



STOR 320 Data Transformation II

Lecture 5

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Data Transformation II Info

- Finish Reading Chapter 5 and Practice the Code in R4DS
- Covers
 - The Pipe
 - Statistical Summaries
 - Grouped Summaries
 - Helpful Functions
- Builds Off Last Tutorial



The Pipe

- Useful for Combining Multiple Steps of Operations
- Represented by `%>%`
- Reads as “Then”
- Works Like a Composite Function From Algebra

$$f(x) = 3x + 4$$

$$g(x) = 2x$$

$$h = 1$$

$$f(g(h)) = 3(2(1)) + 4 = 10$$



$$\begin{aligned} \text{OUT} &= h \%>\% \\ &\quad g() \%>\% \\ &\quad \quad f() \end{aligned}$$

$$\text{OUT} = 10$$

The Pipe

- Chaining with the Pipe



```
```{r,eval=F}
f2e.pipedream =
 # Acknowledge the Original Data
 flights %>%
 # Input Original Data and Perform Mutations
 transmute(dep_hr=dep_time%%100+(dep_time%%100)/60,
 sched_dep_hr=sched_dep_time%%100+(sched_dep_time%%100)/60,
 arr_hr=arr_time%%100+(arr_time%%100)/60,
 sched_arr_hr=sched_arr_time%%100+(sched_arr_time%%100)/60) %>%

 mutate(dep_delay_hr=dep_hr-sched_dep_hr,
 arr_delay_hr=arr_hr-sched_arr_hr) %>%

 mutate(percent_dep_delay_hr=percent_rank(dep_delay_hr)) %>%
 # Input Modified Data and Filter the observations
 filter(percent_dep_delay_hr<0.1|percent_dep_delay_hr>0.9) %>%

 # Input Modified Data and Sort according to percent_dep_delay_hr
 arrange(desc(percent_dep_delay_hr))
```

dep_hr <dbl>	sched_dep_hr <dbl>	arr_hr <dbl>	sched_arr_hr <dbl>	dep_delay_hr <dbl>	arr_delay_hr <dbl>
23.35000	8.166667	1.583333	10.33333	15.18333	-8.750000
22.95000	7.983333	1.350000	10.43333	14.96667	-9.083333
22.71667	8.500000	1.000000	11.10000	14.21667	-10.100000
23.40000	10.266667	1.233333	12.45000	13.13333	-11.216667
19.35000	6.250000	21.583333	8.70000	13.10000	12.883333

5 rows | 1-6 of 7 columns



# The Pipe

## Why use

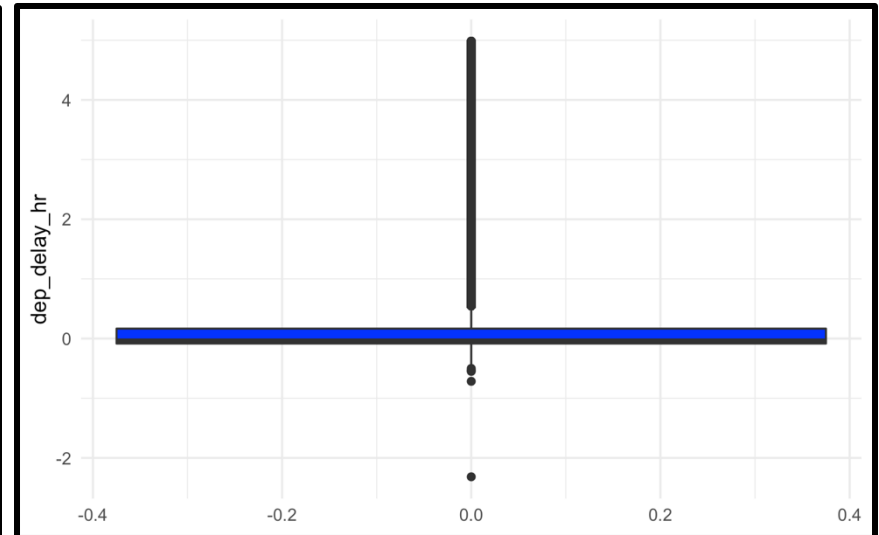
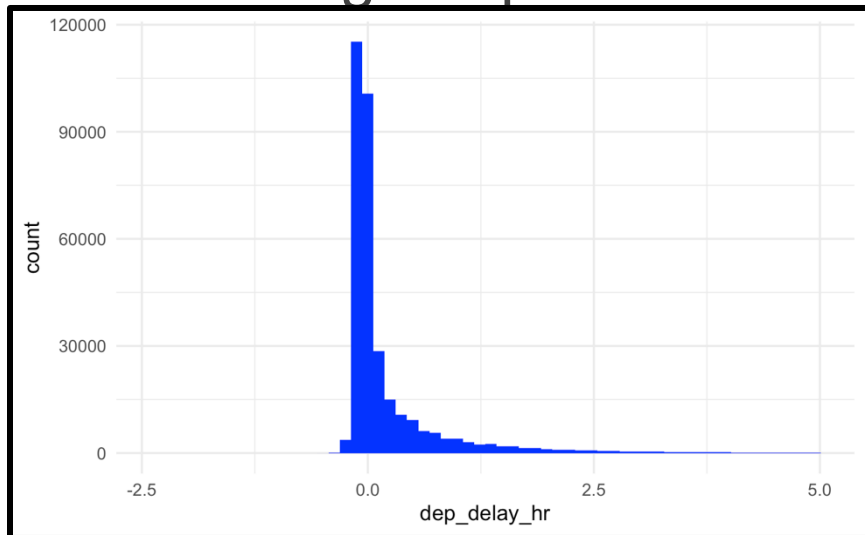
- Avoid nested functions
- Minimize number of local variables
- Easier to add steps in the sequence

## Why not to use

- Debug
- Can't handle multiple inputs
- Can't handle complex code structure

# summarize()

- Summarizing All Data
- Using Graphics



Both the histogram and the boxplot are made from summary statistics.

**(Statistical Transformations in Ch. 3)**



# summarize()

- Summarizing All Data

```
```{r}
dep_delay_hr.summary1 = summarize(f3e.pipedream,
                                   n=n(),
                                   mean=mean(dep_delay_hr, na.rm=T),
                                   var=var(dep_delay_hr, na.rm=T),
                                   sd=sd(dep_delay_hr, na.rm=T))

dep_delay_hr.summary2 =
  f3e.pipedream %>%
  summarize(n=n(),
            min=min(dep_delay_hr),
            Q1=quantile(dep_delay_hr, 0.25),
            Q2=quantile(dep_delay_hr, 0.5),
            Q3=quantile(dep_delay_hr, 0.75),
            max=max(dep_delay_hr),
            IQR=Q3-Q1)
```
```

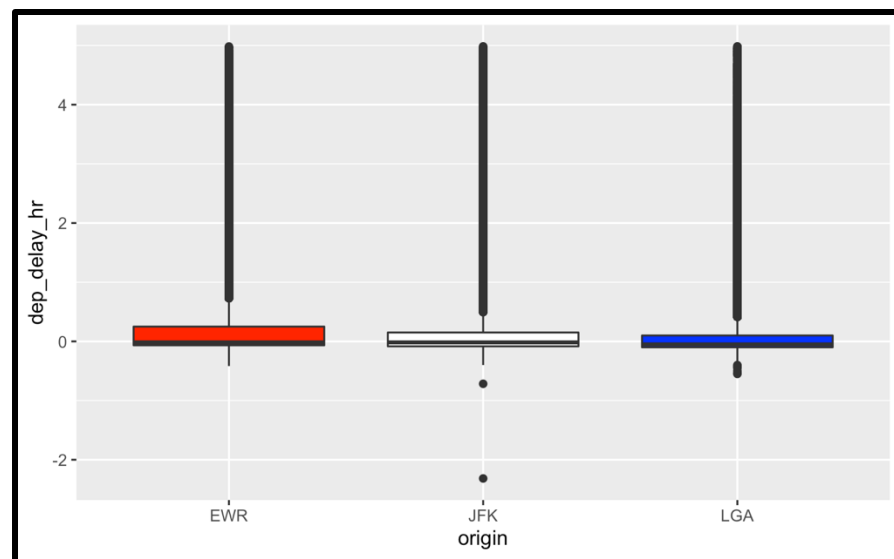
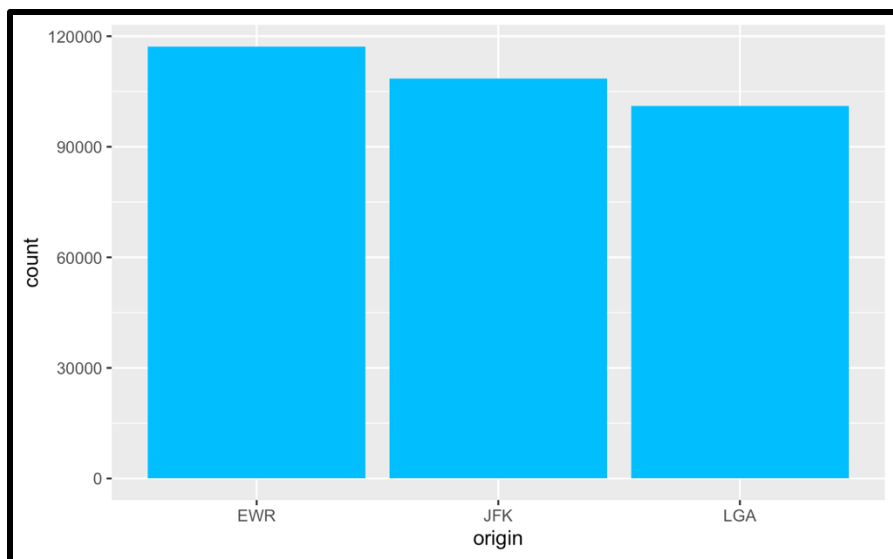
| n      | mean     | var       | sd        |
|--------|----------|-----------|-----------|
| 326848 | 0.192752 | 0.3506166 | 0.5921289 |

| n      | min       | Q1         | Q2         | Q3        | max      | IQR  |
|--------|-----------|------------|------------|-----------|----------|------|
| 326848 | -2.316667 | -0.0833333 | -0.0333333 | 0.1666667 | 4.983333 | 0.25 |



# summarize() with group\_by()

- Summarizing Data by Groups
- Using Graphics







# summarize() with group\_by()

- Summarizing Data by Groups
  - Using Tables

```
```\r\ngroup.summary1 = f3e.pipedream %>%  
  group_by(origin) %>%  
  summarize(n=n())  
  
group.summary2 =  
  f3e.pipedream %>%  
  group_by(origin) %>%  
  summarize(n=n(),  
    min=min(dep_delay_hr),  
    Q1=quantile(dep_delay_hr,0.25),  
    Q2=quantile(dep_delay_hr,0.5),  
    Q3=quantile(dep_delay_hr,0.75),  
    max=max(dep_delay_hr),  
    IQR=Q3-Q1,  
    nLow=sum(dep_delay_hr<Q1-1.5*IQR),  
    propHigh=mean(dep_delay_hr>Q3+1.5*IQR))  
```\r\`
```

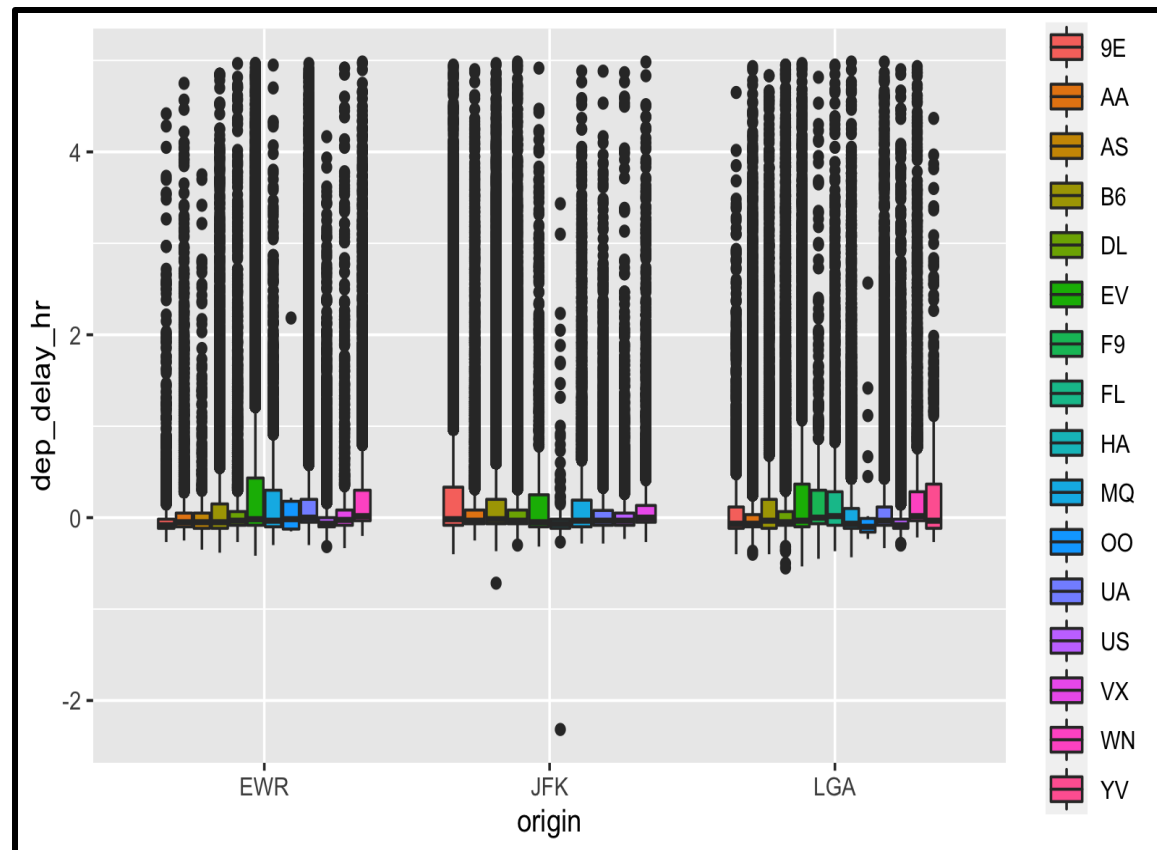
origin	n
EWR	117209
JFK	108486
LGA	101153

origin <chr>	n <int>	min <dbl>	Q1 <dbl>	Q2 <dbl>	Q3 <dbl>	max <dbl>	IQR <dbl>	nL... <int>	propHigh <dbl>
EWR	117209	-0.4166667	-0.06666667	-0.01666667	0.25	4.983333	0.3166667	0	0.1259204
JFK	108486	-2.3166667	-0.08333333	-0.01666667	0.15	4.983333	0.2333333	2	0.1372988
LGA	101153	-0.5500000	-0.10000000	-0.05000000	0.10	4.983333	0.2000000	7	0.1466491



# summarize() with group\_by()

- Multiple Groups
  - Using Graphics





# summarize() with group\_by()

- Multiple Groups
- Using Tables

```
{r}
group.summary3 =
 f3e_ninedream %>%
 group_by(origin, carrier) %>%
 summarize(n=n(),
 min=min(dep_delay_hr),
 Q1=quantile(dep_delay_hr,0.25),
 Q2=quantile(dep_delay_hr,0.5),
 Q3=quantile(dep_delay_hr,0.75),
 max=max(dep_delay_hr),
 IQR=Q3-Q1,
 nLow=sum(dep_delay_hr<Q1-1.5*IQR),
 propHigh=mean(dep_delay_hr>Q3+1.5*IQR))
```

origin	carrier	n	min	Q1	Q2	Q3	max
EWR	9E	1199	-0.2666667	-0.1166667	-0.08333333	-0.01666667	4.416667
EWR	AA	3376	-0.2500000	-0.1000000	-0.05000000	0.05000000	4.750000
EWR	AS	712	-0.3500000	-0.1166667	-0.05000000	0.05000000	3.750000
EWR	B6	6446	-0.3833333	-0.1166667	-0.05000000	0.15000000	4.850000
EWR	DL	4281	-0.2666667	-0.08333333	-0.03333333	0.06666667	4.966667
EWR	EV	41592	-0.4166667	-0.08333333	-0.01666667	0.43333333	4.966667
EWR	MQ	2095	-0.3000000	-0.1000000	-0.03333333	0.30000000	4.950000
EWR	OO	6	-0.1500000	-0.1250000	-0.01666667	0.1791667	2.183333
EWR	UA	45561	-0.3000000	-0.0500000	0.00000000	0.20000000	4.966667
EWR	US	4326	-0.3166667	-0.1000000	-0.06666667	0.00000000	4.166667
EWR	VX	1554	-0.3333333	-0.08333333	-0.02500000	0.08333333	4.916667
EWR	WN	6061	-0.2000000	-0.03333333	0.01666667	0.30000000	4.983333
JFK	9E	13801	-0.4000000	-0.08333333	-0.01666667	0.33333333	4.950000
JFK	AA	13617	-0.2500000	-0.06666667	-0.03333333	0.08333333	4.900000
JFK	B6	41005	-0.7166667	-0.06666667	-0.01666667	0.20000000	4.966667
JFK	DL	20551	-0.3000000	-0.06666667	-0.03333333	0.08333333	4.983333
JFK	EV	1315	-0.3166667	-0.1000000	-0.05000000	0.25000000	4.916667



# Useful Summary Functions

- Measures of Center
  - `mean()`
  - `median()`
  - `mode()`
- Measures of Spread
  - `var()`
  - `sd()`
  - `IQR()`
  - `mad()`
- Measures of Rank
  - `min()`
  - `max()`
  - `quantile()`



# Useful Summary Functions

- Measures of Position
  - Order Matters
  - `first() = x[1]`
  - `last() = x[length(x)]`
  - `nth(,k) = x[k]`
- Counts
  - `n()`
  - `n_distinct()`
- Counts/Proportions for Logical
  - `sum()`
  - `mean()`
  - Example
    - `sum(x>10)`
    - `mean(x>10)`



# Case Study

- Flight Accuracy
  - Accurate Flight Means
    - Departure Delay = 0
    - Arrival Delay = 0

- Bad Metric

$$Accuracy = delay_{dep} + delay_{arr}$$

$$Accuracy = (delay_{dep} + delay_{arr})/2$$

- Good Metrics

$$Accuracy = |delay_{dep}| + |delay_{arr}|$$

$$Accuracy = \sqrt{delay_{dep}^2 + delay_{arr}^2}$$



# Case Study

- Summary Table
  - Step 1: Accuracy Variable
  - Step 2: Grouping
  - Step 3: Summarize Info
    - Mean
    - Standard Error
    - Lower Bound (95% CI)
    - Upper Bound (95% CI)

```
```{r}
accuracy<-
  f.pipedream3 %>%
  transmute(carrier,origin,
    accuracy=abs(dep_delay_hr)+abs(arr_delay_hr)) %>%
  group_by(carrier,origin) %>%
  summarize(n=n(),
    avg=mean(accuracy,na.rm=T),
    se=sd(accuracy,na.rm=T)/sqrt(n),
    low=avg-2*se,
    high=avg+2*se
  )
```
```



# Case Study

- Sorted by Average Accuracy
  - Best Carriers/Origin

```
> head(arrange(accuracy, avg), 5)
A tibble: 5 x 7
Groups: carrier [3]
 carrier origin n avg se low high
 <chr> <chr> <int> <dbl> <dbl> <dbl> <dbl>
1 US EWR 4322 0.505 0.0123 0.481 0.530
2 US JFK 2960 0.509 0.0152 0.479 0.539
3 US LGA 12517 0.544 0.0121 0.520 0.569
4 HA JFK 342 0.556 0.0362 0.483 0.628
5 UA JFK 4367 0.591 0.0173 0.556 0.625
```

- Worst Carriers/Origin

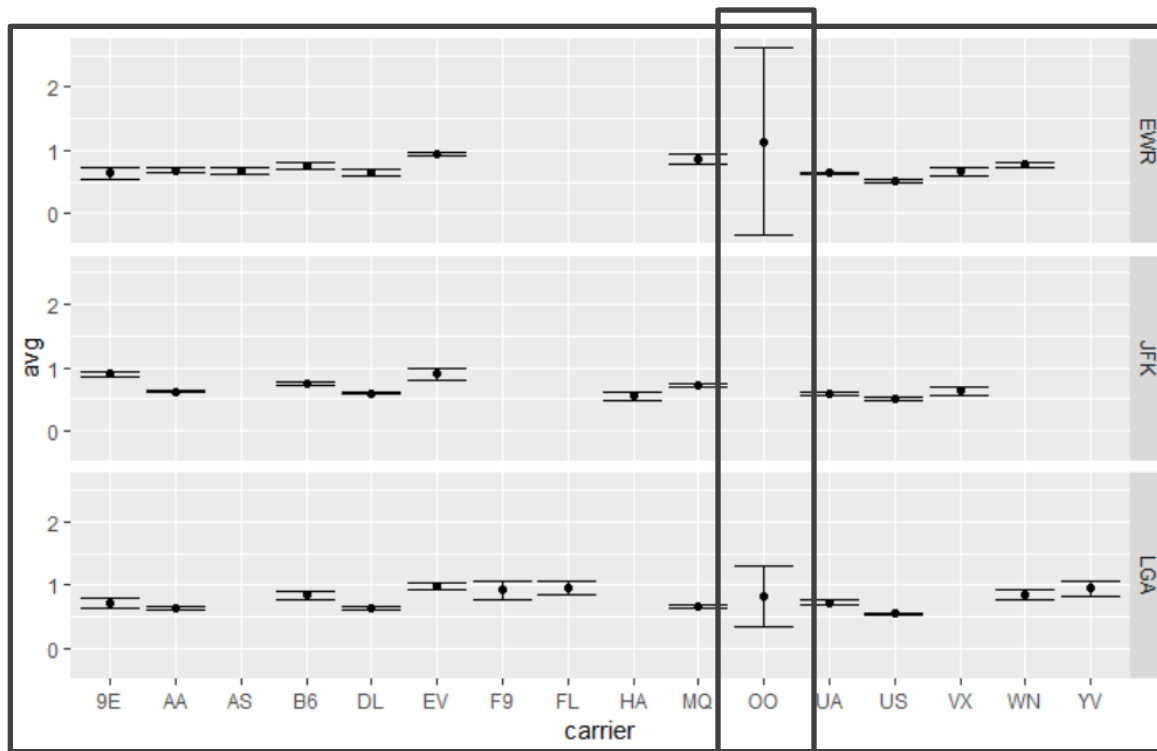
```
> head(arrange(accuracy, desc(avg)), 5)
A tibble: 5 x 7
Groups: carrier [4]
 carrier origin n avg se low high
 <chr> <chr> <int> <dbl> <dbl> <dbl> <dbl>
1 OO EWR 6 1.14 0.737 -0.334 2.61
2 EV LGA 8086 0.986 0.0265 0.933 1.04
3 YV LGA 542 0.954 0.0597 0.835 1.07
4 FL LGA 3136 0.952 0.0545 0.843 1.06
5 EV EWR 40571 0.952 0.0125 0.927 0.977
```





# Case Study

- 95% Confidence Intervals



Carrier “OO” Creates a  
Visual Problem Due to  
Small Sample Size



# Case Study

```
library(ggplot2)
ggplot(filter(accuracy, carrier != "oo")) +
 geom_point(aes(x=carrier, y=avg)) +
 geom_errorbar(aes(x=carrier, ymin=low, ymax=high)) +
 facet_grid(origin~.)
```

- 95% Confidence Intervals

