### **DS 600 Project Data Mining on Flights**

```
In [1]:
            import pandas as pd
         2 import numpy as np
In [2]:
         1 import matplotlib as mpl
         2 import matplotlib.pyplot as plt
         3 import seaborn as sns
         4 import scipy.fftpack as fftpack
         5 import warnings
         6 warnings.filterwarnings("ignore")
         1 df = pd.read_csv('airline.csv')
In [3]:
In [4]:
         1 df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 277374 entries, 0 to 277373
        Data columns (total 21 columns):
                               277374 non-null int64
        year
                               277374 non-null int64
        month
        carrier
                               277374 non-null object
        carrier_name
                               277374 non-null object
                               277374 non-null object
        airport
                              277374 non-null object
        airport_name
        arr_flights
                              276999 non-null float64
        arr_del15
                              276941 non-null float64
        carrier_ct
                              276999 non-null float64
        weather_ct
                              276999 non-null float64
        nas ct
                              276999 non-null float64
        security ct
                              276999 non-null float64
        late_aircraft_ct
                              276999 non-null float64
        arr_cancelled
                              276999 non-null float64
        arr diverted
                              276999 non-null float64
        arr_delay
                               276999 non-null float64
        carrier_delay
                               276999 non-null float64
                               276999 non-null float64
        weather_delay
        nas_delay
                               276999 non-null float64
        security_delay
                               276999 non-null float64
        late_aircraft_delay
                               276999 non-null float64
        dtypes: float64(15), int64(2), object(4)
        memory usage: 44.4+ MB
```

1 df.head() In [5]:

Out[5]:

	year	month	carrier	carrier_name	airport	airport_name	arr_flights	arr_del15	carrier_ct	weather_ct	 security_ct	late_aircraft_ct	arr_cancell
	<b>0</b> 2004	1	DL	Delta Air Lines Inc.	PBI	West Palm Beach/Palm Beach, FL: Palm Beach Int	650.0	126.0	21.06	6.44	 1.0	45.92	۷
	<b>1</b> 2004	1	DL	Delta Air Lines Inc.	PDX	Portland, OR: Portland International	314.0	61.0	14.09	2.61	 0.0	10.05	30
i	<b>2</b> 2004	1	DL	Delta Air Lines Inc.	PHL	Philadelphia, PA: Philadelphia International	513.0	97.0	27.60	0.42	 0.0	17.12	15
;	<b>3</b> 2004	1	DL	Delta Air Lines Inc.	PHX	Phoenix, AZ: Phoenix Sky Harbor International	334.0	78.0	20.14	2.02	 0.0	16.45	3
,	<b>4</b> 2004	1	DL	Delta Air Lines Inc.	PIT	Pittsburgh, PA: Pittsburgh International	217.0	47.0	8.08	0.44	 0.0	16.59	2

5 rows × 21 columns

```
In [6]:
          1 df.isnull().sum()
Out[6]: year
                                     0
                                     0
         month
                                     0
         carrier
         carrier_name
                                     0
                                     0
         airport
         airport_name
                                     0
         arr_flights
                                   375
         arr_del15
                                   433
                                   375
         carrier_ct
         weather_ct
                                   375
         nas_ct
                                   375
                                   375
         security_ct
                                   375
         late_aircraft_ct
                                   375
         arr_cancelled
                                   375
         arr_diverted
         arr_delay
                                   375
         carrier_delay
                                   375
                                   375
         weather_delay
         nas_delay
                                   375
         security_delay
                                   375
         late_aircraft_delay
                                   375
         dtype: int64
In [7]:
             df= df.dropna()
In [8]:
           1 df.isnull().sum()
Out[8]: year
                                   0
         month
                                   0
         carrier
                                   0
                                   0
         carrier_name
         airport
                                   0
                                   0
         airport_name
         arr_flights
                                   0
         arr_del15
                                   0
         carrier ct
                                   0
                                   0
         weather_ct
                                   0
         nas_ct
         security_ct
                                   0
         late_aircraft_ct
         arr_cancelled
                                   0
         arr_diverted
         arr_delay
                                   0
         carrier_delay
                                   0
         weather_delay
                                   0
         nas_delay
                                   0
                                   0
         security_delay
         late_aircraft_delay
         dtype: int64
          1 df.nunique()
In [9]:
Out[9]: year
                                      18
                                      12
         month
         carrier
                                      28
         carrier_name
                                      32
                                     406
         airport
         airport_name
                                     406
         arr_flights
                                    7132
                                    2300
         arr_del15
                                   17520
         carrier_ct
         weather_ct
                                    5461
         nas_ct
                                   20456
         security_ct
                                    1098
         late_aircraft_ct
                                   21564
        arr_cancelled 565
arr_diverted 137
arr_delay 26798
carrier_delay 13519
weather_delay 5149
nas_delay 13755
security_delay 629
         arr_cancelled
                                     565
                                   17078
         late_aircraft_delay
         dtype: int64
```

```
In [10]: 1 df.describe()
```

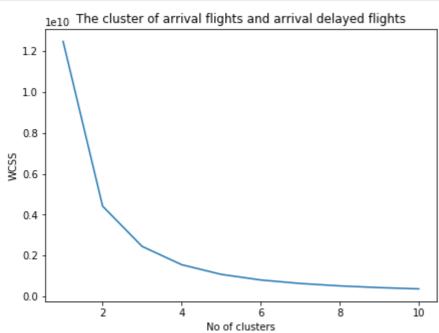
Out[10]:

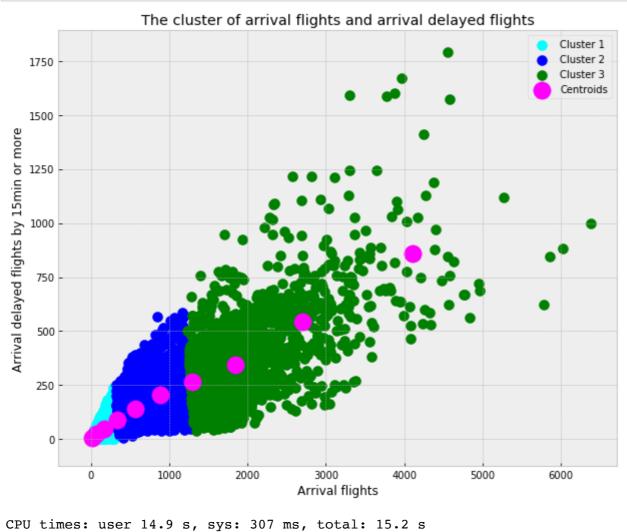
```
security_ct late_aircraft_ct
                                         arr_flights
                                                         arr_del15
                                                                                      weather_ct
                             month
                                                                        carrier ct
                                                                                                         nas ct
                year
count 276941.000000 276941.000000
                                                                   276941.000000 276941.000000 276941.000000 276941.000000 27
                                     276941.000000 276941.000000
         2011.190272
                           6.521432
                                        396.809775
                                                        77.560744
                                                                        21.758212
                                                                                        2.756241
                                                                                                      26.010714
                                                                                                                       0.178455
                                                                                                                                     26.857183
mean
            4.953442
                           3.466662
                                                                        47.700162
                                                                                                      89.750606
  std
                                       1055.373010
                                                       206.911642
                                                                                       10.367470
                                                                                                                      0.839823
                                                                                                                                     78.527039
         2003.000000
                           1.000000
                                          1.000000
                                                          0.000000
                                                                         0.000000
                                                                                        0.000000
                                                                                                       -0.010000
                                                                                                                       0.000000
                                                                                                                                      0.000000
 min
                           3.000000
                                                                         3.570000
                                                                                        0.000000
                                                                                                                      0.000000
 25%
         2007.000000
                                         61.000000
                                                         10.000000
                                                                                                       2.050000
                                                                                                                                      2.000000
                           7.000000
         2011.000000
                                        125.000000
                                                        25.000000
                                                                         8.970000
                                                                                        0.680000
                                                                                                       6.250000
                                                                                                                       0.000000
                                                                                                                                      6.730000
 50%
         2016.000000
                                        285.000000
                                                        60.000000
                                                                        20.630000
                                                                                        2.170000
                                                                                                                                     18.600000
 75%
                          10.000000
                                                                                                      16.780000
                                                                                                                      0.000000
                                      21977.000000
        2020.000000
                          12 000000
                                                                                      717.940000
                                                                                                                                   1885.470000
                                                      6377.000000
                                                                     1792.070000
                                                                                                    4091.270000
                                                                                                                     80.560000
 max
```

```
In [11]:
          1 df.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 276941 entries, 0 to 277373
         Data columns (total 21 columns):
         year
                                276941 non-null int64
                                276941 non-null int64
         month
         carrier
                                276941 non-null object
         carrier_name
                                276941 non-null object
                                276941 non-null object
         airport
         airport_name
                                276941 non-null object
         arr_flights
                                276941 non-null float64
         arr_del15
                                276941 non-null float64
         carrier_ct
                                276941 non-null float64
         weather_ct
                                276941 non-null float64
         nas_ct
                                276941 non-null float64
                                276941 non-null float64
         security_ct
         late_aircraft_ct
                                276941 non-null float64
                                276941 non-null float64
         arr_cancelled
         arr_diverted
                                276941 non-null float64
         arr delay
                                276941 non-null float64
         carrier_delay
                                276941 non-null float64
         weather_delay
                                276941 non-null float64
         nas_delay
                                276941 non-null float64
         security_delay
                                276941 non-null float64
         late_aircraft_delay
                                276941 non-null float64
         dtypes: float64(15), int64(2), object(4)
         memory usage: 46.5+ MB
```

#### **Clusters by Kmeans**

```
In [12]:
             #selected arrival flights and arrival delayed flights
          2
            x = df.iloc[:, [7,8]].values
          3 x
Out[12]: array([[126.
                [ 61. , 14.09],
                [ 97. ,
                          27.6],
                [ 26.
                           4.67],
                [ 35.
                           6.83],
                           1.16]])
In [13]:
          1 from sklearn.cluster import KMeans
             wcss = []
In [14]:
          1 %%time
          3 for i in range(1,11):
          4
                 kmeans = KMeans(n_clusters = i, init = 'k-means++', random_state = 0)
                 kmeans.fit(x)
                 wcss.append(kmeans.inertia )
         CPU times: user 1min 43s, sys: 12.5 s, total: 1min 55s
         Wall time: 57 s
```





CLUSTER 1: Low flights, Lowest delayed CLUSTER and centroid distance is near to each other.

2 : Average arrivals, Average delayed CLUSTER

Wall time: 14.8 s

3: High arrival, High delayed CLUSTER and centroid point distance has been increased over a period

Our target would be cluster 3 wherein carrier companies need to improve their performance for further not getting delayed

### **Exploratory Data Analysis**

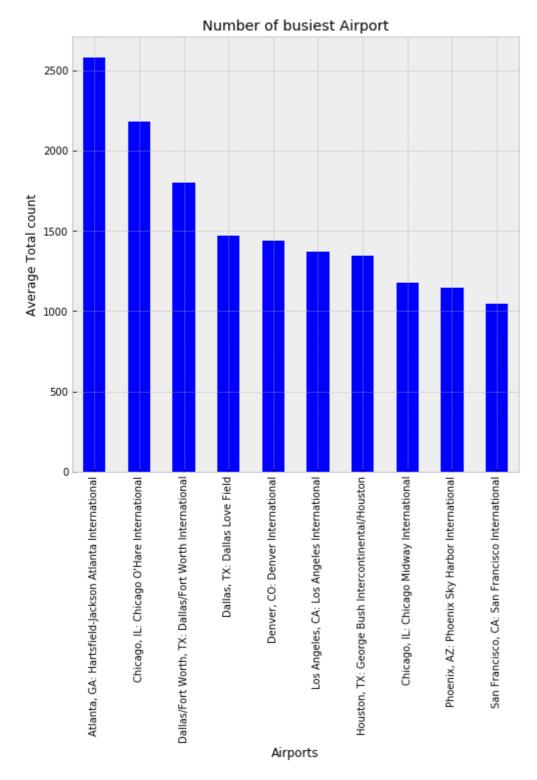
# Which is busiest airport in United States?

Out[20]:

	airport_name	year	month	arr_flights	arr_del15	carrier_ct	weather_ct	nas_ct	security_ct	late_aircraft_ct	arr_cancelled	arr_divert
0	Aberdeen, SD: Aberdeen Regional	2015.551020	6.602041	60.642857	6.285714	2.257449	0.360612	1.765102	0.000000	1.902959	0.979592	0.1734
1	Abilene, TX: Abilene Regional	2010.017544	6.425439	169.258772	36.061404	14.716711	3.961272	5.977544	0.056798	11.348640	3.885965	0.1622
2	Adak Island, AK: Adak	2011.288557	6.537313	8.641791	1.606965	0.509303	0.077413	0.538358	0.016965	0.465025	0.557214	0.0149
3	Aguadilla, PR: Rafael Hernandez	2012.171429	6.534066	52.789011	13.848352	5.371648	0.390725	3.665736	0.072769	4.347560	0.764835	0.0747
4	Akron, OH: Akron-Canton Regional	2011.328330	6.582552	124.475610	26.941839	8.961492	1.359193	5.811135	0.021135	10.788959	2.424953	0.2326

```
In [21]: plt.figure()
    plt.figure(figsize=(8,8))
    dfairpot.groupby("airport_name").arr_flights.mean().sort_values(ascending=False)[:10].plot(kind = 'bar',colors
    plt.title('Number of busiest Airport')
    plt.xlabel('Airports')
    plt.ylabel('Average Total count')
    plt.savefig('Graphl.png',dpi=300, bbox_inches='tight')
    plt.show()
```

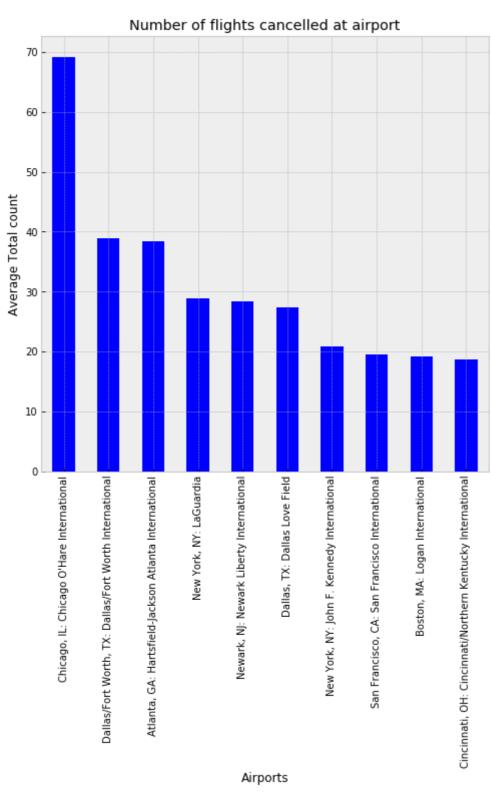
<Figure size 432x288 with 0 Axes>



Atlanta is the busiest airport amongst top 10 airport in United States.

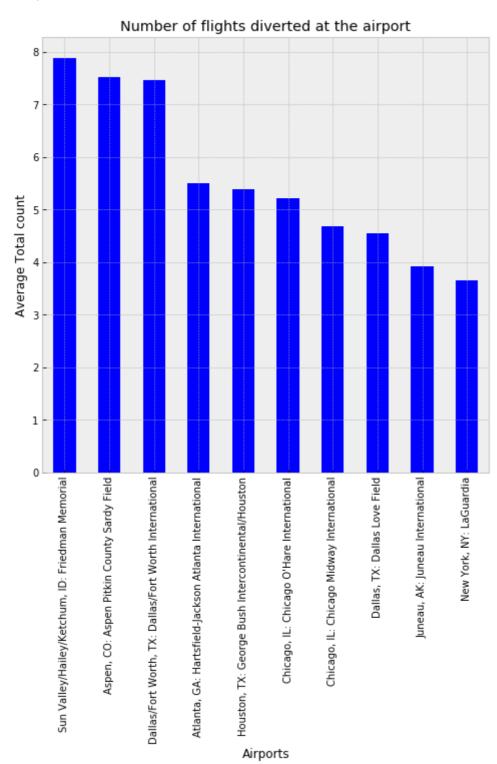
# At Which airport most flight has been cancelled?

<Figure size 432x288 with 0 Axes>



At Chicago O'Hare international airport has been most flights cancelled.

# At which airport most flights has been diverted?



At Sun Valley, Aspen and Dallas fort worth inetrnational airport most flights has been diverted

#### **Problem statement**

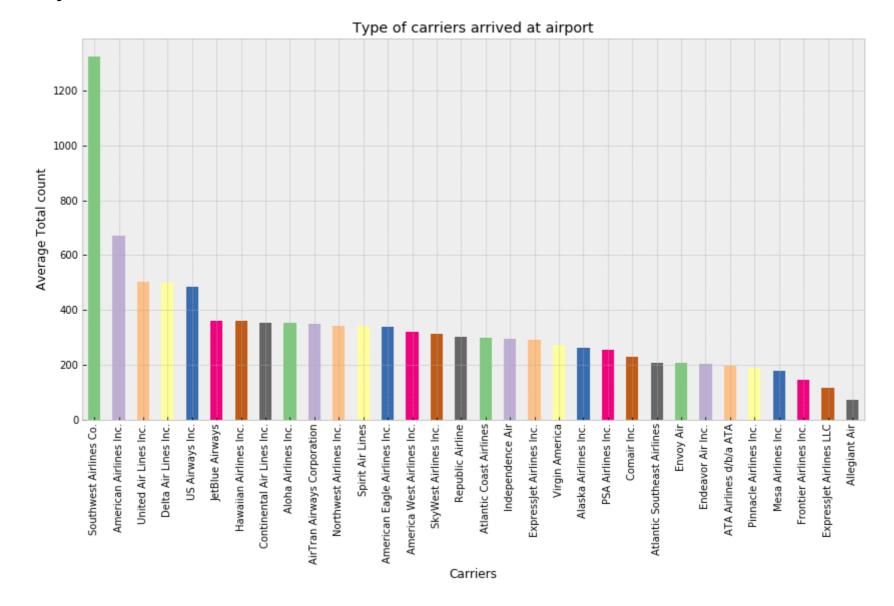
The study focus on which are the best airlines has most reliable carriers.

Out[19]:

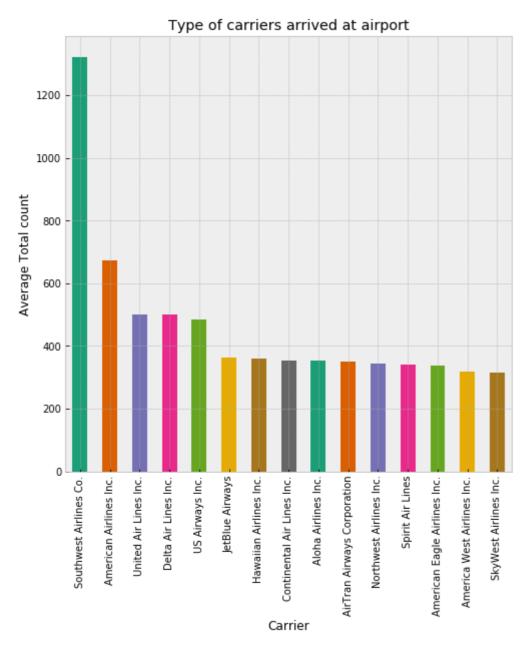
	carrier_name	year	month	arr_flights	arr_del15	carrier_ct	weather_ct	nas_ct	security_ct	late_aircraft_ct	arr_cancelled	arr_dive
0	ATA Airlines d/b/a ATA	2004.399782	6.875817	197.672113	38.450980	7.149967	0.262952	19.800000	0.366187	10.871874	2.028322	0.087
1	AirTran Airways Corporation	2008.772394	6.577139	349.067146	69.921927	13.210239	0.622456	26.814306	0.000000	29.274893	3.417829	0.801
2	Alaska Airlines Inc.	2012.076741	6.549112	260.460264	44.479199	12.357491	0.817403	16.541396	0.320614	14.442357	2.698134	0.716
3	Allegiant Air	2018.621709	6.107897	70.852779	15.079298	4.191355	0.909734	3.671371	0.072099	6.234972	0.488463	0.177
4	Aloha Airlines Inc.	2006.695652	6.652174	353.940711	25.193676	12.877391	0.160316	2.296364	0.211304	9.648419	4.644269	0.142

### Which carrier is most arrived at airports?

<Figure size 432x288 with 0 Axes>



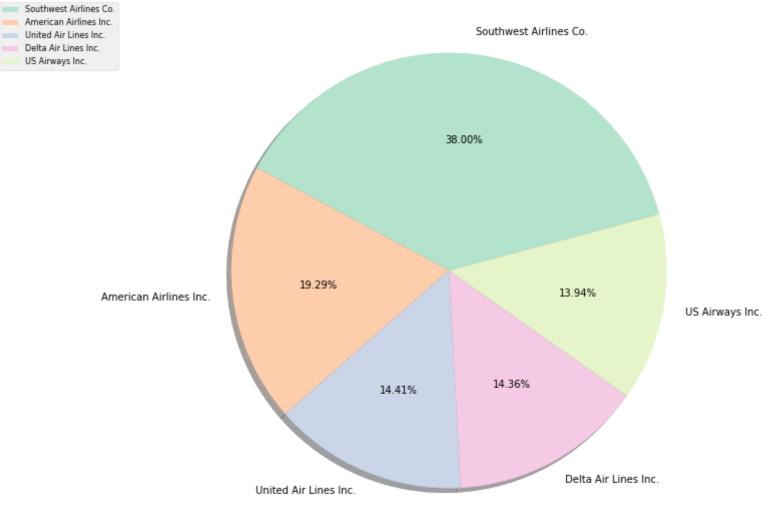
<Figure size 432x288 with 0 Axes>



```
In [35]: arr delay: Total time (minutes) of delayed flights.
          2
         lt3figure()
         lt4figure(figsize=(10,10))
         fcarrier.groupby("carrier_name").arr_flights.mean().sort_values(ascending=False)[:5].plot(kind='pie',
                                                                                                  autopct='%1.2f%%',
          7
                                                                                                     startangle=15,
          8
                                                                                                     shadow = True,
                                                                                                    colors=plt.cm.Pastel2.
         ltOtitle('Percent of arrivals delyed carriers')
         lt1xlabel('Carrier')
         lt2legend(bbox_to_anchor=(-0.1, 1.),fontsize=8,loc="best")
         lt3ylabel("")
         lt4savefig('piechart1.png', bbox_inches='tight')
         15
         lt6show()
```

<Figure size 432x288 with 0 Axes>

Percent of arrivals delyed carriers



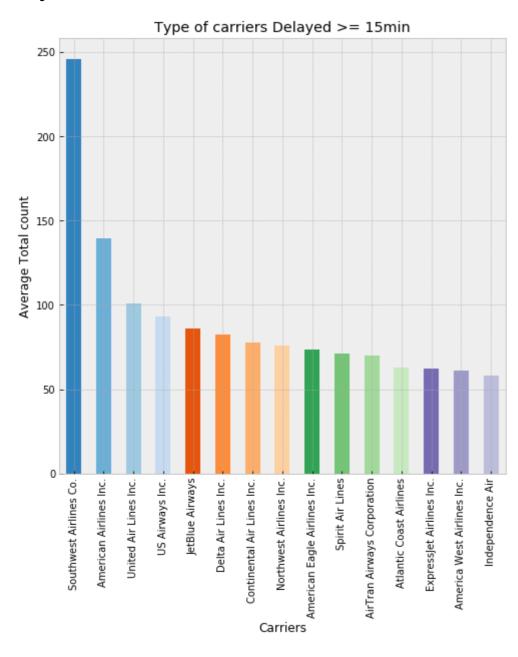
Carrier

Out of 32 carriers I have selected top 15 busiest carriers for Analysis.

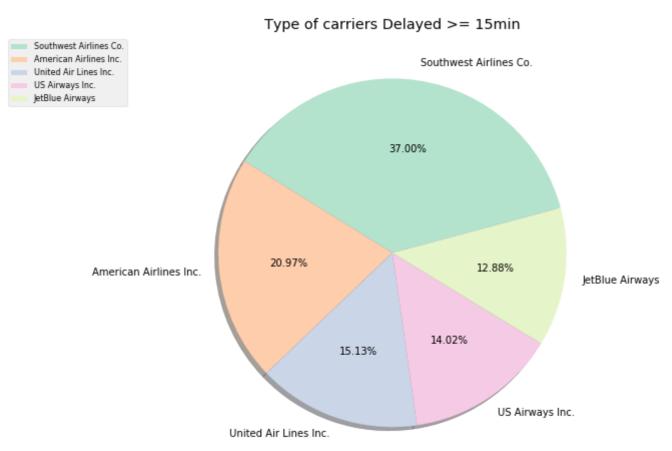
**Southwest Airline comapny** has more number of arrivals comapared to its competitors.

# Which carriers are most delayed at airport?

<Figure size 432x288 with 0 Axes>



```
In [37]:
          1 plt.figure()
          2 plt.figure(figsize=(8,8))
          3 cmap = plt.get_cmap('Spectral')
          4 colors = [cmap(i) for i in np.linspace(0, 1, 8)]
             dfcarrier.groupby("carrier_name").arr_del15.mean().sort_values(ascending=False)[:5].plot(kind='pie',
                                                                                                      autopct='%1.2f%%',
                                                                                                         startangle=15,
          7
          8
                                                                                                         shadow = True,
          9
                                                                                                        colors=plt.cm.Pas
         10 plt.title('Type of carriers Delayed >= 15min')
         11 plt.xlabel('Carriers')
         12 plt.legend(bbox_to_anchor=(-0.1, 1.),fontsize=8,loc="best")
         13 plt.ylabel("")
         plt.savefig('piechart2.png', bbox_inches='tight')
         15
         16 plt.show()
         17
```

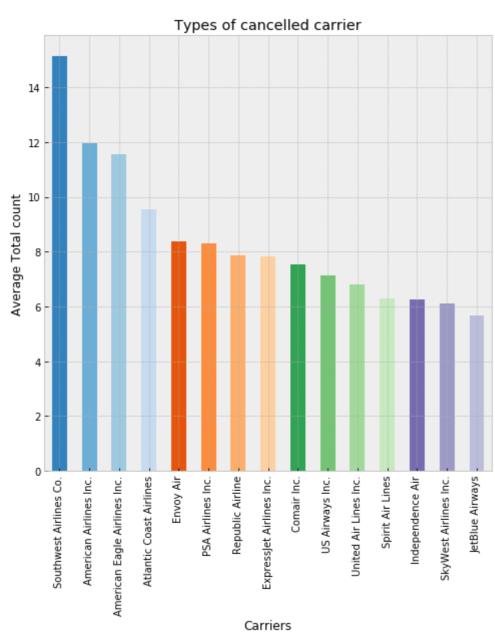


In this part of analysis I observed southwest airlines has most number of delayed compared to its competitors because they have most number of flyings and arrivals count. Whereas only flights are listed above delayed by greater than equal to 15 minutes.

Carriers

### Which carrier has been most cancelled at airports?

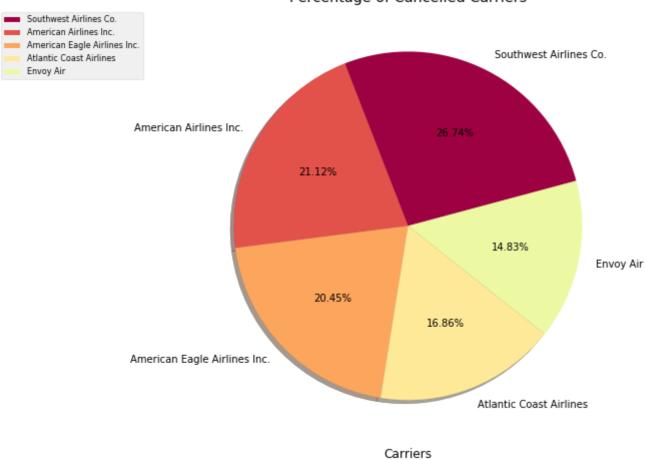
<Figure size 432x288 with 0 Axes>



```
In [34]:
          1 plt.figure()
          2 plt.figure(figsize=(8,8))
          3 cmap = plt.get_cmap('Spectral')
          4 colors = [cmap(i) for i in np.linspace(0, 1, 8)]
             dfcarrier.groupby("carrier_name").arr_cancelled.mean().sort_values(ascending=False)[:5].plot(kind='pie',
                                                                                                      autopct='%1.2f%%',
          7
                                                                                                           startangle=15
          8
                                                                                                           shadow = True
          9
                                                                                                           colors=colors)
         10 plt.title('Percentage of Cancelled Carriers')
         11 plt.xlabel('Carriers')
         12 plt.legend(bbox_to_anchor=(-0.1, 1.),fontsize=8,loc="best")
         13 plt.ylabel("")
         14 plt.savefig('piechart3.png', bbox_inches='tight')
         15
         16 plt.show()
```

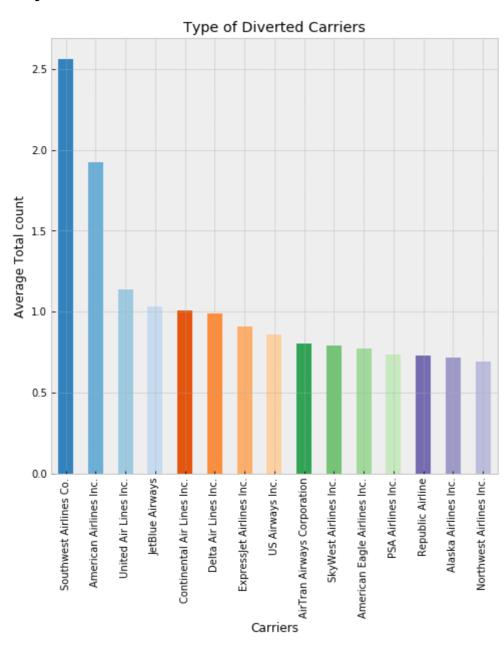
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# Percentage of Cancelled Carriers

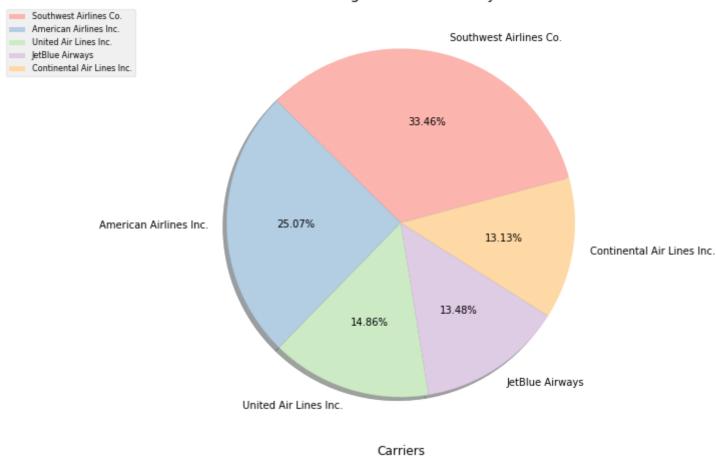


Which carrier has been divered at the airport?

<Figure size 432x288 with 0 Axes>



#### Percentage of diverted delayed



So far we have seen arrival count, delayed count, cancelled count and diverted count in which we can see usual trend for top 3 carrier those are not reliable, whereas trend form Jet Blue to Spirit Airlines are steady.

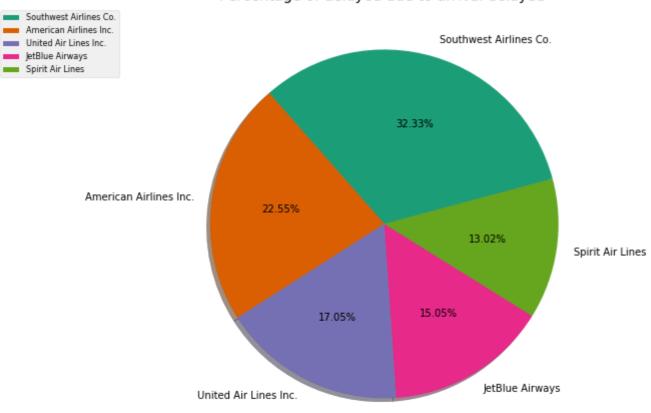
### What are the most delayed reasons for the carrier?

```
In [25]:
                dfdelayed = df[['carrier','carrier_name','arr_delay','carrier_delay',
                                    'weather_delay','nas_delay','security_delay','late_aircraft_delay']]
                dfdelayed= dfdelayed.groupby(['carrier_name']).mean().reset_index()
             4 dfdelayed.head()
Out[25]:
                          carrier_name
                                         arr_delay carrier_delay weather_delay
                                                                               nas_delay security_delay late_aircraft_delay
            0
                   ATA Airlines d/b/a ATA 2225.654684
                                                    441.725490
                                                                   25.202614
                                                                               949.856209
                                                                                             14.063181
                                                                                                             794.807190
              AirTran Airways Corporation 3926.668871
                                                    623.882962
                                                                    48.281524
                                                                             1267.291445
                                                                                              0.000000
                                                                                                             1987.212940
                      Alaska Airlines Inc. 2197.137733
                                                                               582.223305
            2
                                                    712.314884
                                                                   59.492035
                                                                                             11.258352
                                                                                                             831.849158
                           Allegiant Air 1080.610010
            3
                                                    394.692558
                                                                   75.208970
                                                                               160.726357
                                                                                              2.650959
                                                                                                              447.331167
                                                                                62.976285
                                                                                              7.395257
            4
                       Aloha Airlines Inc. 1037.237154
                                                    572.814229
                                                                    9.079051
                                                                                                             384.972332
In [26]:
                import matplotlib.pyplot as plt
```

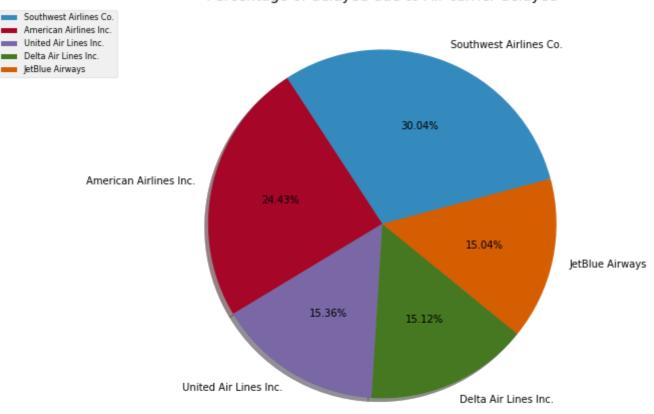
```
In [27]:
             #arr_delay: Total time (minutes) of delayed flights.
          3 plt.figure()
          4 plt.figure(figsize=(10,8))
          5 | dfdelayed.groupby("carrier_name").arr_delay.mean().sort_values(ascending=False)[:5].plot(kind='pie',
                                                                                                      autopct='%1.2f%%',
          7
                                                                                                       startangle=15,
          8
                                                                                                       shadow = True,
          9
                                                                                                      colors=plt.cm.Dark
         10 plt.title('Percentage of delayed due to arrival delayed ')
         11 plt.xlabel('Carriers')
         plt.legend(bbox_to_anchor=(-0.1, 1.),fontsize=8,loc="best")
         13 plt.ylabel("")
         14 plt.savefig('piechart5.png', bbox_inches='tight')
         15 plt.show()
         16
```

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#### Percentage of delayed due to arrival delayed



#### Percentage of delayed due to Air carrier delayed

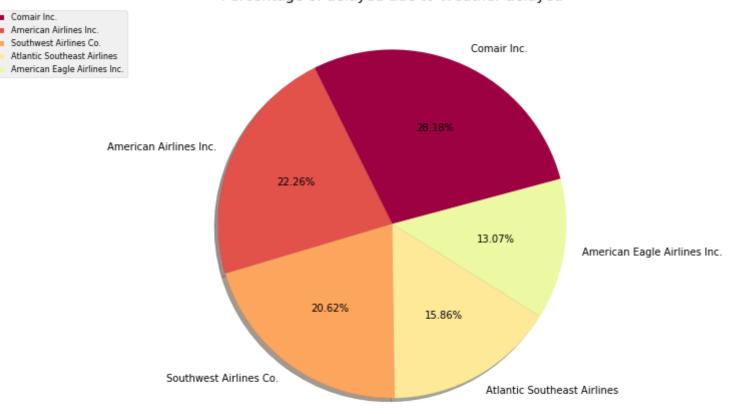


```
In [29]:
          1 #weather delay: Total time (minutes) of delayed flights due to weather.
          2 plt.figure()
          3 plt.figure(figsize=(10,8))
          4 cmap = plt.get_cmap('Spectral')
          5 colors = [cmap(i) for i in np.linspace(0, 1, 8)]
             dfdelayed.groupby("carrier_name").weather_delay.mean().sort_values(ascending=False)[:5].plot(kind='pie',
                                                                                                     autopct='%1.2f%%',
          8
                                                                                                          startangle=15
          9
                                                                                                          shadow = True
         10
                                                                                                         colors=colors)
         plt.title('Percentage of delayed due to Weather delayed')
         12 plt.xlabel('Carriers')
         plt.legend(bbox_to_anchor=(-0.1, 1.),fontsize=8,loc="best")
         14 plt.ylabel("")
         plt.savefig('piechart7.png', bbox_inches='tight')
         16 plt.show()
```

<Figure size 432x288 with 0 Axes>

Comair Inc.

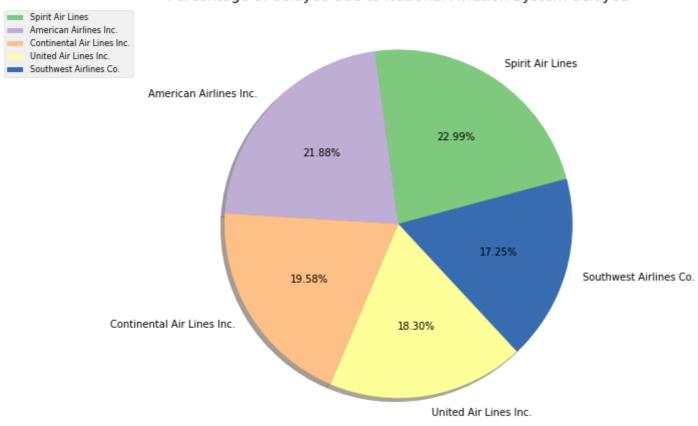
#### Percentage of delayed due to Weather delayed



```
In [30]:
          #nas_delay: Total time (minutes) of delayed flights due to National Aviation System.
          plt.figure()
          plt.figure(figsize=(10,8))
          dfdelayed.groupby("carrier_name").nas_delay.mean().sort_values(ascending=False)[:5].plot(kind='pie',
                                                                                                     autopct='%1.2f%%',
           6
                                                                                                      startangle=15,
           7
                                                                                                      shadow = True,
           8
                                                                                                     colors=plt.cm.Accent.
          plt.title('Percentage of delayed due to National Aviation System delayed')
          1plt.xlabel('Carriers')
          iplt.legend(bbox_to_anchor=(-0.1, 1.),fontsize=8,loc="best")
          1plt.ylabel("")
          lplt.savefig('piechart8.png', bbox_inches='tight')
          lplt.show()
```

<Figure size 432x288 with 0 Axes>

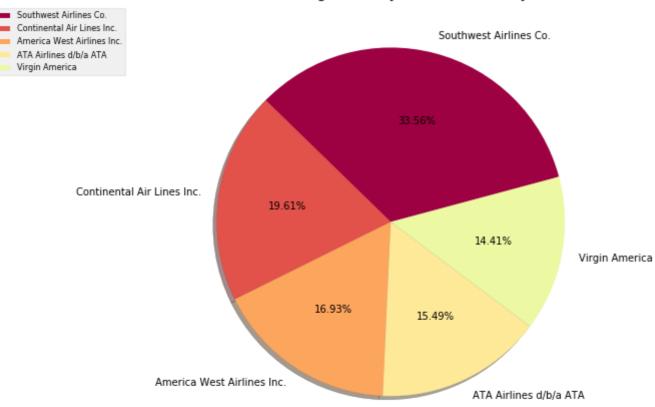
#### Percentage of delayed due to National Aviation System delayed



```
In [31]:
          1 #security_delay: Total time (minutes) of delayed flights due to security.
          2 plt.figure()
          3 plt.figure(figsize=(10,8))
          4 cmap = plt.get_cmap('Spectral')
          5 colors = [cmap(i) for i in np.linspace(0, 1, 8)]
             dfdelayed.groupby("carrier_name").security_delay.mean().sort_values(ascending=False)[:5].plot(kind='pie',
                                                                                                     autopct='%1.2f%%',
          8
                                                                                                           startangle=1
          9
                                                                                                           shadow = True
         10
                                                                                                          colors=colors
         11
         12 plt.title('Percentage of delayed due to security')
         13 plt.xlabel('Carriers')
         plt.legend(bbox_to_anchor=(-0.1, 1.),fontsize=8,loc="best")
         15 plt.ylabel("")
         plt.savefig('piechart9.png', bbox_inches='tight')
         17 plt.show()
```

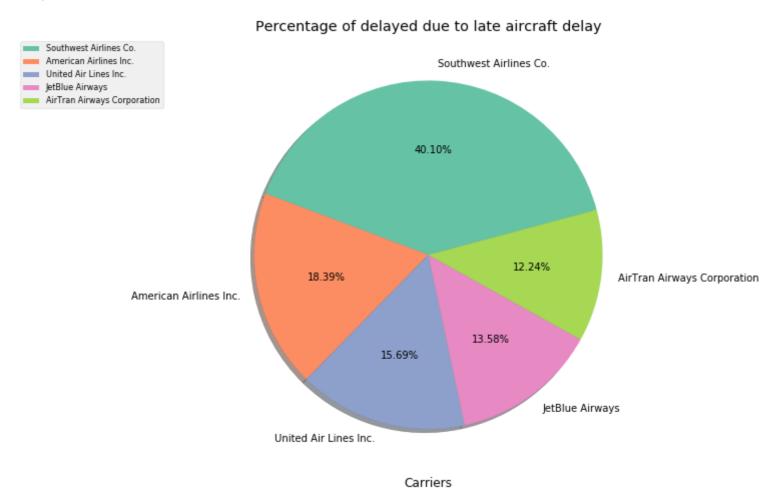
<Figure size 432x288 with 0 Axes>

#### Percentage of delayed due to security



Carriers

```
In [32]:
          1 #late aircraft delay: Total time (minutes) of delayed flights due to a previous flight using the same aircr
          2 plt.figure()
          3 plt.figure(figsize=(10,8))
             dfdelayed.groupby("carrier_name").late_aircraft_delay.mean().sort_values(ascending=False)[:5].plot(kind='pic
                                                                                                      autopct='%1.2f%%',
          6
                                                                                                                 startan
          7
                                                                                                                 shadow
          8
                                                                                                                 colors=
          9 plt.title('Percentage of delayed due to late aircraft delay')
         10 plt.xlabel('Carriers')
         plt.legend(bbox_to_anchor=(-0.1, 1.),fontsize=8,loc="best")
         12 plt.ylabel("")
         13 plt.savefig('piechart10.png', bbox_inches='tight')
         14 plt.show()
```



Mainly there are 6 reasons for flights delayed those are listed and shown above in form of pie charts.

In analysis I observed Southwest airline and American Airline are top 2 airlines with most number of delays.

#### **Conclusion**

Based on the data, Analysis and visualization I concluded that Southwest Airlines and American airlines are not reliable airlines because of most number of delays.

Whereas Jet blue, Spirit airline and other few airlines are reliable to choose for flying based on their arrival count, delayed count, cancelled count, diverted count and their delayed reasons.

# Suggestion

In suggestion for the airlines are getting delayed like **Southwest airline**, **American airline** etc. should improve thier performance based on delayed causes and increase quality of service and thereby gain customer satisfaction.

```
In [ ]: 1
```