

Flights Data Visualization

2024-07-28

Show data set

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr   1.5.1
## v ggplot2     3.5.1      v tibble    3.2.1
## v lubridate   1.9.3      v tidyr     1.3.1
## v purrr       1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(nycflights13)
```

```
library(glue)
```

```
flights
```

```
## # A tibble: 336,776 x 19
```

```
##   year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time
##   <int> <int> <int>   <int>         <int>         <dbl>   <int>         <int>
## 1  2013     1     1     517             515           2     830             819
## 2  2013     1     1     533             529           4     850             830
## 3  2013     1     1     542             540           2     923             850
## 4  2013     1     1     544             545          -1    1004            1022
## 5  2013     1     1     554             600          -6     812             837
## 6  2013     1     1     554             558          -4     740             728
## 7  2013     1     1     555             600          -5     913             854
## 8  2013     1     1     557             600          -3     709             723
## 9  2013     1     1     557             600          -3     838             846
## 10 2013     1     1     558             600          -2     753             745
```

```
## # i 336,766 more rows
```

```
## # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>,
## #   tailnum <chr>, origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>,
## #   hour <dbl>, minute <dbl>, time_hour <dtm>
```

```
airlines
```

```
## # A tibble: 16 x 2
```

```
##   carrier name
##   <chr>   <chr>
## 1 9E      Endeavor Air Inc.
## 2 AA      American Airlines Inc.
## 3 AS      Alaska Airlines Inc.
## 4 B6      JetBlue Airways
## 5 DL      Delta Air Lines Inc.
```

```
## 6 EV      ExpressJet Airlines Inc.
## 7 F9      Frontier Airlines Inc.
## 8 FL      AirTran Airways Corporation
## 9 HA      Hawaiian Airlines Inc.
## 10 MQ     Envoy Air
## 11 OO     SkyWest Airlines Inc.
## 12 UA     United Air Lines Inc.
## 13 US     US Airways Inc.
## 14 VX     Virgin America
## 15 WN     Southwest Airlines Co.
## 16 YV     Mesa Airlines Inc.
```

Data Preparation

```
flights = flights %>%
  sample_frac(0.1) %>%
  filter(!rowSums(is.na(.)))
```

Top 10 Popular Routes (show origin and destination)

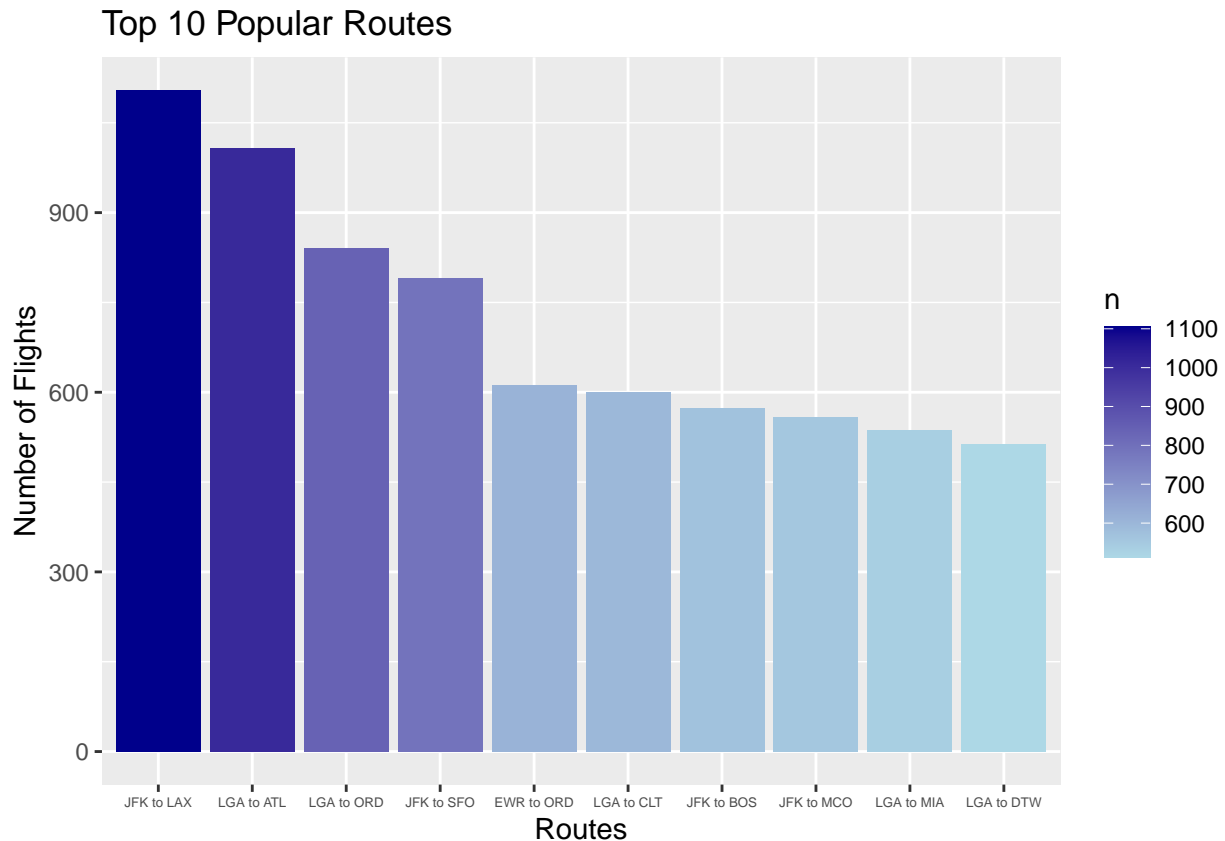
```
# Create a new data frame with only the necessary columns
flights_routes = flights %>%
  select(origin, dest)

# Count the frequency of each route and Select the top 10 routes
route_counts = flights_routes %>%
  count(origin, dest) %>%
  arrange(desc(n)) %>%
  head(10)
```

```
# Create new column "route" showing origin and destination
route = route_counts %>%
  mutate(route = glue("{origin} to {dest}")) %>%
  arrange(desc(n))
```

```
# Convert route to factor for correct ordering
route$route = factor(route$route, levels = route$route)
```

```
ggplot(route, aes(route, n, fill = n)) +
  geom_col() +
  scale_fill_gradient(low = "lightblue", high = "darkblue") +
  theme(axis.text.x = element_text(size = 5)) +
  labs(title = "Top 10 Popular Routes", x = "Routes", y = "Number of Flights")
```



Correlation between departure delay time and arrival delay time for the 3 airlines with highest number of flights

```
# Form a table with information from 'flights' and 'airlines' tables
flights_airlines = flights %>%
  left_join(airlines, by = 'carrier')

# Find 3 airlines with highest number of flight with the help of table 'airlines'
pop_airlines = flights_airlines %>%
  group_by(name) %>%
  count() %>%
  arrange(-n) %>%
  head(3)

# Select only rows of our three popular airlines and convert time from minutes to hours
delay_airlines = flights_airlines %>%
  filter(name %in% pop_airlines$name) %>%
  mutate(dep_delay = dep_delay/60, arr_delay = arr_delay/60) %>%
  rename(Airline = name)

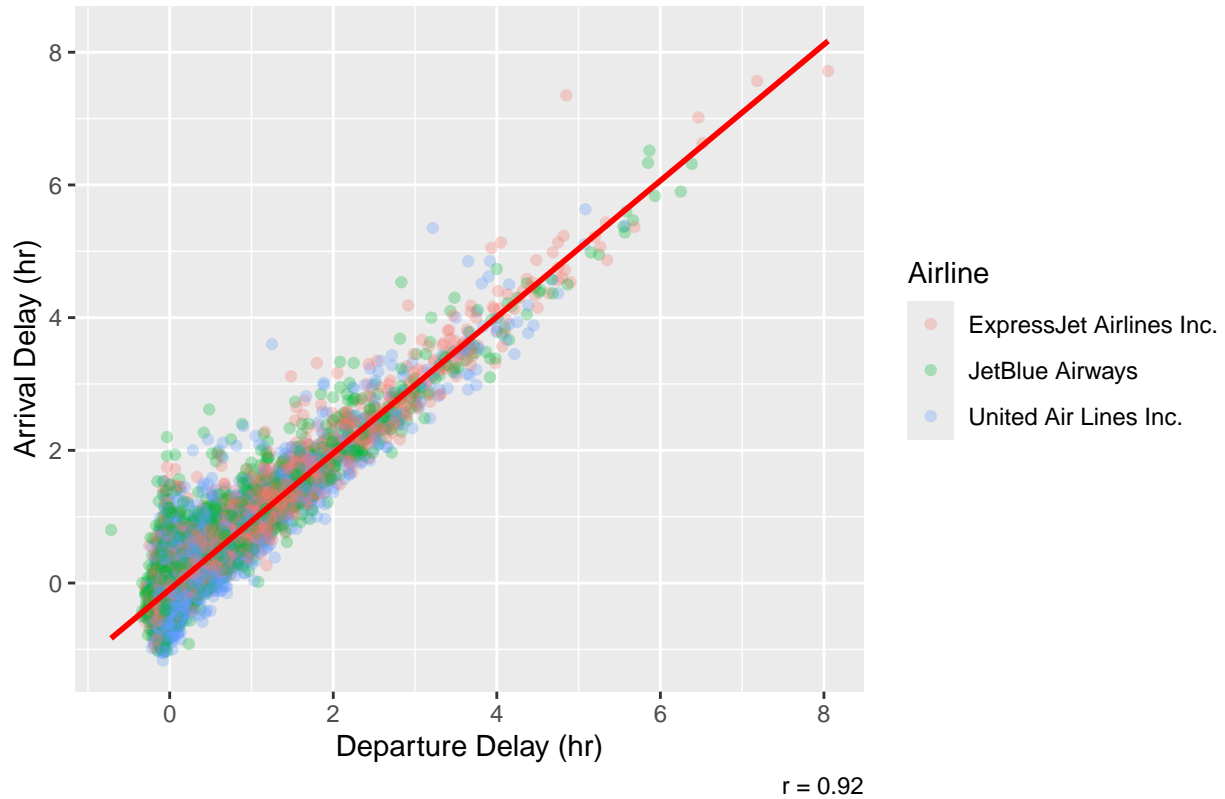
# Calculate correlation coefficient
correlation = cor(flights_airlines$dep_delay, flights_airlines$arr_delay)

ggplot(delay_airlines, aes(dep_delay, arr_delay, col=Airline)) +
  geom_point(alpha = 0.3) +
```

```
geom_smooth(method = "lm", col = "red") +  
labs(subtitle = "Correlation between departure delay and arrival delay of the three most popular airlines")
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

Correlation between departure delay and arrival delay of the three most popular airlines



```
ggplot(delay_airlines, aes(dep_delay, arr_delay, col = Airline)) +  
  geom_point(alpha = 0.3) +  
  geom_smooth(method = "lm", col = "red") +  
  facet_wrap(~Airline) +  
  labs(x = "Departure Delay (hr)", y = "Arrival Delay (hr)")
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

