hw1

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1. Import

```
library(haven)
library(ggplot2)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
library(tidyr)
library(purrr)
library(cowplot)
flights <- read_dta("version10_RA.dta") %>% tibble()
```

2. Data Definitions

```
market - the combination of origin and destination airports
origin - the origin airport
dest - the destination airport
year - year when the travel occurred
quarter - quarter when the travel occurred
carrier - the carrier who transported the passengers (e.g. American - AA)
nonstopmiles mkt - nonstop distance in miles
totalpassengers - passengers transported by the carrier
medianmktfare mkt - median fare charged by the carrier
meanmktfare mkt - mean fare charged by the carrier
```

3.

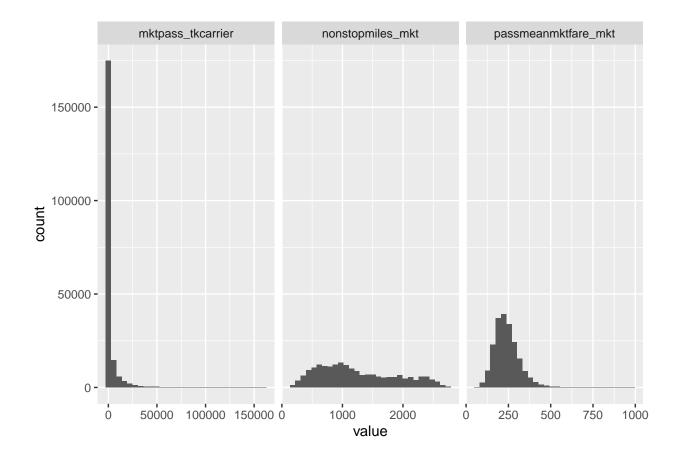
```
summary.stats <- function(x,...){
    c(mean=mean(x, ...),
    sd=sd(x, ...),
    min=min(x, ...),
    lower = quantile(x, ..., 1/4),
    median=median(x, ...),
    upper = quantile(x, ..., 3/4),
    max=max(x,...))
}
num.cols <- c("passmeanmktfare_mkt","mktpass_tkcarrier","nonstopmiles_mkt")
flights.summary <- sapply(select(flights, all_of(num.cols)), summary.stats) %>% as.data.frame()
flights.summary
```

##		passmeanmktfare_mkt	mktpass_tkcarrier	nonstopmiles_mkt
##	mean	239.77945	2491.877	1242.8215
##	sd	72.96611	7120.953	637.9157
##	min	59.97292	90.000	151.7254
##	lower.25%	189.53795	210.000	726.0604
##	median	230.44464	480.000	1102.4904
##	upper.75%	279.65482	1280.000	1733.1265
##	max	979.01361	160100.000	2719.2703

We can see that for passmeanmktfare_mkt and mktpass_tkcarrier, there are many outliers that bring up the maximum significantly over the median. From the histograms, we see that the distribution of nonstopmiles_mkt is fairly even, and confirms lack of outliers.

```
flights %>%
  select(all_of(num.cols)) %>%
  gather() %>%
  ggplot(aes(value)) +
  geom_histogram() +
  facet_grid(~key, scales = 'free_x')
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



4

We'll first map all the carriers to a group, and check its distribution

```
carrier_groups = list(AA='Legacy',
                      DL='Legacy',
                      UA='Legacy',
                      WN='LCC',
                      B6='LCC')
flights$carrier_group = flights$tkcarrier %>% recode(!!!carrier_groups, .default='Other')
flights$carrier_group %>% table()
##
##
      LCC Legacy Other
    23887 107672 73214
group_stats =
for (g in flights$carrier_group %>% unique()){
  print(paste0('Carrier Group: ',g))
  group.stats <- sapply(flights %>%
           filter(carrier_group == g) %>%
           select(all_of(num.cols)),
         summary.stats) %>%
```

```
as.data.frame()
print(group.stats)
}
```

## [1] "Carrier Group	: Legacy"				
## passmear	mktfare_mkt	mktpass_tkcarrier	nonstopmiles_mkt		
## mean	254.82647	1957.027	1240.8139		
## sd	74.09149	6117.566	630.3758		
## min	68.07591	90.000	152.0161		
## lower.25%	203.06381	200.000	731.0565		
## median	245.52037	440.000	1101.3259		
## upper.75%	294.55953	1080.000	1723.6611		
## max	923.81018	146590.000	2719.2703		
## [1] "Carrier Group	o: Other"				
## passmear	mktfare_mkt	${\tt mktpass_tkcarrier}$	nonstopmiles_mkt		
## mean	228.54031	2090.861	1247.4475		
## sd	71.72090	5671.783	642.5461		
## min	59.97292	90.000	151.7254		
## lower.25%	178.99164	190.000	726.7436		
## median	217.61833	420.000	1101.7332		
## upper.75%	268.78526	1090.000	1741.6509		
## max	979.01361	114960.000	2719.2703		
## [1] "Carrier Group: LCC"					
## passmear	mktfare_mkt	${\tt mktpass_tkcarrier}$	nonstopmiles_mkt		
## mean	206.40230	6131.865	1237.6926		
## sd	51.07686	12344.255	657.1104		
## min	68.82486	90.000	153.2130		
## lower.25%	171.48658	480.000	693.3195		
## median	206.88779	1310.000	1111.8335		
## upper.75%	241.41648	5830.000	1745.8091		
## max	462.05280	160100.000	2714.5547		