Data Analysis Flights NYC 2013 😂

Import Data

```
flights <- read.csv("flights.csv")
```

Data preparation

```
Rows: 336,776
Columns: 19
$ year
                <int> 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2...
                $ month
$ day
                $ dep_time
                <int> 517, 533, 542, 544, 554, 554, 555, 557, 557, 558, 558, ...
$ sched_dep_time <int> 515, 529, 540, 545, 600, 558, 600, 600, 600, 600, 600, ...
                <int> 2, 4, 2, -1, -6, -4, -5, -3, -3, -2, -2, -2, -2, -2, -1...
$ dep_delay
$ arr_time
                <int> 830, 850, 923, 1004, 812, 740, 913, 709, 838, 753, 849,...
$ sched_arr_time <int> 819, 830, 850, 1022, 837, 728, 854, 723, 846, 745, 851,...
               <int> 11, 20, 33, -18, -25, 12, 19, -14, -8, 8, -2, -3, 7, -1...
$ arr_delay
               <chr> "UA", "UA", "AA", "B6", "DL", "UA", "B6", "EV", "B6", "...
$ carrier
               <int> 1545, 1714, 1141, 725, 461, 1696, 507, 5708, 79, 301, 4...
$ flight
               <chr> "N14228", "N24211", "N619AA", "N804JB", "N668DN", "N394...
$ tailnum
                <chr> "EWR", "LGA", "JFK", "JFK", "LGA", "EWR", "EWR", "LGA",...
$ origin
$ dest
               <chr> "IAH", "IAH", "MIA", "BQN", "ATL", "ORD", "FLL", "IAD",...
                <int> 227, 227, 160, 183, 116, 150, 158, 53, 140, 138, 149, 1...
$ air_time
               <int> 1400, 1416, 1089, 1576, 762, 719, 1065, 229, 944, 733, ...
$ distance
$ hour
                <int> 5, 5, 5, 5, 6, 5, 6, 6, 6, 6, 6, 6, 6, 6, 6, 5, 6, 6, 6.
$ minute
               <int> 15, 29, 40, 45, 0, 58, 0, 0, 0, 0, 0, 0, 0, 0, 59, 0...
```

Finding the percentage of missing values in flight

- complete value = 97.2%
- missing value = 2.8%

```
sum(complete.cases(flights))/nrow(flights)
```

0.971999192341497

Drop missing value

If the missing value is less than 5%, it's acceptable to drop it.

```
clean_df <- drop_na(flights)
clean_df %>% head(3)
```

A data.frame: 3 × 19

	year	month	day	dep_time	sched_dep_time	dep_delay	arr_time	sched_arr_time	arr_delay	carrier	flight	tailnum	ori
	<int></int>	<int></int>	<int></int>	<int></int>	<int></int>	<int></int>	<int></int>	<int></int>	<int></int>	<chr></chr>	<int></int>	<chr></chr>	<cl< td=""></cl<>
1	2013	1	1	517	515	2	830	819	11	UA	1545	N14228	EW
2	2013	1	1	533	529	4	850	830	20	UA	1714	N24211	LG
3	2013	1	1	542	540	2	923	850	33	AA	1141	N619AA	JFK

Data Analysis 🞇

Q1: What were the top 5 months that had the highest number of delayed arrivals and departure?

```
delay_m_a <- clean_df %>%
group_by(month) %>%
select(month, arr_delay) %>%
filter(arr_delay > 0) %>%
summarize(num_delay = n()) %>%
arrange(desc(num_delay))
```

delay_m_a %>% head(5)

A tibble: 5×2

month	num_delay		
<int></int>	<int></int>		
12	14394		
7	13304		
4	12522		
6	12490		
8	11629		

```
delay_m_d <- clean_df %>%
group_by(month) %>%
select(month, dep_delay) %>%
filter(dep_delay > 0) %>%
summarize(num_delay = n()) %>%
arrange(desc(num_delay))

delay_m_d %>% head(5)
```

A tibble: 5×2

month	num_delay
<int></int>	<int></int>
7	13773
12	13490
6	12558
8	11665
5	11227

Q2: What were the top 5 destination in December

```
top_d_dec <- clean_df %>%
filter(month == 12, year == 2013) %>%
count(dest) %>%
arrange(desc(n))

top_d_dec %>% head(5)
```

A data.frame: 5 ×

	2		
	dest	n	
	<chr></chr>	<int></int>	
1	ATL	1429	
2	LAX	1390	
3	MCO	1203	
4	SFO	1159	
5	CLT	1155	

```
low_at <- clean_df %>% group_by(carrier) %>%
select(carrier, air_time) %>%
summarize(air_time = n()) %>%
arrange(air_time)

low_at %>% head(5)
```


Q4:Which origin airport had the least amount of arrival "delay" time in October?

```
air_l_delay <- clean_df %>% filter(month == 10, year == 2013,) %>%
group_by(origin) %>%
select(origin, arr_delay) %>%
summarise(arr_delay =n()) %>%
arrange(arr_delay)
air_l_delay %>% head(3)
```

```
A tibble: 3 × 2

origin arr_delay

<chr> <int>
JFK 9096

LGA 9516

EWR 10006
```

```
avg_depdelayinJune <- clean_df %>%
filter(between(month,6,7), year == 2013,dep_delay > 0) %>%
group_by(carrier) %>%
summarize(mean_depdealy = mean(dep_delay),
median_depdealy = median(dep_delay)) %>%
arrange(desc(mean_depdealy))
avg_depdelayinJune %>% head (16)
```

A tibble: 16×3

carrier	mean_depdealy	median_depdealy		
<chr></chr>	<dbl></dbl>	<dbl></dbl>		
00	131.00000	131		
FL	63.77044	25		
YV	62.61404	39		
9E	58.58470	34		
EV	57.67910	35		
VX	56.40036	16		
MQ	55.79965	35		
В6	52.03178	30		
DL	47.80134	22		
US	44.90656	24		
AA	44.71014	20		
WN	44.66869	20		
F9	41.06897	25		
UA	38.46728	17		
НА	33.62500	8		
AS	31.37778	8		