# 101\_wk4\_Adv\_Vis

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### DS4B 101-R: R FOR BUSINESS ANALYSIS —-

### ADVANCED BUSINESS PLOTS —-

```
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.1 --
## v ggplot2 3.3.5 v purrr 0.3.4
## v tibble 3.1.6 v dplyr 1.0.7
## v tidyr 1.1.4 v stringr 1.4.0
## v readr 1.4.0
                   v forcats 0.5.1
## Warning: package 'tibble' was built under R version 4.1.1
## Warning: package 'tidyr' was built under R version 4.1.1
## Warning: package 'forcats' was built under R version 4.1.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                 masks stats::lag()
library(lubridate)
## Warning: package 'lubridate' was built under R version 4.1.1
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
##
      date, intersect, setdiff, union
```

```
library(tidyquant)
## Loading required package: PerformanceAnalytics
## Loading required package: xts
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
      as.Date, as.Date.numeric
##
## Attaching package: 'xts'
## The following objects are masked from 'package:dplyr':
##
##
      first, last
##
## Attaching package: 'PerformanceAnalytics'
## The following object is masked from 'package:graphics':
##
##
      legend
## Loading required package: quantmod
## Loading required package: TTR
## Registered S3 method overwritten by 'quantmod':
##
    method
                     from
##
    as.zoo.data.frame zoo
## Business Science offers a 1-hour course - Learning Lab #9: Performance Analysis & Portfolio Optimiza
## </> Learn more at: https://university.business-science.io/p/learning-labs-pro </>
library(formattable)
options(digits = 2, scipen = 99)
bike_orderlines_tbl <- read_rds("~/Desktop/University_business_science/DS4B_101/00_data/bike_sales/data
glimpse(bike_orderlines_tbl)
```

```
## Rows: 15,644
## Columns: 13
                   <dttm> 2011-01-07, 2011-01-07, 2011-01-10, 2011-01-10, 2011-0~
## $ order date
                   <dbl> 1, 1, 2, 2, 3, 3, 3, 3, 4, 5, 5, 5, 5, 6, 6, 6, 6, 7~
## $ order_id
## $ order_line
                   <dbl> 1, 2, 1, 2, 1, 2, 3, 4, 5, 1, 1, 2, 3, 4, 1, 2, 3, 4, 1~
## $ quantity
                   <dbl> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1
## $ price
                   <dbl> 6070, 5970, 2770, 5970, 10660, 3200, 12790, 5330, 1570,~
                   <dbl> 6070, 5970, 2770, 5970, 10660, 3200, 12790, 5330, 1570,~
## $ total_price
                   <chr> "Jekyll Carbon 2", "Trigger Carbon 2", "Beast of the Ea~
## $ model
                   <chr> "Mountain", "Mountain", "Mountain", "Road",~
## $ category_1
## $ category_2
                   <chr> "Over Mountain", "Over Mountain", "Trail", "Over Mounta~
## $ frame_material <chr> "Carbon", "Carbon", "Aluminum", "Carbon", "Carbon", "Ca-
## $ bikeshop_name <chr> "Ithaca Mountain Climbers", "Ithaca Mountain Climbers",~
                   <chr> "Ithaca", "Ithaca", "Kansas City", "Kansas City", "Loui~
## $ city
## $ state
                   <chr> "NY", "NY", "KS", "KS", "KY", "KY", "KY", "KY", "KY", "~
```

## 1.0 Lollipop Chart: Top N Customers

• — Great for showing order

Question: How much purchasing power is in top 5 customers? Goal: Visualize top N customers in terms of Revenue, include cumulative percentage

### **Data Manipulation**

```
n <- 10
top_customers_table <- bike_orderlines_tbl %>%
    select(bikeshop_name, total_price) %>%
    mutate(bikeshop_name = as_factor(bikeshop_name) %% fct_lump(n = n, w = total_price)) %>%
    group_by(bikeshop_name) %>%
    summarise(revenue = sum(total_price)) %>% ungroup() %>%
   mutate(bikeshop_name = bikeshop_name %>% fct_reorder(revenue)) %>%
   mutate(bikeshop_name = bikeshop_name %>% fct_relevel("Other", after = 0)) %>%
   arrange(desc(bikeshop_name)) %>%
    # Revenue Text
   mutate(revenue_text = scales::dollar(revenue, scale = 1e-6, suffix = "M")) %>%
    # Cumulative Percent
   mutate(cum pct = cumsum(revenue)/sum(revenue)) %>%
   mutate(cum_pct_text = scales::percent(cum_pct)) %>%
    # Rank
    mutate(rank = row_number()) %>%
    # NA_integer_: NA Values must match the data type. The NA_integer_ adds an NA value to integer data
    mutate(rank = case_when(
        rank == max(rank) ~ NA_integer_,
        TRUE ~ rank
       )) %>%
```

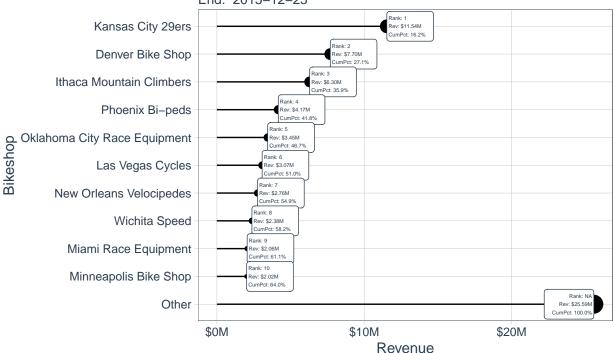
```
#Label Text
   mutate(label_text = str_glue("Rank: {rank}\nRev: {revenue_text}\nCumPct: {cum_pct_text}"))
top customers table
## # A tibble: 11 x 7
     bikeshop_name revenue_text cum_pct cum_pct_text rank label_text
##
##
     <fct>
                      <dbl> <chr>
                                          <dbl> <chr>
                                                          <int> <glue>
## 1 Kansas City 29~ 1.15e7 $11.54M
                                          0.162 16.2%
                                                                1 "Rank: 1\nRe~
## 2 Denver Bike Sh~ 7.70e6 $7.70M
                                          0.271 27.1%
                                                                2 "Rank: 2\nRe~
## 3 Ithaca Mountai~ 6.30e6 $6.30M
                                          0.359 35.9%
                                                                3 "Rank: 3\nRe~
## 4 Phoenix Bi-peds 4.17e6 $4.17M
                                          0.418 41.8%
                                                                4 "Rank: 4\nRe~
## 5 Oklahoma City ~ 3.45e6 $3.45M
                                          0.467 46.7%
                                                               5 "Rank: 5\nRe~
## 6 Las Vegas Cycl~ 3.07e6 $3.07M
                                          0.510 51.0%
                                                                6 "Rank: 6\nRe~
## 7 New Orleans Ve~ 2.76e6 $2.76M
                                          0.549 54.9%
                                                               7 "Rank: 7\nRe~
                                                               8 "Rank: 8\nRe~
## 8 Wichita Speed
                     2.38e6 $2.38M
                                          0.582 58.2%
                                          0.611 61.1%
                                                               9 "Rank: 9\nRe~
## 9 Miami Race Equ~ 2.06e6 $2.06M
## 10 Minneapolis Bi~ 2.02e6 $2.02M
                                          0.640 64.0%
                                                               10 "Rank: 10\nR~
## 11 Other
                     2.56e7 $25.59M
                                          1
                                                100.0%
                                                             NA "Rank: NA\nR~
```

#### Data Visualization

```
top customers table %>%
    # Geometrics
   ggplot(aes(revenue, bikeshop_name)) +
    geom_segment(aes(xend = 0, yend = bikeshop_name), size = 0.5) +
   geom_point(aes(size = revenue)) +
    # Label
    geom_label(aes(label = label_text),
               size = 1.5,
               hjust = "inward",
               colour = palette_light()[1]) +
    # Formatting
    scale_x_continuous(labels = scales::dollar_format(scale = 1e-6, suffix = "M")) +
    labs(
        title = str_glue("top {n} Customers"),
        subtitle = str_glue("Start: {min(bike_orderlines_tbl$order_date)}
                             End: {max(bike_orderlines_tbl$order_date)}"),
        x = "Revenue".
        y = "Bikeshop",
        caption = str_glue("Top 6 customers contribute 51% of purchasing power.")
   ) +
   theme_tq() +
    theme(
        legend.position = "none",
        plot.title = element_text(face = "bold"),
       plot.caption = element_text(face = "bold.italic")
```

### top 10 Customers

Start: 2011-01-07 End: 2015-12-25



Top 6 customers contribute 51% of purchasing power.

# 2.0 Heatmaps

• — Great for showing details in 3 dimensions

Question: Do specific customers have a purchasing prefernce?

Goal: Visualize heatmap of proportion of sales by Secondary Product Category

### **Data Manipulation**

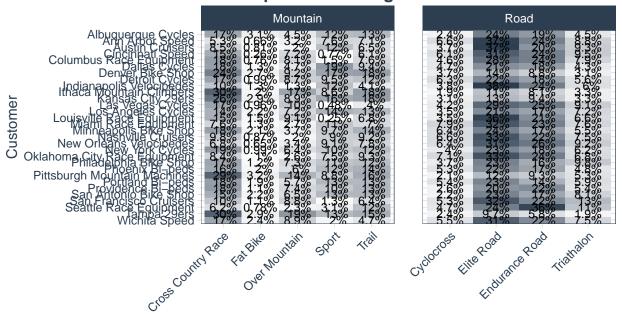
```
pct_sales_by_customer <- bike_orderlines_tbl %>%
    select(bikeshop_name, category_1, category_2, quantity) %>%
    group_by(bikeshop_name, category_1, category_2) %>%
    summarise(total_qty = sum(quantity)) %>% ungroup() %>%
    group_by(bikeshop_name) %>%
    mutate(pct = total_qty/sum(total_qty)) %>% ungroup() %>%
    mutate(bikeshop_name = as_factor(bikeshop_name) %>% fct_rev()) %>%
    mutate(bikeshop_name_num = as.numeric(bikeshop_name))
```

## 'summarise()' has grouped output by 'bikeshop\_name', 'category\_1'. You can override using the '.grou

## **Data Visualization**

```
pct_sales_by_customer %>%
    ggplot(aes(category_2, bikeshop_name)) +
    geom_tile(aes(fill = pct)) +
    geom_text(aes(label = scales::percent(formattable(pct, digits =2))), size = 3) +
    facet_wrap(~category_1, scales = "free_x") +
    # formatting
    scale_fill_gradient(low = "white", high = palette_light()[1]) +
    labs(
        title = "Heatmap of Purchasing Habits",
       x = "Bike Type (Category 2)",
       y = "Customer",
       caption = str_glue(
        "Customers that prefer Road:
        Ann Arbor speed, Austin Cruisers, & Indianapolis Velocipedes
       Customers that prefer Mountain:
        Ithaca Mountain Climbers, Pittsburgh Mountain Machines, & Tampa 29ers")
    ) +
    theme_tq() +
    theme(
       axis.text.x = element_text(angle = 45, hjust = 1),
       legend.position = "none",
       plot.title = element_text(face = "bold"),
       plot.caption = element_text(face = "bold.italic")
```

# **Heatmap of Purchasing Habits**



Bike Type (Category 2)

Customers that prefer Road: Ann Arbor speed, Austin Cruisers, & Indianapolis Velocipedes

Customers that prefer Mountain: Ithaca Mountain Climbers, Pittsburgh Mountain Machines, & Tampa 29ers