## **Final case Analysis**

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#### **Business statistics 1**

The final case analysis is based on aviation industry dataset itineraries sold in the first quarters of years 2009, 2017, 2018. The rates of flight tickets vary based on the month of the year for example the rates of tickets are high in December and august. The months selected are consistent for all the years mentioned.

The variables used in this dataset are:

ItniID - unique ID for each transaction

Year – year when flight took place

Origin – starting destination airport

originStateName – state where airport is located

roundtrip- if its roundtrip then 1 or else 0

online – if booked online then 1 or else 0

FarePerMile- charges per mile

RPcarrier – the airline

Passengers

ItinFare

Distance

DistanceGroup

MilesFlown

Region

Division

Pop

## **Data cleaning**

## **Conditions**

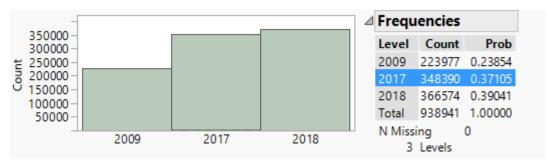
#### **Continuous variables:**

- a) Itinfare less than 9,800
- b) Passengers less than 100
- c) Distance less than 10,000

## **Categorical Variables:**

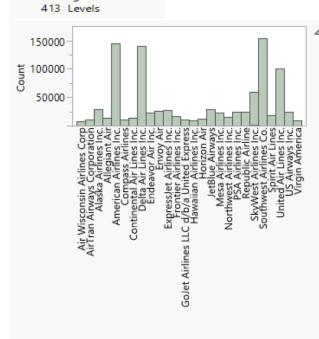
- a) OriginStateNames: Remove "U.S. Pacific Trust Territories and Possessions", "Puerto Rico" and "U.S. Virgin Islands"
- b) RPCarrier: remove all carriers that have less than 5000 itineraries listed in the dataset

The data was cleaned by applying appropriate filters for 3 years combined. After getting the subset we get total number of rows as 938,941. We can observe variance in the overall tickets sold in these 3 years. 2009 shows a lower count since it is the year following the economic depression. 2017 shows growth in the sales of the tickets but the number of tickets being sold in 2018 as compared to 2017 is not impressing as there is only slight growth of 18,184.



The number of flights taking off from LAX airport are 30448 which is highest than any other airport that is 30448. And the lowest are in AIA that is 1.

Frequencies							
Level	Count	Prob					
SLC	12493	0.01331					
BWI	12552	0.01337					
TPA	13027	0.01387					
FLL	13529	0.01441					
IAH	13553	0.01443					
SAN	13631	0.01452					
LAS	13899	0.01480					
PDX	14247	0.01517					
DCA	14354	0.01529					
JFK	15301	0.01630					
PHL	16493	0.01757					
PHX	17008	0.01811					
MCO	17062	0.01817					
DTW	18037	0.01921					
LGA	18624	0.01984					
DFW	19126	0.02037					
MSP	19475	0.02074					
EWR	19751	0.02104					
SFO	21811	0.02323					
ATL	22129	0.02357					
SEA	22397	0.02385					
BOS	22835	0.02432					
DEN	23110	0.02461					
ORD	29118	0.03101					
LAX	30448	0.03243					
Total	938941	1.00000					
N Missing 0							



Frequencies					
Level	Count	Prob			
Air Wisconsin Airlines Corp	6541	0.00697	$^{\wedge}$		
AirTran Airways Corporation	9353	0.00996			
Alaska Airlines Inc.	26900	0.02865			
Allegiant Air	12968	0.01381			
American Airlines Inc.	143529	0.15286			
Compass Airlines	8647	0.00921			
Continental Air Lines Inc.	12091	0.01288			
Delta Air Lines Inc.	139595	0.14867			
Endeavor Air Inc.	21453	0.02285			
Envoy Air	24196	0.02577			
ExpressJet Airlines Inc.	26306	0.02802			
Frontier Airlines Inc.	15065	0.01604			
GoJet Airlines LLC d/b/a United Express	8609	0.00917			
Hawaiian Airlines Inc.	7820	0.00833			
Horizon Air	10813	0.01152			
JetBlue Airways	27510	0.02930			
Mesa Airlines Inc.	21569	0.02297			
Northwest Airlines Inc.	13704	0.01460			
PSA Airlines Inc.	22421	0.02388			
Republic Airline	23278	0.02479			
SkyWest Airlines Inc.	58013	0.06179			
Southwest Airlines Co.	152595	0.16252			
Spirit Air Lines	16456	0.01753			
United Air Lines Inc.	98731	0.10515			
US Airways Inc.	22418	0.02388			
Virgin America	8360	0.00890	~		
N Missing 0 26 Levels		4 00000			

As we can see from the image above the Airlines which most of the people preferred travelling with was southwest airlines Co with count of 15,2595. And the least preferred one was AirWisconsin airline Co with a count of 6541.

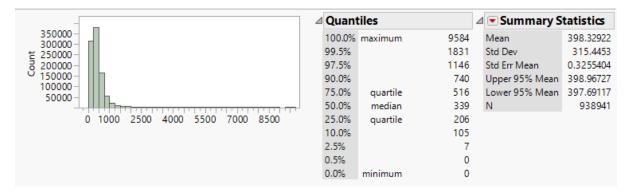
## **Data Analysis**

## **Univariate Analysi**

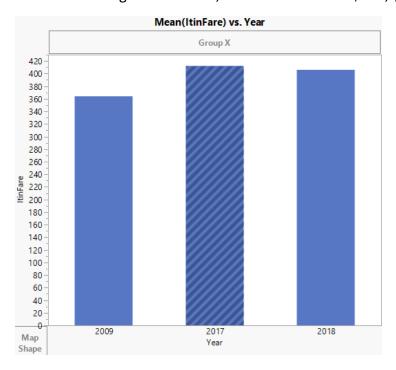
Describe shape and other descriptive statistics of Itinfare, Distance and Passengers variables

## **Itinfare**

As we can see from the image the average fare for all the three years combined with a 95% confidence level is 398.32.

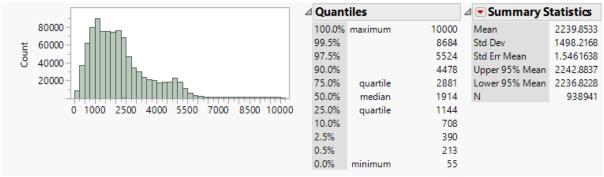


As we can see the average fare has dropped in the year 2018 compared to the previous year 2017. The average fare in 2009, 2017 and 2018 was \$364, \$412 and \$406, respectively

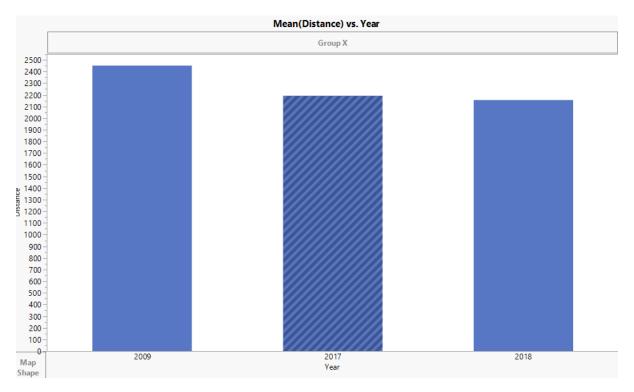


#### Distance

The average distance travelled in 3 years is 2239.8533 miles

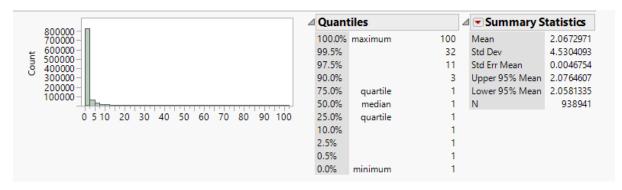


We can see from the graph that there is a decrease in the distance travelled with each year passing. Comparing the three years separately, the average distance travelled in 2009 was 2451 miles while it was 2192 miles in 2017 and 2156 miles in 2018

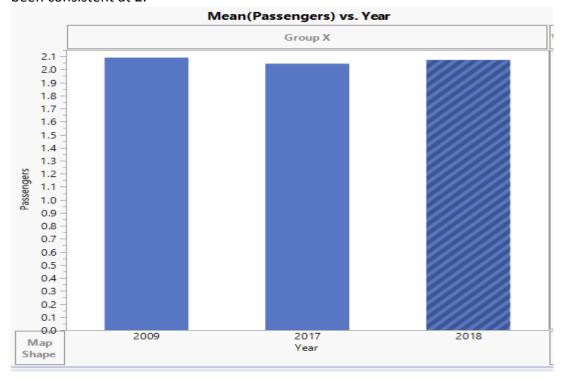


## **Passengers**

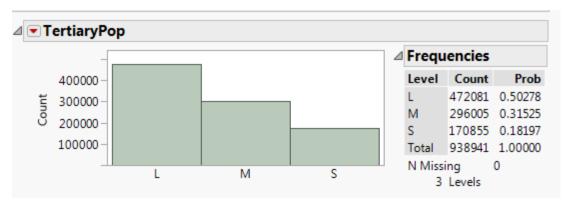
The average number of passengers travelling for 3 years combined is 2.0672971 per itinerary. 90% of the times no of passenger travelling was 3.



There is only slight difference in the number of passengers travelling in these 3 years. It has been consistent at 2.

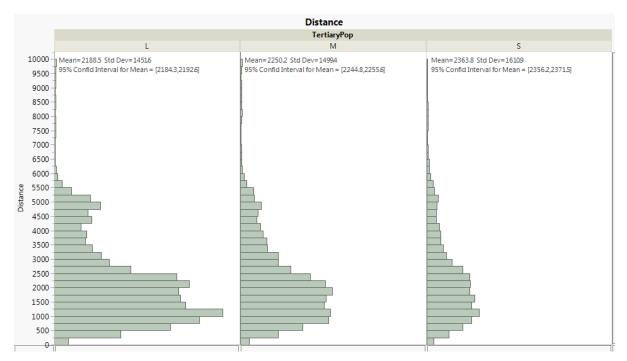


2.Create a new tertiary (three categories) flag variable using population "Pop" variable for states that have populations more than 10 million, 5-10 million and those less than that. Call this variable TertiaryPop (stands for Tertiary population).



A new column was created by sorting states that have population more than 10 million, population between 5-10 million and less than that.

# 3. Run descriptive analysis including confidence intervals on Distance, Itinfare and FareperMile and comment if there are differences in the above by the size of the state



## Distance-

From the above graph we can state that the average distance travelled for large states is 2188.5 miles 95% CI at [2184, 2192]), medium states is 2250.2 miles (CI- [2245, 2256]) and small states is 2363.8 miles Confidence Interval- (2356, 2372). Thus with increase in the area of the state, the average distance travelled has decreased.

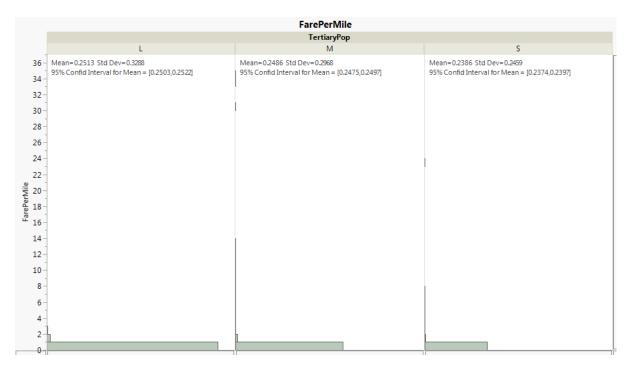
#### **FarePerMile**

The fare per mile has decreased with the size of the state but the difference is very small as we can observe from the graph.

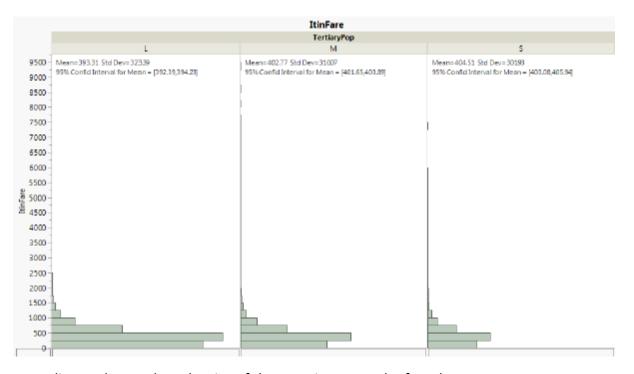
CI for large states- [0.2503, 0.2522]

CI for medium states- [0.2475, 0.2497]

CI for small states- [0.2374, 0.2397]



## **ItinFare**



According to the graph as the size of the state increases the fare decreses.

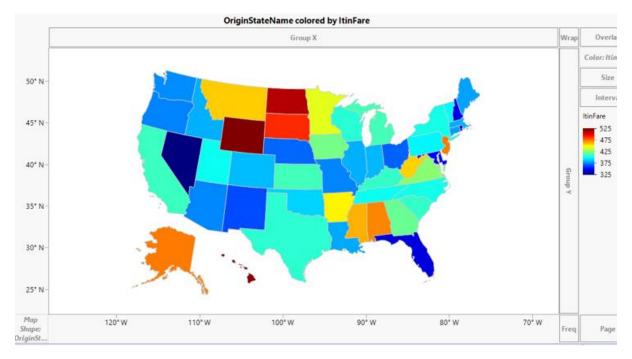
Tickets from large states cost an average of \$393 with a 95% CI of [392, 394]. Medium sized states had their tickets priced at an average of \$403 with CI of [402, 404] and small states averaged at \$405 with CI of [404, 406].

4. It is of interest for consumers to know if there are differences in airline ticket fares by the origin of the flight. In other words is there a difference in ticket fares to fly from different states? Run a comprehensive analysis of comparing average ticket price and average ticket price / mile by state and *visualize* the differences using mapping facility in JMP and comment on the trends.

- a. For all three years combined
- b. Separately for three years

#### itinFare

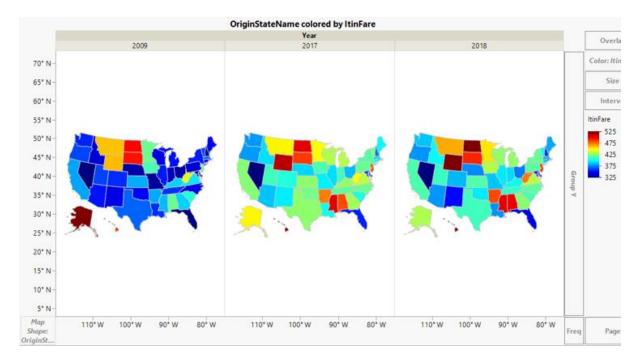
# 1)



Wyoming has the highest ticket fare of 528\$ followed by North Dakota. Nevada has the cheapest fares of 321\$ followed by florida.

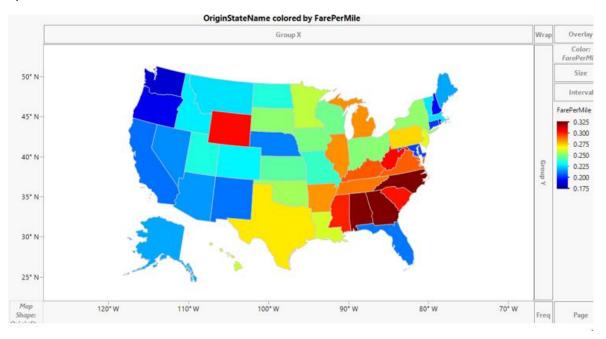
## 2) three years combined

As seen in the graph in 2009, Alaska was the most expensive with a fare of \$573 average while Florida was the cheapest at \$319. In 2017, Hawaii was on top for the highest ticket rates at \$535 while Nevada became the cheapest at \$324. In 2018, Wyoming was the most expensive source of flight with average prices at \$587 as Nevada remained as the cheapest at \$318.



## **FarePerMile**

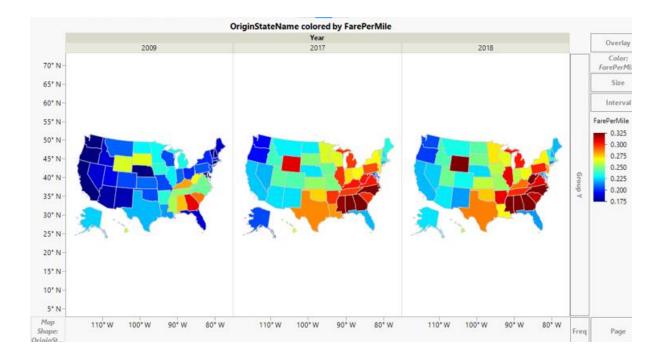
1)



North Carolina had the highest fare per mile at \$0.3296 and Washington State had the lowest at \$0.1870.

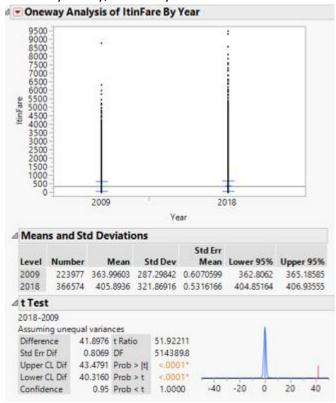
## 2) 3 years combined

As seen in the graph in 2009, New Hampshire had the lowest fare per mile at \$0.1448 and Georgia had the highest at \$0.3090. In 2017, Washington State was the cheapest at \$0.1932 while North Carolina was highest with \$0.3463. In 2018, California was the lowest \$0.2024 and North Carolina remained at the top at \$0.3583.



## **Bivariate Analysis**

a) Have prices changed over time Is there a significant difference in airfare for years 2018 and 2009? If yes why, if not why not?



```
i. H_0: \mu_{2018} = \mu_{2009}

H_1: \mu_{2018} \neq \mu_{2009}

ii. H_0: \mu_{2018} \geq \mu_{2009}

H_1: \mu_{2018} < \mu_{2009}

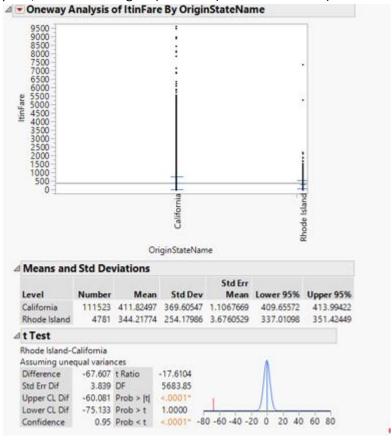
iii. H_0: \mu_{2018} \leq \mu_{2009}

H_1: \mu_{2018} > \mu_{2009}
```

The p value for Prob > |t| is .0001, which is negligible so the first condition that both the averages are equal can be rejected. Similarly the  $2^{nd}$  condition that average of 2018 is greater than the average of 2009 can also be rejected. However, the p-value of 1.0000 is very high for Prob < t, that means the null hypothesis that average of 2018 is lesser than that of 2009, has to be considered.

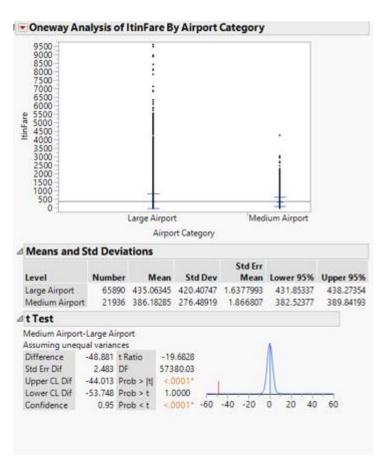
We can state that the average itinerary fare for 2009 was less than 2018. Prices have changed overtime, prices have increased over the years.

**b)** Is it more expensive to fly out of smaller states Test the theory on average ticket prices and price/mile for the largest (California) and the smallest (Rhode Island) states in the nation.



The p-value for Prob > t is 1.0000 the p-value for the hypothesis that the airfare from Rhode Island is less than or equal to that of California has a very high value and cannot be rejected. The null hypotheses that the values are equal and that airfare from Rhode Island is greater than or equal to that of California have both p-values of .0001 which is very less. Here it cannot be proved that it is expensive to fly from smaller states.

Some airports within the same state are more expensive to fly out of- State of California has ~ 30 airports with multiple airports present in larger cities. LAX, SAN, and SFO are three of the largest airports and SNA, SMF and OAK are three medium size airports. Combine the largest and medium size airports in California into two categories and perform a test of hypothesis on the average airfare for the two groups.



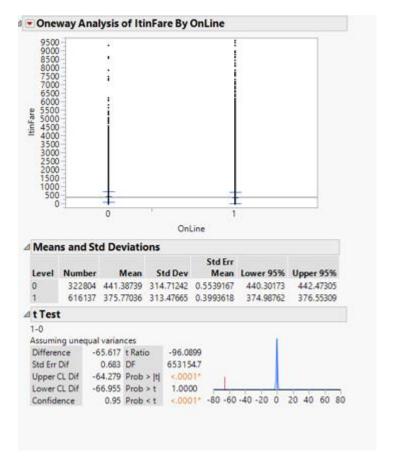
The null hypothesis that airfares from medium airports are less than or equal to airfares from large airports has the high p-value of 1.0000 and cannot be rejected. The null hypotheses that the values are equal and that the airfares from medium airports are greater than or equal to airfares from larger airports can be ignored on the basis of p-value of .0001 in favour of the hypotheses that they are unequal and that airfares from medium airports are cheaper than that of large airports, so it proves that large airports from California are expensive to fly out compared to medium sized airport.

4) Online purchases are cheaper than otherwise Run tests of hypothesis to check if there are significant differences between airfares and miles flown for itineraries that were either purchased online or not.

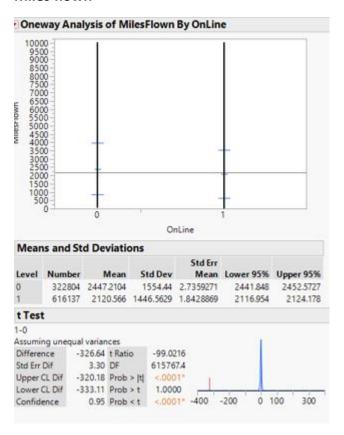
#### AirFare

The null hypothesis that online tickets are cheaper than or equal to tickets purchased offline has value of 1.0000 cannot be rejected. However, hypotheses that online tickets are more expensive or equal to offline tickets can be ignored because of p-value of .0001. This means that the alternate hypothesis that **online tickets are cheaper than offline tickets** is true.

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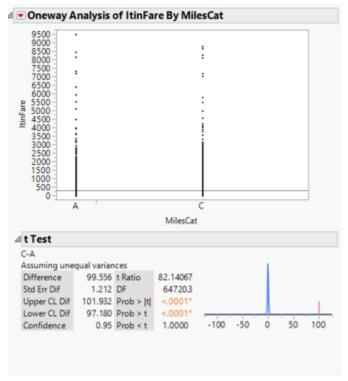


## Miles flown



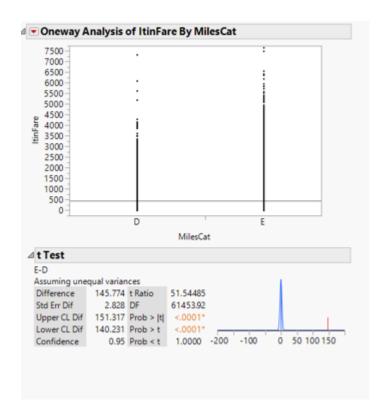
The p-value for the null hypothesis that online tickets have higher or equal miles as that of tickets purchased offline is .0001. So this null hypothesis can be rejected of the hypothesis that online tickets have lower miles flown compared to offline tickets.

- c) The longer the flight, the pricier it is Test this theory by creating a four category variable from MilesFLown variable: "< 500 miles", "500 1200 miles", "1200 2000 miles", "2000 3000 miles", "3000+ miles". Run the following tests of hypothesis:
  - a. Is there a difference in average fare cost between flights that flew "<500 miles" and those that flew "1200-2000 miles"?



The null hypothesis that average airfare for flights that flew less than 500 miles is greater than or equal to those that flew between 1200 and 2000 miles is .0001 and can be rejected. And the hypothesis that that the average airfare for flights that flew between less than 500 miles is less than those that flew between 1200 and 2000 miles has a high p-value and has to be considered.

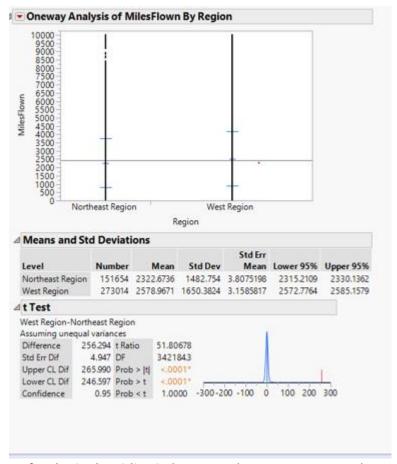
b. Test the hypothesis that there is a difference in average fare cost between flights that flew "2000 – 3000 miles" and "3000+ miles" in the Northeast region.



From the above image we can see that the p-value for the null hypothesis that average fare cost between flights that flew between 2000 and 3000 miles is greater than or equal to that of 3000+ miles in the Northeast region is .0001. Hence, this hypothesis can be rejected and the hypothesis that average fare cost between flights that flew "2000 – 3000 miles" is less than "3000+ miles" in the Northeast region has to be considered.

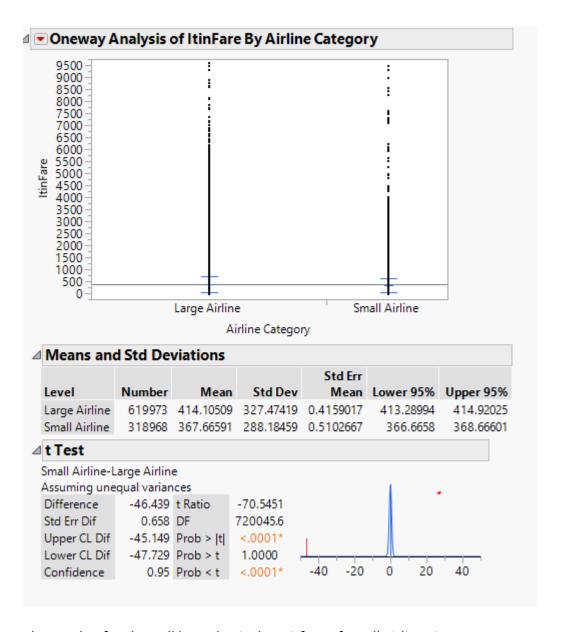
**3.** Run a test of hypothesis to see if there is a significant difference in the miles flown from airports in the Northeast region compared to the Western region.

The t test shows that the p-value for the hypothesis that the miles flown from airports in the Northeast region is higher or equal compared to the Western region is .0001. Hence, it proves that **miles flown** from airports in the Northeast region is less compared to the Western region.



d) Economy of scales in the airline industry Southwest, American, Delta, United, Skywest and JetBlue are six of the largest domestic airline companies. The last question of the case relates to studying the phenomenon of "economy of scales", i.e. are larger airlines on average able to offer more affordable prices for tickets compared to smaller airlines. Create two groups: "Large airlines" and "Small airlines". Large Airlines consists of the six companies mentioned above and the Medium/small group consists of all other airlines that have less than 10k rows of data in the filtered dataset.

Run a test of hypothesis testing if there is a significant difference in the average airfare price for the two groups



The p-value for the null hypothesis that airfare of small airlines is greater than or equal to that of large airlines is .0001. Hence, the null hypothesis can be rejected. And the alternate hypothesis that airfares of small airlines is lesser than airfares of larger airlines is to be considered.