Data Analysis on Bike Store

Introduction: Welcome to the dynamic world of a Bike Store in USA, where the thrill of cycling meets the precision of data. In this exciting chapter of our journey, we delve into the intricate realm of data analysis to understand and optimize every aspect of our bike store's operations. Join us on this expedition as we navigate the data trail, unlocking insights that propel us forward in delivering an unparalleled cycling experience.

Data-Driven Pedals: our commitment to excellence extends beyond the workshop floor and into the realm of data analytics. By harnessing the power of data, we aim to fine-tune our operations, enhance customer experiences, and elevate our understanding of the ever-evolving cycling landscape.

Exploring the Analytics Landscape: Just as cyclists explore diverse terrains, we explore the vast landscape of data analytics. From sales trends to customer behavior, inventory management to marketing effectiveness, our data analysis project is a comprehensive expedition that aims to uncover patterns, optimize strategies, and ensure that every gear shift in our store is backed by informed decisions.

Your Ride, Your Data: In this data-driven era, we understand the importance of personalization. Our data analysis project isn't just about numbers; it's about tailoring the cycling experience to your preferences. By analyzing your interactions, purchases, and feedback, we strive to create a more customized and enjoyable journey for every cyclist who steps into the Bike Store.

Joining Tables Using MySQL

Introduction: In the realm of data analysis for our bike sales project, SQL joins emerge as the crucial gears that seamlessly connect disparate datasets, enabling us to extract valuable insights from the intricate tapestry of information. This brief explores how SQL joins elevate our capacity to understand, analyze, and enhance the performance of our bike sales at [Your Bike Store].

Understanding the Terrain: Our bike sales data spans various dimensions — from customer information and product details to transaction records and inventory status. SQL joins act as the navigational tool, allowing us to traverse this multidimensional landscape with precision and clarity.

Accelerating Analysis:

- Efficiency: SQL joins streamline the process, allowing us to retrieve relevant information from multiple tables in a single query.
- Accuracy: By establishing relationships between tables, we ensure that our analyses are based on comprehensive and interconnected datasets.
- Insight Generation: SQL joins empower us to unveil intricate patterns, correlations, and trends that drive informed decision-making in product placement, marketing strategies, and inventory management.

MySQL Query:

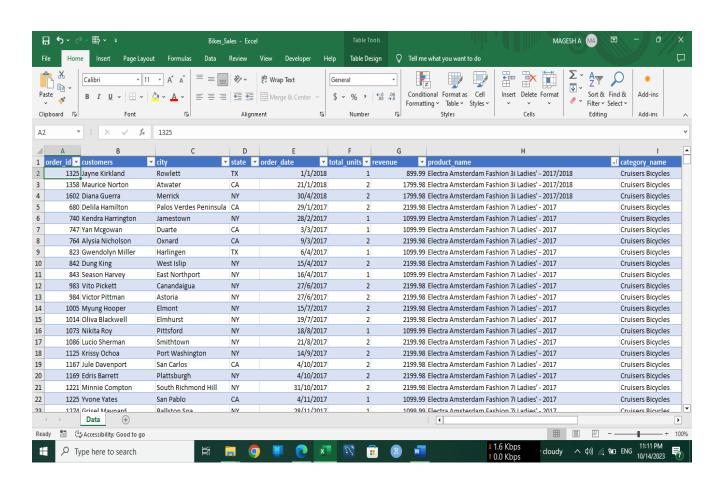
```
SELECT
      ord.order id,
      CONCAT(cus.first_name," ",cus.last_name) AS "customer",
      cus.city,
      cus.state,
      SUM(ite.quantity) AS "total_units",
      SUM(ite.quantity * ite.list_price) AS "revenue",
      pro.product_name,
      cat.category_name,
      sto.store_name,
      CONCAT(sta.first_name , " " , sta.last_name) AS "Sales_rep"
FROM orders ord
   JOIN customers cus
      ON ord.customer id = cus.customer id
   JOIN order_items ite
      ON ord.order_id = ite.order_id
   JOIN products pro
      ON ite.product id = pro.product id
   JOIN categories cat
      ON pro.category_id = cat.category_id
   JOIN stores sto
      ON ord.store_id = sto.store_id
   JOIN staffs sta
      ON ord.staff id = sta.staff id
 GROUP BY
         ord.order id,
         CONCAT(cus.first_name , " " , cus.last_name),
         cus.city,
         cus.state,
         ord.order date,
         pro.product_name,
         cat.category name,
         CONCAT(sta.first_name , " " , sta.last_name)
```

Data cleaning

Introduction: In the world of data analysis, the importance of clean, well-organized data cannot be overstated. Excel, a powerhouse in spreadsheet software, provides robust tools for data cleaning, allowing analysts to transform raw datasets into refined gems. This brief explores the essential techniques and tools within Excel for effective data cleaning.

Identifying Data Issues: Before diving into the cleaning process, it's crucial to identify common data issues such as missing values, duplicates, inconsistencies, and formatting errors. Excel's data cleaning features are designed to address these challenges and enhance the overall quality of your dataset.

Dataset -> click here



Analyzing using R

Installing Packages

```
install.packages("dplyr")
install.packages("tidyverse")
install.packages("tidyr")
install.packages("ggplot2")
install.packages("rmarkdown")
install.packages('tinytex')
library(dplyr)
library(tidyverse)
library(tidyverse)
library(ggplot2)
library(gmarkdown)
library(tinytex)
```

Reading input

```
Bikes_Sales <-read_excel("Bikes_Sales.xlsx")
```

Column names

head()

```
head(Bikes Sales)
```

```
## # A tibble: 6 x 12
                             city state order date
                                                              total units revenue
## order id customers
        <dbl> <chr>
                               <chr> <chr> <dttm>
                                                                    <dbl> <dbl>
        1325 Jayne Kirkland
## 1
                               Rowl... TX
                                           2018-01-01 00:00:00
                                                                       1
                                                                            900.
        1358 Maurice Norton
## 2
                               Atwa... CA
                                           2018-01-21 00:00:00
                                                                        2 1800.
## 3
        1602 Diana Guerra
                               Merr... NY
                                          2018-04-30 00:00:00
                                                                        2
                                                                           1800.
## 4
         680 Delila Hamilton Palo... CA
                                                                        2
                                                                           2200.
                                          2017-01-29 00:00:00
## 5
         740 Kendra Harrington Jame... NY
                                          2017-02-28 00:00:00
                                                                       1 1100.
         747 Yan Mcgowan
                               Duar... CA
## 6
                                           2017-03-03 00:00:00
                                                                           1100.
## # i 5 more variables: product_name <chr>, category_name <chr>,
      brand_name <chr>, store_name <chr>, sales_rep <chr>
```

str()

```
str(Bikes Sales)
```

```
## tibble [4,722 x 12] (S3: tbl df/tbl/data.frame)
## $ order id : num [1:4722] 1325 1358 1602 680 740 ...
## $ customers : chr [1:4722] "Jayne Kirkland" "Maurice Norton" "Diana Guerra" "Delila Hamilton" ...
## $ city
                : chr [1:4722] "Rowlett" "Atwater" "Merrick" "Palos Verdes Peninsula" ...
## $ state : chr [1:4722] "TX" "CA" "NY" "CA" ...
## $ order date : POSIXct[1:4722], format: "2018-01-01" "2018-01-21" ...
## $ total units : num [1:4722] 1 2 2 2 1 1 2 1 2 1 ...
## $ revenue
                 : num [1:4722] 900 1800 1800 2200 1100 ...
## $ product name : chr [1:4722] "Electra Amsterdam Fashion 3i Ladies' - 2017/2018" "Electra Amsterdam Fashion 3i Ladies' -
2017/2018" "Electra Amsterdam Fashion 3i Ladies' - 2017/2018" "Electra Amsterdam Fashion 7i Ladies' - 2017" ...
## $ category name: chr [1:4722] "Cruisers Bicycles" "Cruisers Bicycles" "Cruisers Bicycles" ...
## $ brand name : chr [1:4722] "Electra" "Electra" "Electra" "...
## $ store_name : chr [1:4722] "Rowlett Bikes" "Santa Cruz Bikes" "Baldwin Bikes" "Santa Cruz Bikes" ...
## $ sales rep : chr [1:4722] "Kali Vargas" "Mireya Copeland" "Marcelene Boyer" "Genna Serrano" ...
```

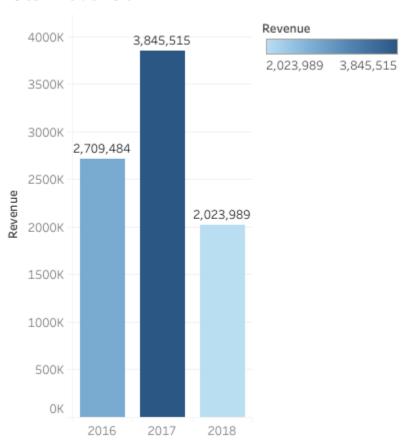
summary()

```
summary(Bikes_Sales)
```

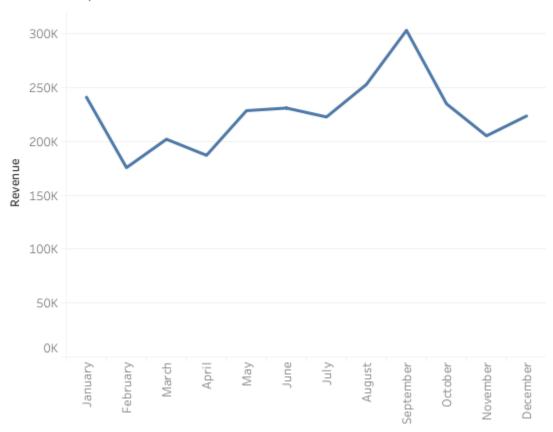
```
order id
                    customers
                                         city
                                                          state
                   Length: 4722
                                     Length:4722
                                                       Length: 4722
## Min. : 1.0
## 1st Qu.: 423.2
                   Class :character
                                     Class :character
                                                       Class : character
                   Mode :character
## Median: 828.5
                                     Mode :character
                                                       Mode :character
   Mean : 821.3
   3rd Qu.:1226.0
   Max. :1615.0
     order date
                                   total units
                                                     revenue
   Min. :2016-01-01 00:00:00.00
                                  Min. :1.000
                                                 Min. : 89.99
                                  1st Qu.:1.000
   1st Qu.:2016-09-06 12:00:00.00
                                                  1st Qu.: 539.98
   Median :2017-04-09 00:00:00.00
                                  Median :1.000
                                                  Median: 939.98
                                  Mean :1.499 Mean : 1816.81
   Mean :2017-04-04 22:18:26.99
   3rd Qu.:2017-11-04 00:00:00.00
                                  3rd Qu.:2.000 3rd Qu.: 1999.98
         :2018-12-28 00:00:00.00
                                  Max.
                                        :2.000 Max.
                                                       :23999.98
   product name
                                        brand name
                     category name
                                                          store name
                     Length: 4722
   Length:4722
                                       Length: 4722
                                                         Length:4722
   Class :character Class :character
                                       Class : character Class : character
   Mode :character Mode :character
                                       Mode :character Mode :character
##
##
##
    sales rep
   Length:4722
   Class :character
   Mode :character
##
##
##
```

Data Viz

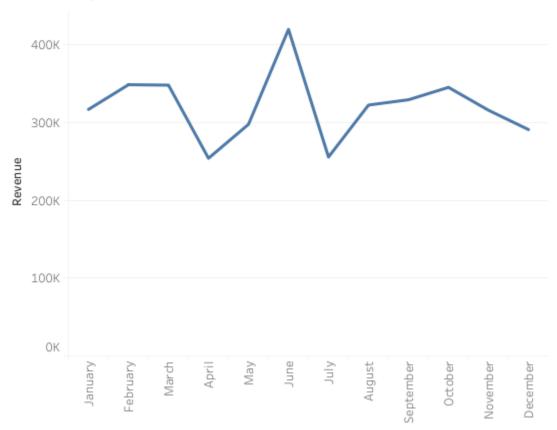
Total Revenue



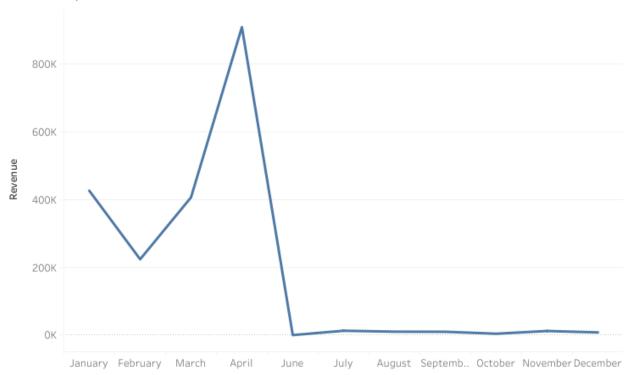
Revenue per month - 2016



Revenue per month - 2017



Revenue per month - 2018



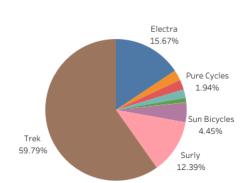
Revenue by State

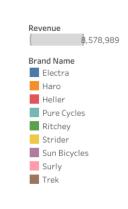


Revenue by category

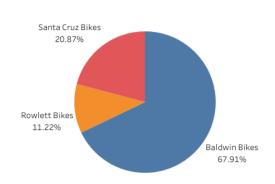
Mountain Bikes 35.33% 3,030,776	Cruisers Bicycles 12.93% 1,109,151	Electric Bikes 11.89% 1,020,237	327,888	3,030,776
Road Bikes 21.59% 1,852,556	Cyclocross Bicycles 9.32% 799,875	Comfort Bicycles 5.11% 438,507		

Revenue per Brand



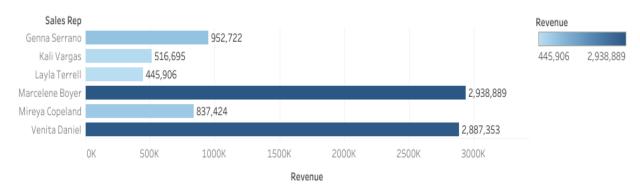


Revenue per Store

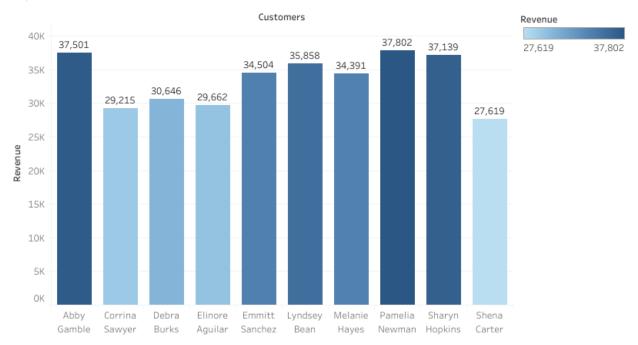




Top Sales Rep



Top Customers



Data Driven Decision Making:

- The Electric vehicles category is low and contributing major part in the revenue So, Increase the category of E – Vehicles.
- The number of stores of Rowlett Bikes is low and ratio of revenue of this store is marginally higher than other store So, Increase the number of stores.
- The number of stores in the California city is high and the revenue ratio is low So, reduce the number of the stores in California.
- Customer's are preferring Electra Brand the more than the other brand but their revenue ratio is considerably low So, launch new products from Electra.
- The number of times visiting the store by a customer is pretty low So, by giving give away and gift to the selected repeated customers will increase the revenue.
- By increasing the salary of the top sales representative create a boost among the sales rep.
- The revenue graph is not uniform, company need to work on the customer feedback and give what customer exactly needed.