Research and Model to understand the US aviation market

US FLIGHT ROUTES

RESEARCH QUESTION

- There are more than 10 million flights within US per year, which makes it a massive industry.
- > 6.5M jobs are directly supported by the air transport sector.
- Accounts for around 5% of the GDP of the US.
- Many companies offering similar services in national routes.

Are there any business opportunities for an Airline company by opening a brand new route within two unconnected US cities? If so, can we predict how many flights per month should we run the route with?

CONSIDERATIONS & HYPOTHESES

- Restricted information availability.
 - We have routes, amount of flights per route, per month, per airline.
 - Merge with economic and demographic data to find out if we can explain the amount of flights between a pair of cities given the information we have.

ΔGDPShare -> ΔFlights

ΔGDPpc -> ΔFlights

ΔPopulation -> ΔFlights

Touristic areas -> △Flights

Summer months-> \(\Delta \) Flights

Positive synergy between tourism and summer months.

MODEL SETUP

- **Dummy variables:**
 - Hub == 1 if the airport is a hub.
 - Away == 1 if in Alaska or Hawaii.
 - Within == 1 within the same state.
 - **Summer == 1 in June, July or August.**
 - Coast == 1 if the state has coast.
 - > SumCoa == Summer * Coast, to control for the potential synergy.
- > State GDP per capita and % of the US GDP Share.
- City population and distance between cities.
- Airport clusters (e.g. [LAX, SAN, SNA, ONT], [SFO, SMF, SJC, OAK], [JFK, EWR, LGA]...)

MODEL RESULTS + PREDICTION

```
OLS Regression Results
                                                                           0.729
Dep. Variable:
                                   R-squared (uncentered):
                           Flights
                                                                          0.729
Model:
                                   Adj. R-squared (uncentered):
Method:
                                                                       1.515e+04
                  Least Squares
                                   F-statistic:
                   Tue, 15 Dec 2020 Prob (F-statistic):
                                                                            0.00
Date:
                                                                      -3.9964e+05
                                   Log-Likelihood:
                          18:11:11
Time:
No. Observations:
                                                                        7.993e+05
                             67613
                                    AIC:
Df Residuals:
                             67601
                                     BIC:
                                                                        7.994e+05
Df Model:
Covariance Type:
                         nonrobust
               coef
                                             P>|t|
                      std err
                                                       [0.025
                                                                  0.975]
                                                                  64.016
Carrier
            63.4309
                        0.298
                                 212.526
                                                       62.846
                                             0.000
Dist
           -38.4432
                        0.662
                                             0.000
                                                      -39.742
                                 -58.028
                                                                 -37.145
Hub
         37.0673
                         0.748
                                49.538
                                             0.000
                                                       35.601
                                                                  38.534
            44.4714
                        2.178
                                20.419
                                             0.000
                                                     40.203
                                                                  48.740
Away
Within
                                7.633
            12.3053
                        1.612
                                             0.000
                                                     9.146
                                                                  15.465
            21.9261
                                 20.812
                                             0.000
                                                                  23.991
Coast
                        1.054
                                                     19.861
             6.8082
                        1.193
                                   5.707
                                             0.000
                                                        4.470
                                                                   9.146
Summer
             6.5405
                                   5.793
                                                        4.327
                                                                   8.754
                        1.129
                                             0.000
GDP
                        0.182
                                70.163
                                                     12.389
            12.7450
                                             0.000
                                                                  13.101
Pop
                                 -18.517
                         0.248
                                             0.000
                                                       -5.077
Share
            -4.5908
                                                                  -4.105
Share2
            0.2125
                         0.008
                                25.180
                                             0.000
                                                     0.196
                                                                   0.229
            -3.0352
                        1.578
                                  -1.923
                                             0.054
                                                       -6.128
                                                                   0.058
SumCoa
                         28121.045
                                   Durbin-Watson:
Omnibus:
                                                                   0.257
Prob(Omnibus):
                             0.000
                                    Jarque-Bera (JB):
                                                              229735.339
                        1.799 Prob(JB): 0.00
Skew:
                                                                1.23e+03
                            11.282 Cond. No.
Kurtosis:
```

RESEARCH QUESTION

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MODEL RESULTS + PREDICTION

We have taken 3 of the routes which are not connected yet.

	Route	Carrier	Dist	Hub	Away	Within	Coast	Summer	GDP	Pop	Share	Share2	SumCoa
0	MIA-PDX	1	2.70	1	0	0	1	0	1.047	1.51	6.27	39.31	0
1	MIA-PDX	1	2.70	1	0	0	1	1	1.047	1.51	6.27	39.31	1
2	BOS-SAT	1	1.76	1	0	0	1	0	1.442	1.05	11.43	130.64	0
3	BOS-SAT	1	1.76	1	0	0	1	1	1.442	1.05	11.43	130.64	1
4	PIT-MCI	1	0.77	0	0	0	0	0	1.151	2.64	5.43	29.49	0
5	PIT-MCI	1	0.77	0	0	0	0	1	1.151	2.64	5.43	29.49	0

model.predict(xp)								
0	24.289652							
1	28.062687							
2	52.866719							
3	56.639753							
4	56.343079							
5	63.151320							

FUTURE IMPROVEMENTS + LEARNING

- Improvement of the model:
 - More reliable information.
 - > Passengers per route to interpret demand; we know #flights but what is the size?
 - Pricing will definitely influence the demand for flights. Not enough data.
 - > We could create various models instead of a model with several dummies.
 - Carriers and Flights can be correlated both ways, we should control for that.
- **Learning:**
 - Tableau Dashboards, Python. Machine Learning and Feature Engineering.
 - Time management and focusing on establishing an idea and providing an MVP.