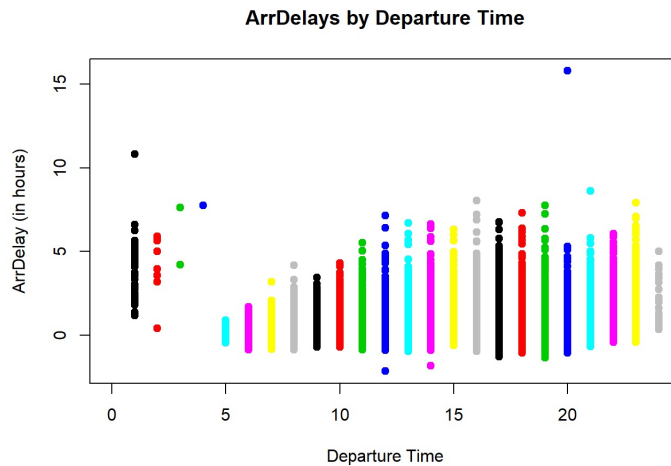


Flights at ABIA

```
library(ggplot2)
par(mfrow=c(2,2))
data=read.csv('C:/Users/rache/OneDrive/Documents/MSBA/Academic/next session/STA380-master/STA380-master/data/ABIA.csv')
data$Austin=ifelse(data$Origin=='AUS',1,0)
```

What is the best time of day to fly to minimize delays?

```
data$hour=round(data$DepTime/100)
plot(data$hour, data$ArrDelay/60, pch=19, col=data$hour, xlab='Departure Time', ylab='ArrDelay (in hours)', main='ArrDelays by Departure Time')
```



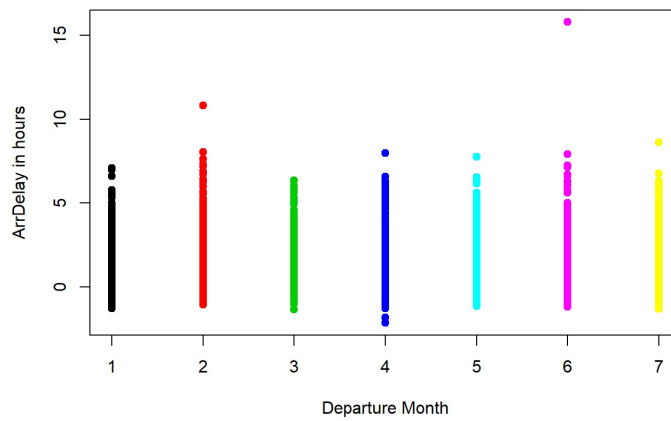
```
median_day=rep(NA,24)
std_day=rep(NA, 24)
for(i in 1:24){
  newsub=subset(data, hour==i)
  median_day[i]<-median(newsub$ArrDelay/60, na.rm=TRUE)
  std_day[i]<-sd(newsub$ArrDelay/60, na.rm=TRUE)
}
choose_day=data.frame(coll=median_day,col2=std_day)
choose_day[
  with(choose_day, order(choose_day$col2),decreasing=FALSE),
]
```

```
##      coll      col2
## 5 -0.01666667 0.1750175
## 6 -0.08333333 0.1908791
## 7 -0.06666667 0.2293273
## 8 -0.06666667 0.2724264
## 9 -0.03333333 0.3566187
## 11 -0.03333333 0.4266867
## 10 -0.03333333 0.4345897
## 12 -0.01666667 0.4677728
## 13 -0.05000000 0.4838855
## 15 -0.03333333 0.5049383
## 14 -0.03333333 0.5275240
## 17 -0.01666667 0.5860976
## 16 0.00000000 0.5875870
## 18 0.03333333 0.6161010
## 19 0.01666667 0.6354975
## 20 0.05000000 0.7367572
## 21 0.16666667 0.8395833
## 22 0.41666667 1.0999831
## 24 1.46666667 1.2421436
## 23 0.26666667 1.4330274
## 1 3.75000000 1.7487295
## 2 5.00000000 1.7973188
## 3 4.21666667 1.9726134
## 4 7.76666667      NA
```

The table above indicates the standard deviation and median of different Arrival Delays in hours in each departing hour. Here, we could see that 5 am has the lowest standard deviation and low median. ## What is the best day of week to fly to minimize delays?

```
plot(data$DayOfWeek, data$ArrDelay/60, pch=19, col=data$DayOfWeek, xlab='Departure Month', ylab='ArrDelay in hours', main='ArrDelays by Day in a Week')
```

ArrDelays by Day in a Week



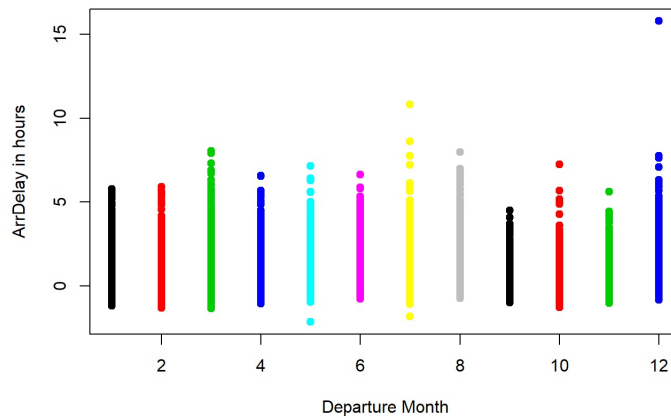
```
median_week=rep(NA,7)
std_week=rep(NA, 7)
for(i in 1:7){
  newsb=subset(data, DayOfWeek==i)
  median_week[i]<-median(newsub$ArrDelay/60, na.rm=TRUE)
  std_week[i]<-sd(newsub$ArrDelay/60, na.rm=TRUE)
}
choose_week=data.frame(col1=median_week,col2=std_week)
choose_week[
  with(choose_week, order(choose_week$col2),decreasing=FALSE),
]
```

```
##      col1      col2
## 3 -0.03333333 0.5035203
## 4 -0.01666667 0.5429525
## 6 -0.05000000 0.5654067
## 1 -0.01666667 0.5684814
## 5  0.00000000 0.5758807
## 2 -0.03333333 0.5830499
## 7 -0.03333333 0.5858247
```

The table above indicates the standard deviation and median of different Arrival Delays in hours in each day of the week. Here, we could see that Wednesday has the lowest standard deviation and low median. ## What is the best month of year to fly to minimize delays?

```
plot(data$Month, data$ArrDelay/60, pch=19, col=data$Month, xlab='Departure Month', ylab='ArrDelay in hours', main='ArrDelays by Month in a Year')
```

ArrDelays by Month in a Year

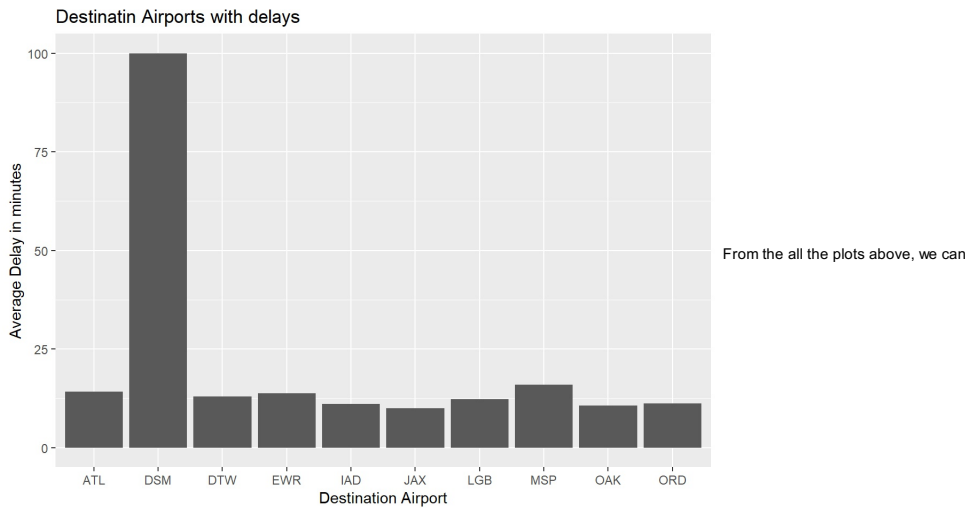


```
median_month=rep(NA,12)
std_month=rep(NA, 12)
for(i in 1:12){
  newsb=subset(data, Month==i)
  median_month[i]<-median(newsub$ArrDelay/60, na.rm=TRUE)
  std_month[i]<-sd(newsub$ArrDelay/60, na.rm=TRUE)
}
choose_month=data.frame(col1=median_month,col2=std_month)
choose_month[
  with(choose_month, order(choose_month$col2),decreasing=FALSE),
]
```

```
##      col1      col2
## 9 -0.06666667 0.3734475
## 10 -0.06666667 0.4066320
## 11 -0.06666667 0.4213849
## 1 -0.03333333 0.5306185
## 5  0.00000000 0.5317505
## 4 -0.01666667 0.5465534
## 2 -0.01666667 0.5549020
## 8 -0.01666667 0.6064537
## 7 -0.03333333 0.6136267
## 6  0.01666667 0.6183460
## 3  0.01666667 0.6417106
## 12 0.00000000 0.7068274
```

The table above indicates the standard deviation and median of different Arrival Delays in hours in each Month. Here, we could see that september has the lowest standard deviation and low median. ## What are the bad airports to fly to?

```
airport=aggregate(data$ArrDelay, by=list(Category=data$Dest), FUN=mean, na.rm='True')
airport=airport[order(airport$x,decreasing = TRUE),]
airport=subset(airport,x>10)
ggplot(aes(x = Category, y = x), data = airport ) +geom_bar(stat = "identity") + xlab("Destination Airport") +
  ylab("Average Delay in minutes") + ggtitle("Destinatin Airports with delays")
```



From the all the plots above, we can see that the best time to fly to minimize delay time is round 5 in the morning on Wednesday during September. \ In addition, we found that the Des Moines International Airport (DSM) airport has the most delay time as a destination airport from Austin Airport.\

How do patterns of flights to different destinations or parts of the country change over the course of the year?

Here, we could take a look at the pattern of the flight (to Austin or from Austin in January, June and December)

```
library(igraph)

##
## Attaching package: 'igraph'

## The following objects are masked from 'package:stats':
##
##      decompose, spectrum

## The following object is masked from 'package:base':
##
##      union

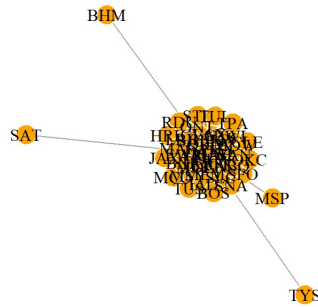
index=c(2,17,18)
network_flight=data[index]

##looking at January
network_1=subset(network_flight, Month==1 )
index_1=c(2,3)
network_1=network_1[index_1]
network_1=as.matrix(network_1)
#####creating a graph object####
graph_network_1=graph.edgelist(network_1, directed=FALSE)
sort(degree(graph_network_1),decreasing = TRUE)

## AUS DAL DFW IAH PHX ATL ORD HOU DEN JFK LAX ELP LAS MEM MCI
## 8726 1194 906 609 469 421 416 404 392 242 240 234 210 186 179
## BNA SJC CVG EWR BWI MDW LBB SAN MSY MAF ONT ABQ SLC IAD CLE
## 167 164 159 157 124 123 123 122 118 118 117 116 108 93 64
## TUL TUS SFO SNA BOS TPA JAX HRL MCO STL OKC RDU MSP SAT TYS
## 64 62 62 62 62 62 62 62 62 62 61 54 11 1 1
## BHM
## 1

#start plotting the network

V(graph_network_1)$color = "orange"
V(graph_network_1)$frame.color = 0
V(graph_network_1)$label.color = "black"
plot(graph_network_1, edge.curved=FALSE)
```

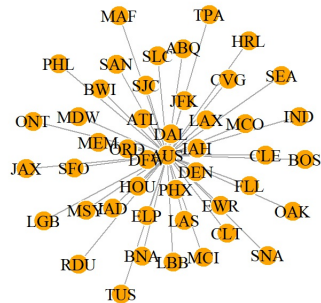


```
##looking at June
network_6=subset(network_flight, Month==6 )
index_1=c(2,3)
network_6=network_6[index_1]
network_6=as.matrix(network_6)
#####creating a graph object####
graph_network_6=graph.edgelist(network_6, directed=FALSE)
sort(degree(graph_network_6),decreasing = TRUE)
```

```
## AUS DAL DFW IAH DEN PHX ORD ATL HOU LAX JFK ELP LAS MEM SJC
## 9090 1149 889 623 493 472 464 402 381 299 236 222 214 180 172
## IAD EWR MCO BWI CLT MDW SFO SAN FLL LBB ABQ MSY MCI SLC BNA
## 170 161 128 120 120 120 120 120 112 112 112 111 110 110
## CVG CLE OAK MAF ONT BOS SNA TPA PHL LGB TUS IND RDU HRL SEA
## 109 102 60 60 60 60 60 60 60 60 60 60 60 59
## JAX
## 58
```

```
#start plotting the network
```

```
V(graph_network_6)$color = "orange"
V(graph_network_6)$frame.color = 0
V(graph_network_6)$label.color = "black"
plot(graph_network_6, edge.curved=FALSE)
```

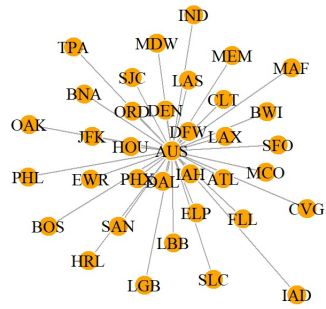


```
##looking at December
network_12=subset(network_flight, Month==12 )
index_1=c(2,3)
network_12=network_12[index_1]
network_12=as.matrix(network_12)
#####creating a graph object####
graph_network_12=graph.edgelist(network_12, directed=FALSE)
sort(degree(graph_network_12),decreasing = TRUE)
```

```
## AUS DFW DAL IAH PHX DEN HOU ORD LAX ATL ELP CLT LAS JFK EWR
## 7248 974 714 622 484 424 377 308 292 276 215 182 179 178 156
## SJC MCO SFO BWI SAN FLL LBB BNA MDW MEM BOS MAF SLC OAK PHL
## 154 130 124 124 124 124 120 116 106 94 63 62 62 62 62
## CVG IND HRL LGB TPA IAD
## 62 62 62 62 62 30
```

```
#start plotting the network
```

```
V(graph_network_12)$color = "orange"
V(graph_network_12)$frame.color = 0
V(graph_network_12)$label.color = "black"
plot(graph_network_12, edge.curved=FALSE)
```



From the graph, we can see that all

other flights flew frequently from or too Austin except TYS, SAT, MSP, BHM in January, and the top 5 airports having high degree with Austin are DAL, DFW, IAH, PHX, and ATL. The top 5 airports having high degree with Austin in June are DAL, DFW, IAH, DEN, and PHX. The top 5 airports having high degree with Austin in December are DFW, DAL, IAH, PHX, and DEN ""