Introduction to R

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This document was made using R-Markdown.

- R-Markdown is great because you can combine text with R code and results.
 - Text is formatted using markdown syntax,
 - It's saved as an . Rmd file.
- You can also just simply write .R code that is analogous to a .do file where there's a list of commands with comments. No need for this fancy .Rmd Markdown business.

This document was made using R-Markdown.

• It's pretty cool to integrate code with results right into the presentation.

Input

```
```{r}
sin(3)
```
```

Output

```
sin(3)
## [1] 0.14112
```

Learn more about R-Markdown

- If you are curious about markdown formatting, you can check out this handy guide. https://www.markdownguide.org/basic-syntax/
- To learn more about R-Markdown, this "cheat-sheet" https://www.rstudio.com/wp-content/uploads/2015/02/rmarkdown-cheatsheet.pdf, is a good starting place.

Running code

Code Chunks

This is a code chunk that will evaluate the command sin(3)

```
```{r}
sin(3)
```
```

The output of this will appear as follows:

```
## [1] 0.14112
```

Inline Code

You can also create calculations inline for example r sin(3) will evaluate sin(3). With output rendering like this 0.14112.

Loading packages

```
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
```

You can install packages using the install.packages() syntax, but see the section on pacman below before worrying about that.

head(diamonds)

```
## # A tibble: 6 x 10
##
     carat cut
                     color clarity depth table price
                                                         Χ
     <dbl> <ord>
                     <ord> <ord>
                                   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
## 1
     0.23 Ideal
                           SI2
                                    61.5
                                            55
                                                 326
                                                      3.95
                                                            3.98
                                                                  2.43
## 2
     0.21 Premium
                           SI1
                                    59.8
                                                 326
                                            61
                                                      3.89
                                                            3.84
                                                                  2.31
## 3
     0.23 Good
                           VS1
                                    56.9
                                            65
                                                 327
                                                      4.05
                                                            4.07
                                                                  2.31
## 4
     0.29 Premium
                           VS2
                                    62.4
                                                            4.23
                                                                  2.63
                                            58
                                                 334
                                                      4.2
## 5
     0.31 Good
                           SI2
                                    63.3
                                                 335
                                                           4.35 2.75
                                            58
                                                     4.34
                                                 336
## 6
     0.24 Very Good J
                           VVS2
                                    62.8
                                            57
                                                      3.94
                                                            3.96 2.48
```

We can get a list of column names fairly easily too

```
names(diamonds)
## [1] "carat" "cut" "color" "clarity" "depth" "table" "price"
## [8] "x" "y" "z"
```

```
diamonds %>% group_by(color) %>% summarise(mean(price))
## # A tibble: 7 x 2
     color `mean(price)`
     <ord>
                   <dbl>
## 1 D
                   3170.
## 2 E
                   3077.
## 3 F
                   3725.
## 4 G
                   3999.
## 5 H
                   4487.
## 6 I
                   5092.
## 7 J
                   5324.
```

Here the pipe command %>% will take the output from the command on the left and "pipes" it as input to the command on the right.

```
diamonds %>%
   group_by(color) %>%
   summarise(mean(price))
## # A tibble: 7 x 2
   color `mean(price)`
     <ord>
                   <dbl>
##
## 1 D
                   3170.
## 2 E
                   3077.
## 3 F
                   3725.
## 4 G
                   3999.
## 5 H
                   4487.
## 6 I
                   5092.
## 7 J
                   5324.
```

- 1. You can just move onto a new line without an error. This is neat because it allows us to write cleaner code and you don't need a delimiter or the /// you may be used to from stata. So we can rewrite the above like this.
- 2. In more recent versions of R (4.1 and above), you don't need to have dpylr installed to pipe. You can do the same thing using |>.

Summary statistics table

First, load the model summary package

library(modelsummary)

Summary statistics table

Now we can use the datasummary_skim function.

datasummary_skim(diamonds)

| | Unique (#) | Missing (%) | Mean | SD | Min | Median | Max | |
|-------|------------|-------------|--------|--------|-------|--------|---------|------------|
| carat | 273 | 0 | 0.8 | 0.5 | 0.2 | 0.7 | 5.0 | - |
| depth | 184 | 0 | 61.7 | 1.4 | 43.0 | 61.8 | 79.0 | |
| table | 127 | 0 | 57.5 | 2.2 | 43.0 | 57.0 | 95.0 | |
| price | 11602 | 0 | 3932.8 | 3989.4 | 326.0 | 2401.0 | 18823.0 | L |
| X | 554 | 0 | 5.7 | 1.1 | 0.0 | 5.7 | 10.7 | . 🛌 |
| у | 552 | 0 | 5.7 | 1.1 | 0.0 | 5.7 | 58.9 | 4 |
| Z | 375 | 0 | 3.5 | 0.7 | 0.0 | 3.5 | 31.8 | L . |

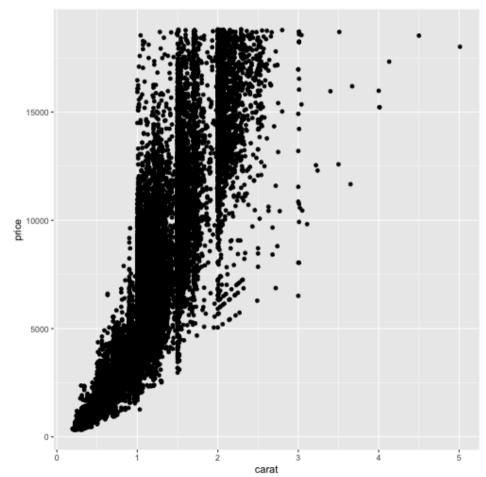
Summary statistics table

But not every variable is numeric. datasummary has this covered with the type = "categorical" sub-option.

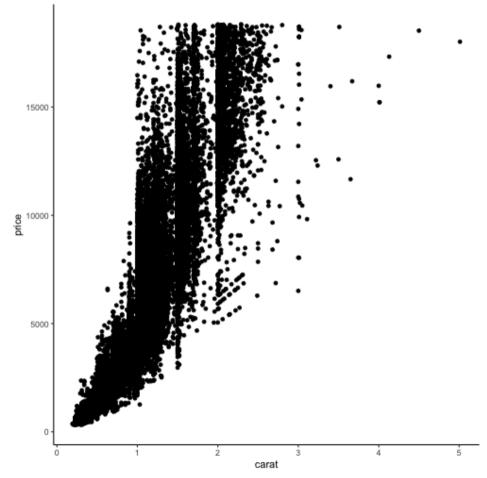
datasummary_skim(diamonds, type = "categorical")

| | | N | % |
|-------|-----------|-------|------|
| cut | Fair | 1610 | 3.0 |
| | Good | 4906 | 9.1 |
| | Very Good | 12082 | 22.4 |
| | Premium | 13791 | 25.6 |
| | Ideal | 21551 | 40.0 |
| color | D | 6775 | 12.6 |
| | E | 9797 | 18.2 |
| | F | 9542 | 17.7 |

```
ggplot(data = diamonds, aes(y = price, x =
    geom_point()
```

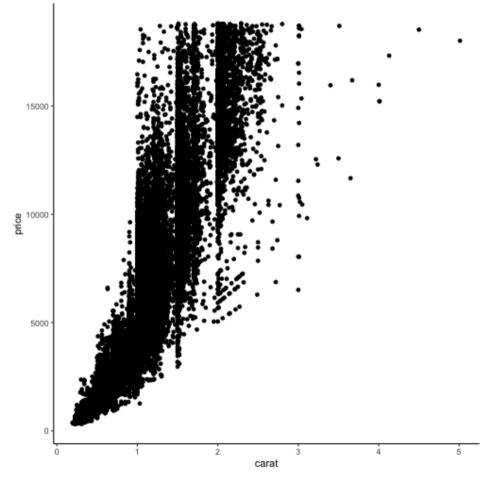


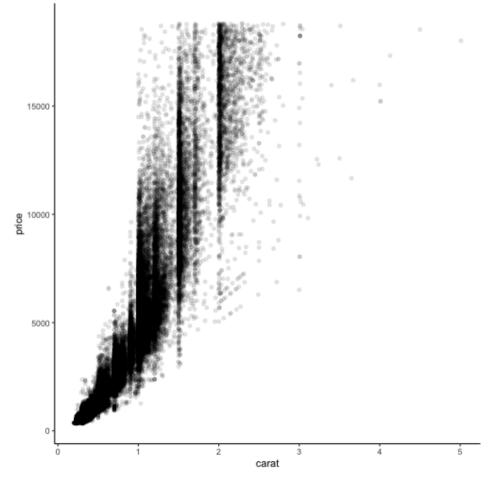
```
ggplot(data = diamonds, aes(y = price, x =
    geom_point() +
    theme_classic()
```



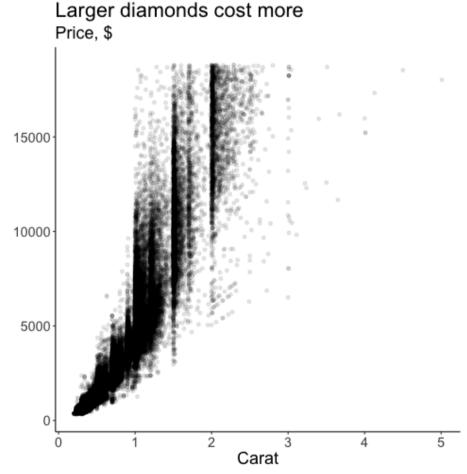
```
base_plot = ggplot(data = diamonds, aes(y =
     geom_point()

base_plot +
   {{theme_classic()}}
```

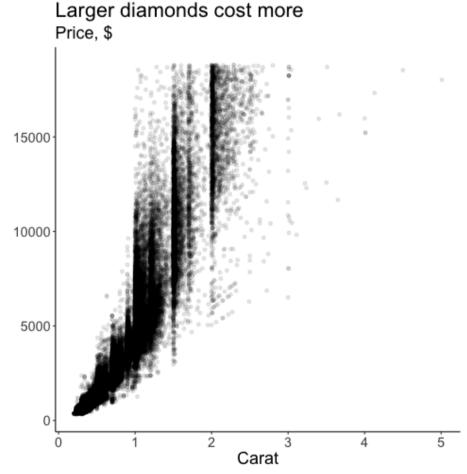




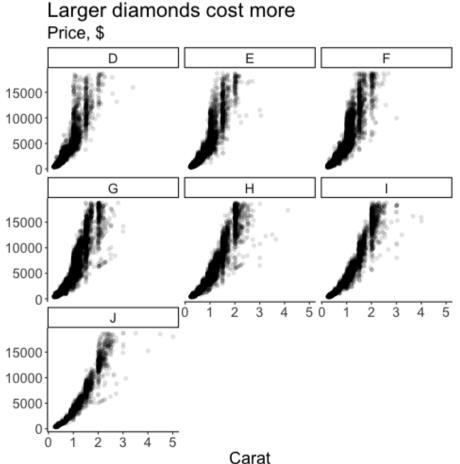
```
ggplot(data = diamonds, aes(y = price, x =
    geom_point(alpha = .1) +
    theme_classic() +
    theme(text = element_text(size = 18)) +
    labs(title = "Larger diamonds cost more
        subtitle = "Price, $",
        y = "",
        x = "Carat")
```



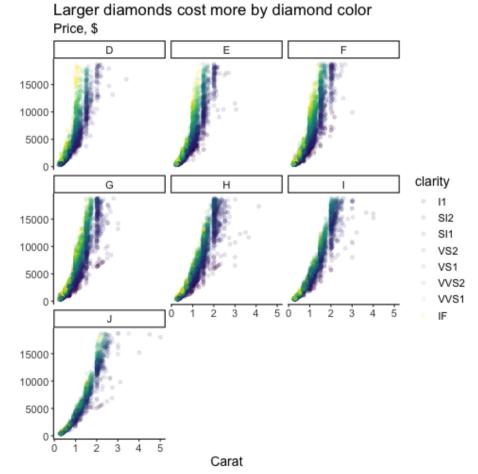
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ggplot(data = diamonds, aes(y = price, x =
    geom_point(alpha = .1) +
    theme_classic() +
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    labs(title = "Larger diamonds cost more
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        y = "",
        x = "Carat")
```

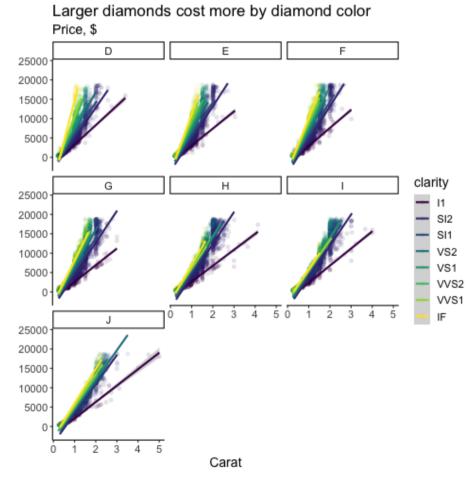


```
ggplot(data = diamonds, aes(y = price, x =
    geom_point(alpha = .1) +
    facet_wrap(~color) +
    theme_classic() +
    theme(text = element_text(size = 18)) +
    labs(title = "Larger diamonds cost more
        subtitle = "Price, $",
        y = "",
        x = "Carat")
```

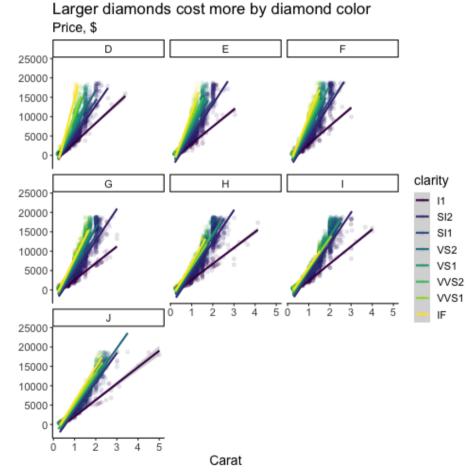


```
ggplot(data = diamonds, aes(y = price, x =
    geom_point(alpha = .1) +
    facet_wrap(~color) +
    theme_classic() +
    theme(text = element_text(size = 14)) +
    labs(title = "Larger diamonds cost more
        subtitle = "Price, $",
        y = "",
        x = "Carat")
```

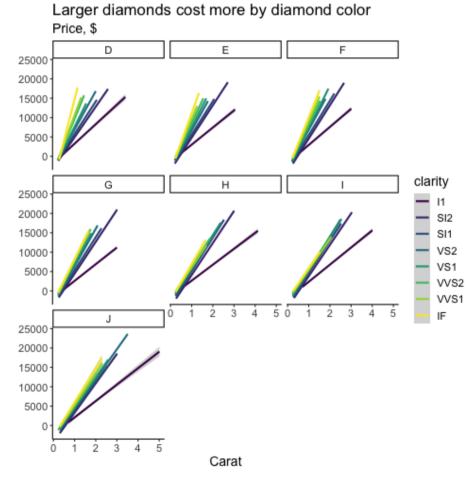




```
ggplot(data = diamonds, aes(y = price, x =
    geom_point(alpha = .1) +
    facet_wrap(~color) +
    geom_smooth(method = "lm") +
    theme_classic() +
    theme(text = element_text(size = 14)) +
    labs(title = "Larger diamonds cost more
        subtitle = "Price, $",
        y = "",
        x = "Carat")
```



```
ggplot(data = diamonds, aes(y = price, x =
    facet_wrap(~color) +
    geom_smooth(method = "lm") +
    theme_classic() +
    theme(text = element_text(size = 14)) +
    labs(title = "Larger diamonds cost more
        subtitle = "Price, $",
        y = "",
        x = "Carat")
```



Let's switch over to an R-Markdown document

Also look. Some latex

$$f(x) = rac{1}{\sigma\sqrt{2\pi}} \mathrm{exp}igg(-rac{1}{2}igg(rac{x-\mu}{\sigma}igg)^2igg)$$