedSearchCV(cv=3,
                        estimator=XGBClassifier(base_score=None, booster=None,
                                                callbacks=None,
                                                colsample_bylevel=None,
                                                colsample_bynode=None,
                                                colsample_bytree=None,
                                                early_stopping_rounds=None,
                                                enable_categorical=False,
                                                eval_metric=None, gamma=None,
                                                gpu_id=None, grow_policy=None,
                                                importance_type=None,
                                                interaction_constraints=None,
                                                learning_rate=None, max_bin=None, ...
                                                min_child_weight=None, missing=nan,
                                                monotone_constraints=None,
                                                n_estimators=100, n_jobs=None,
                                                num_parallel_tree=None,
                                                predictor=None, random_state=None,
                                                reg_alpha=None, reg_lambda=None, ...),
                        n_iter=50,
                        param_distributions={'colsample_bytree': [0.6, 0.7, 0.8, 0.9,
                                                                    1],
                                             'learning_rate': [0.4, 0.45, 0.5, 0.55,
                                                             0.6],
                                             'max_depth': range(6, 10),
                                             'n_estimators': range(8, 20)},
                        scoring='accuracy', verbose=1)

```

```

In [88]: print("Best parameters found: ", xgb_random.best_params_)
         print("Best accuracy found: ", xgb_random.best_score_)

Best parameters found: {'n_estimators': 15, 'max_depth': 9, 'learning_rate': 0.4, 'colsample_bytree': 0.8}
Best accuracy found: 0.6622805010843407

```

```

In [89]: xgb = XGBClassifier(n_estimators=15, max_depth=9, learning_rate=0.4, colsample_bytree=0.8)
         xgb.fit(x_train,y_train).score(x_train,y_train)

```

```

Out[89]: 0.6858553785431708

```

```

In [90]: # 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))

```

```

0.595640715932799
0.651065114746898
0.6627868672449764

```

The best Result is from XGBClassifier

```

In [91]: final_df.to_excel("output.xlsx")

```

```

In [ ]:

```