### Heatmaps in R

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### Section 1

What is a heatmap?

## What is a heatmap?

A heatmap is graphic that will display your data in a colorful grid. This is great for seeing trends and patterns in your data.

Generally, heatmaps are used to represent data where you have 2 categorical variables with a third continuous variable, though this is not explicitly necessary.

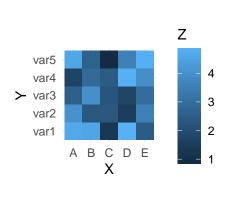
#### Advantages

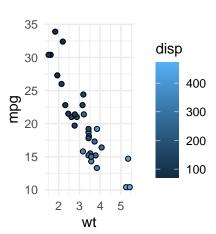
- Adds another variable
- Easier to interpret
- Color is easy to parse

#### Disadvantages

- Need right choice of colors and data
- Only useful if clear
- A different plot may be better

## **Examples**





### When to use a heatmap

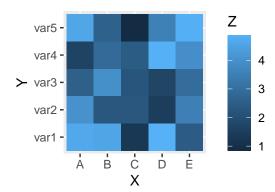
Heatmaps are applicable to both continuous and discrete variables. When working with discrete variables, distinguishing between similar colors can be challenging. To improve visibility, it is recommended to use a color palette of no more than nine colors, such as ROYGBIV, black, and white.

In the case of continuous variables, you have the option to use a sequential color scale, transitioning from a lighter hue to a darker one or vice versa. Alternatively, you can create a divergent color scale, transitioning between two distinct colors. A divergent color scale may be more useful when the center and ends of the value range are meaning full.

## ggplot2

You can create a simple and quick heatmap using the geom\_tile() function in ggplot2, but these heatmaps are very limited in their functionality.

```
ggplot(data, aes(X, Y, fill = Z)) +
geom_tile() + coord_equal()
```



### Section 2

# Pheatmap Package

## Pheatmap Package

Pheatmap (Pretty Heatmaps) is a package for R that supercharges heatmaps and allows you to create incredibly complex and helpful heatmaps. Although it is slightly more difficult to use when compared to ggplot, it is highly specialized and focused. It can plot more than 3 variables in a single heatmap, and more importantly, it groups rows and/or columns.

The grouping specifically can be incredibly helpful for visually spotting trends and similarities.

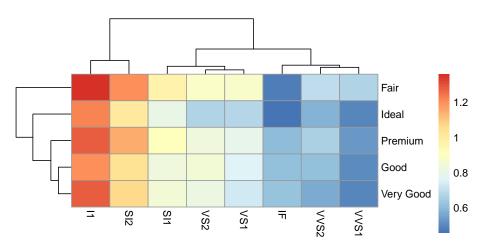
### Dendrogram

Dendrograms, create a higherarchy of similar groups. Eventhough they can be standalone, and are not always seen with heatmaps, pheatmap includes dendrograms by default. Groups that are most similar get pooled together until there are no more groupings possible. The distance between groups directly reflects the difference between groups. By default these groupings are made using the rows or columns euclidean distance.

#### A Bad Tournament

You can think of it as a tournament bracket where seeding determines your matchups, your initial matches will be the strongest teams against each other and the weakest teams against each other. The distance between the connections symbolizes the overall difference in skill.

### Pheatmap with no Arguments



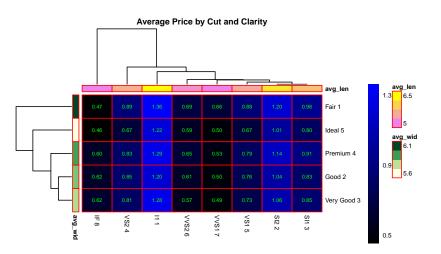
#### Make Annotations

```
# Annotations
ann1 <- diamonds %>%
  group by(clarity) %>%
  summarise(avg len = mean(x)) %>%
  column to rownames("clarity")
ann2 <- diamonds %>%
  group by(cut) %>%
  summarise(avg wid = mean(y)) %>%
  column to rownames("cut")
# Annotations colors
ann_colors <- list(
  avg_len = c("violet", "yellow"),
  avg_wid = c(brewer.pal(9, "YlGn"))
```

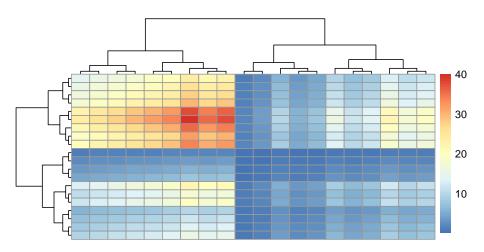
#### Lots of Function

```
pheatmap(df,
  cluster rows = TRUE, cluster cols = TRUE,
  show_rownames = TRUE, show_colnames = TRUE,
  labels_row = paste(rownames(df), 1:5),
  labels col = paste(colnames(df), 1:8), # overlay
  color = colorRampPalette(c("black", "blue"))(100),
  border color = "red",
  clustering distance cols = "correlation",
 main = "Average Price by Cut and Clarity",
  annotation col = ann1,
  annotation row = ann2,
  annotation colors = ann colors,
  legend breaks = c(.5, .9, 1.3),
 display numbers = TRUE,
 number color = "green",
 fontsize = 6,
```

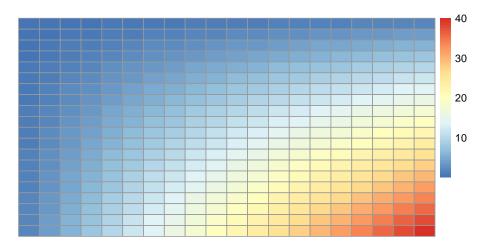
#### Lots of Function



# Just Pheatmap



# No Dendrogram



### Color

0	1	1	1	2	2	2	3	3	4	4	4	5	5	5	6	6	6	7	7
1	1	2	3	4	4	5	6	6	7	8	8	9	10	10	11	12	13	13	14
1	2	3	4	5	6	7	8	9	10	12	13	14	15	16	17	18	19	20	21
1	3	4	6	7	8	10	11	13	14	15	17	18	20	21	22	24	25	27	28
2	4	5	7	9	10	12	14	16	18	19	21	23	24	26	28	30	32	33	35
2	4	6	8	10	13	15	17	19	21	23	25	27	29	32	34	36	38	40	42
2	5	7	10	12	15	17	20	22	24	27	29	32	34	37	39	42	44	47	49
3	6	8	11	14	17	20	22	25	28	31	34	36	39	42	45	48	50	53	56
3	6	9	13	16	19	22	25	28	32	35	38	41	44	47	50	54	57	60	63
4	7	10	14	18	21	24	28	32	35	38	42	46	49	52	56	60	63	66	70
4	8	12	15	19	23	27	31	35	38	42	46	50	54	58	62	65	69	73	77
4	8	13	17	21	25	29	34	38	42	46	50	55	59	63	67	71	76	80	84
5	9	14	18	23	27	32	36	41	46	50	55	59	64	68	73	77	82	86	91
5	10	15	20	24	29	34	39	44	49	54	59	64	69	74	78	83	88	93	98
5	10	16	21	26	32	37	42	47	52	58	63	68	74	79	84	89	94	100	105
6	11	17	22	28	34	39	45	50	56	62	67	73	78	84	90	95	101	106	112
6	12	18	24	30	36	42	48	54	60	65	71	77	83	89	95	101	107	113	119
6	13	19	25	32	38	44	50	57	63	69	76	82	88	94	101	107	113		126
7	13	20	27	33	40	47	53	60	66	73	80	86	93	100	106	113	120	126	133
7	14	21	28	35	42	49	56	63	70	77	84	91	98	105	112	119	126	133	140

40

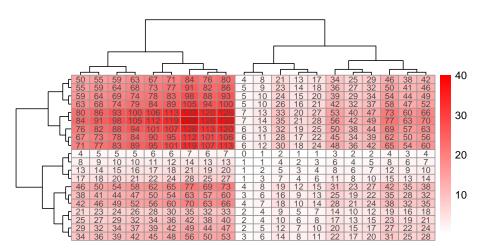
40

30

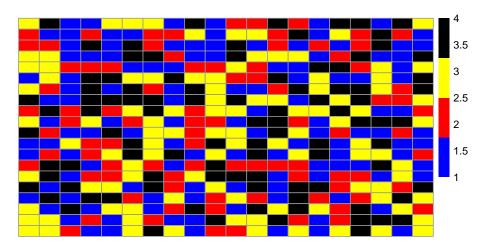
20

10

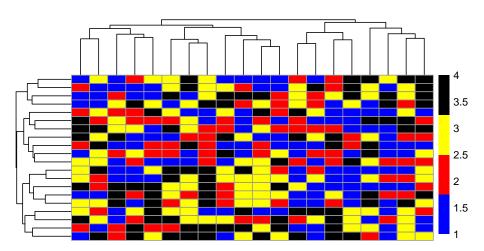
### Pheatmap example



# Pheatmap example



# Pheatmap example



### Section 3

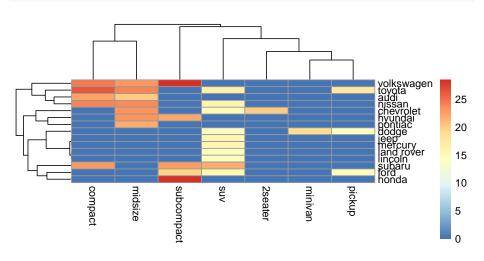
# Applying Heatmaps

Let's do it with a simple dataset we're familiar with.

```
mpg2 <- mpg %>%
  group_by(manufacturer, class) %>%
  mutate(avg_mpg = (cty + hwy) / 2) %>%
  summarize(avg_mpg = mean(avg_mpg)) %>%
  spread(class, avg_mpg) %>%
  column_to_rownames("manufacturer")
mpg2[is.na(mpg2)] = 0
mpg2[1:5, 1:6]
```

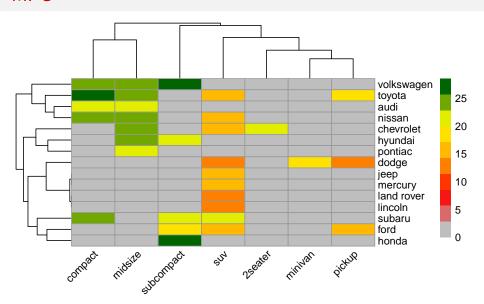
```
##
            2seater
                    compact midsize
                                    minivan
                                             pickup subcompact
               0.0 22.43333
                              20.0
                                    0.00000
                                            0.00000
                                                       0.00000
## andi
## chevrolet
           20.1
                    0.00000
                              23.2
                                    0.00000
                                            0.00000
                                                       0.00000
## dodge
               0.0 0.00000
                               0.0 19.09091 14.07895
                                                       0.00000
## ford
               0.0
                    0.00000
                               0.0
                                    0.00000 14.71429
                                                      19.55556
                                    0.00000
## honda
               0.0
                    0.00000
                               0.0
                                             0.00000
                                                      28.50000
```

#### pheatmap(mpg2)



That didn't look great. Let's clean it up.

```
pheatmap(mpg2,
    angle_col = 45,
    color = colorRampPalette(
        c("gray", "red", "orange", "yellow", "darkgreen")
    )(10),
)
```

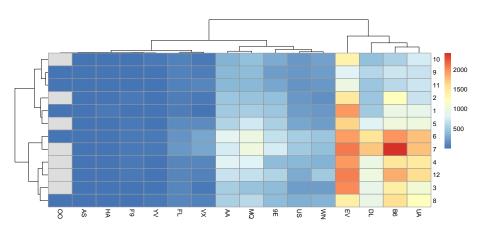


We want to plot the total number of hours flights were late arrivals each month, broken up by each carrier.

```
flights2 <- flights %>%
  filter(arr_delay > 0) %>%
  group_by(month, carrier) %>%
  summarize(total_delay = sum(arr_delay / 60)) %>%
  spread(carrier, total_delay) %>%
  column_to_rownames("month")
```

We can now plot it.

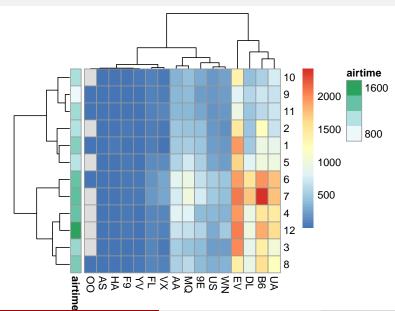
#### pheatmap(flights2)



Additionally, we would like to overlay the total airtime (in days) for each month. We can do this like as well.

```
airtime <- flights %>%
  filter(arr_delay > 0) %>%
  group_by(month) %>%
  summarize(airtime = sum(air_time / (60 * 24))) %>%
  column_to_rownames("month")
```

```
pheatmap(flights2, annotation_row = airtime)
```

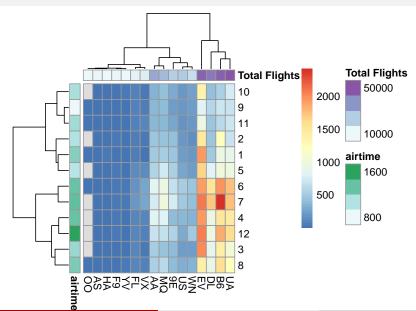


We also know certain airlines are preferred, so we would like to see the total number of flights by airline.

```
flightcounts <- flights %>%
  group_by(carrier) %>%
  summarize(total_flights = n()) %>%
  column_to_rownames("carrier")

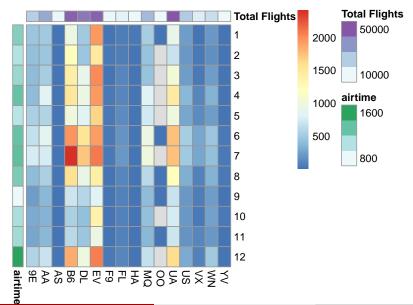
colnames(flightcounts) = "Total Flights"
```

```
pheatmap(flights2, annotation_row = airtime, annotation_col = flight
```



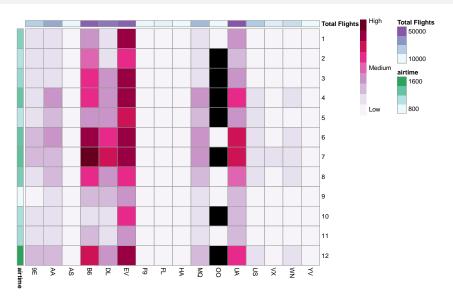
If we're not interested in the dendrogram, we can remove it easily.

```
pheatmap(flights2,
   annotation_row = airtime,
   annotation_col = flightcounts,
   cluster_row = FALSE,
   cluster_cols = FALSE
)
```

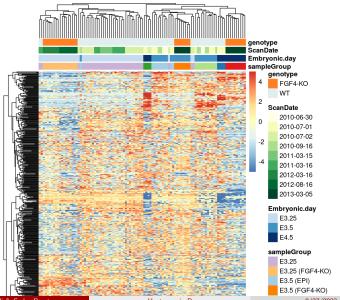


Finally, we can clean up our main legend, change the color scheme, and very clearly show missing values for our final figure.

```
pheatmap(flights2,
   annotation_row = airtime,
   annotation_col = flightcounts,
   cluster_row = FALSE,
   cluster_cols = FALSE,
   legend_breaks = c(150, 1200, 2400),
   legend_labels = c("Low", "Medium", "High"),
   color = RColorBrewer::brewer.pal(9, "PuRd"),
   na_col = "black"
)
```



# You can also choose to go bananas



#### Section 4

Wrapping up