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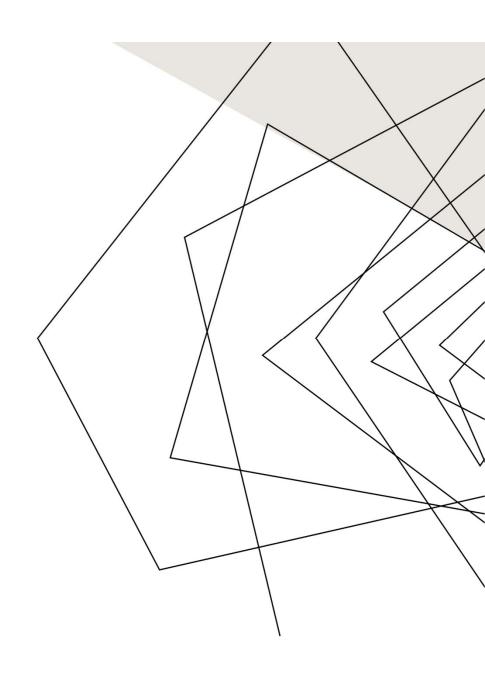
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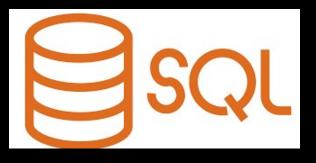
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# **SKILLSETS**









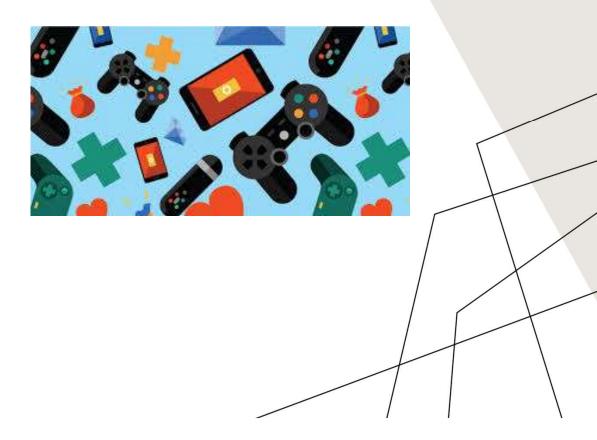




Project #1 – GameCo

Excel

Analyzing Global Video Game Sales by Region



### Goals

- Assess regional differences in video game sales over time
- Assess trends in regional markets
- Provide insight into GameCo's next year's marketing budget based on region

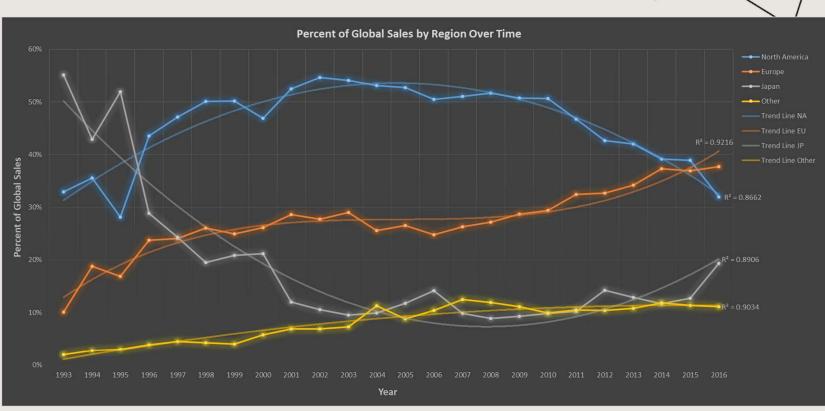
- ~16,000 game titles
- 1980 to 2016
- North America, European, Japanese, and Other regional sales categories measured in million units sold
- Data was cleaned, consolidated, and made uniform in Excel
- Pivot tables and graphs were used to derive key insights and conduct statistical analysis

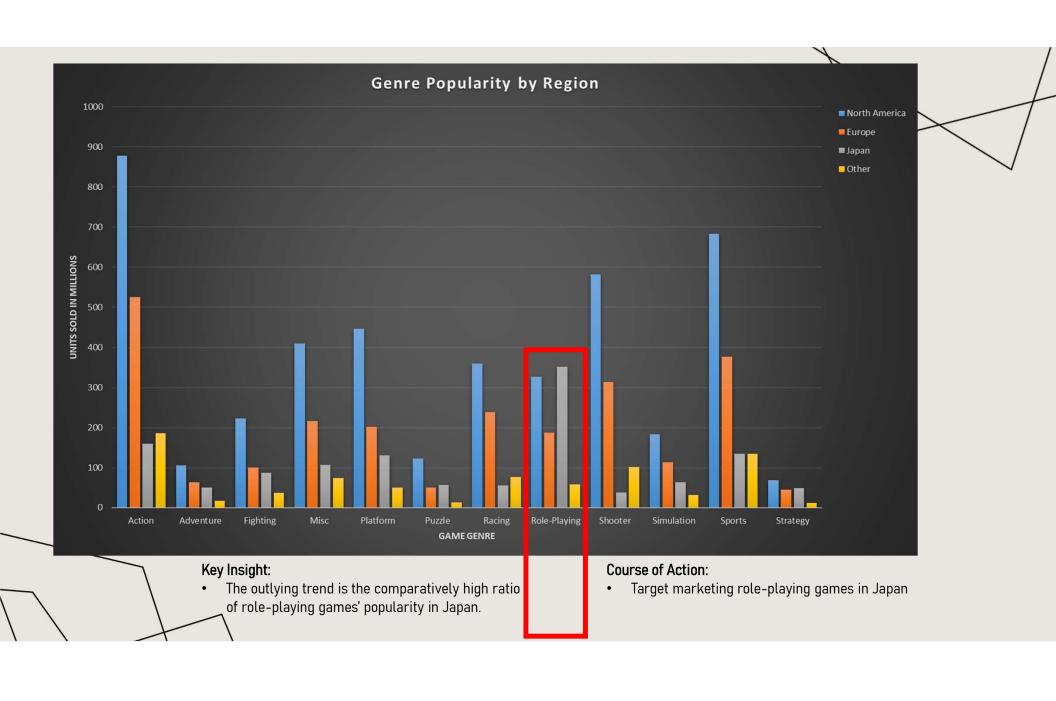
### Key Insights:

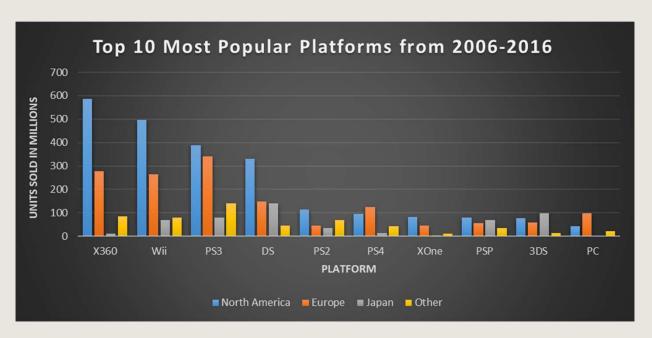
- Japanese and European sales are rising
- North American sales are declining

#### Course of Action:

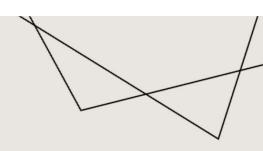
 Add resources to the North American budget depending on overall marketing strategy







Platform	North America	Europe	Japan	Other
X360	587	278	12	84
Wii	497	264	68	79
PS3	389	340	79	141
DS	331	148	141	44
PS2	115	46	36	69
PS4	97	124	14	43
XOne	83	46	0	12
PSP	80	56	70	34
3DS	78	58	97	12
PC	43	98	0	21



### Key Insights:

- X360, Wii, and PS3 are top selling platforms
- The DS sells particularly strong in North America

#### Course of Action:

• Focus marketing budget on top selling platforms

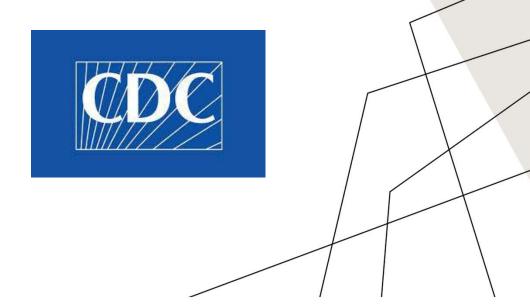
# Project #2 - Center for Disease Control (CDC)

# Excel/Tableau

Finding Insights for a Medical Staffing Agency to Prepare for Influenza Season in the United States

Link to Tableau Story

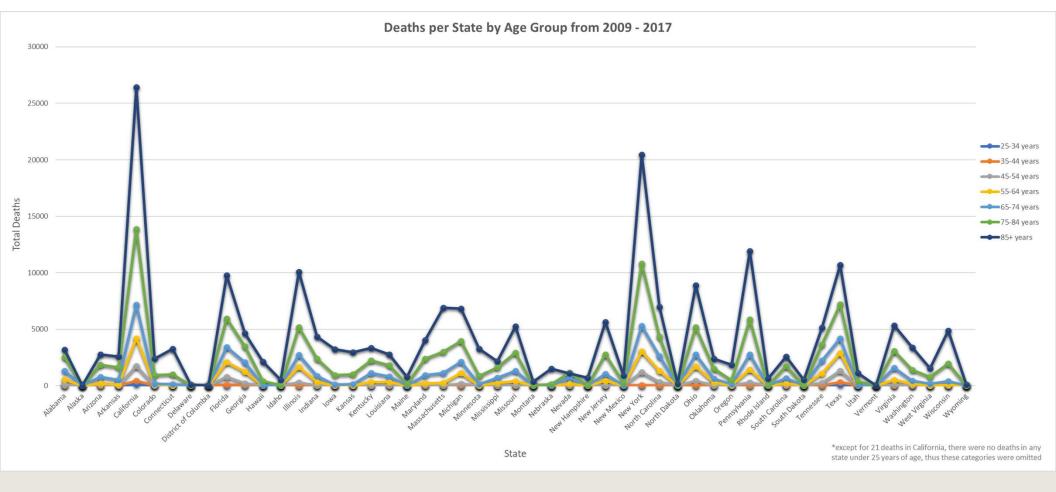
My Storytelling to the Problem



#### Goals

- Assess regional and seasonal differences in the flu's effects across the United States
- Assess trends in deaths from flu across time and region in the United States
- Provide insight into at risk populations and how they are affected
- Provide insight into where, when, and how much staff to send to each state

- USA Census Data
  - Drawn from the US Government
  - 2009–2017 all U.S. county populations broken down by age group and year
- CDC Influenza Deaths Data
  - Drawn from the CDC
  - 2009–2017 all U.S. deaths by flu broken down by state, age group, month, and year
  - Death records with values between 0-9 were "suppressed" so death counts of this data were brought to a value of "0"
- Excel tools and Tableau, including pivot tables and statistical analysis and charts, were used to clean and extrapolate meaning from the data



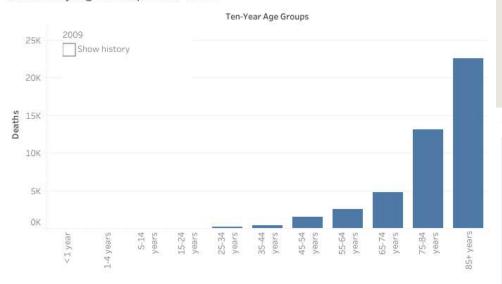
### Key Insights:

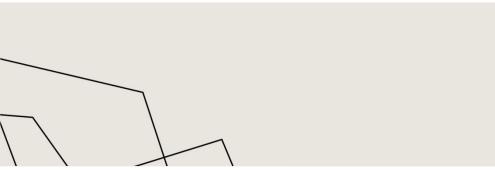
- Regardless of state, 65+ years of age is at much higher risk of death from influenza Course of Action:
- Send more resources to states with higher populations

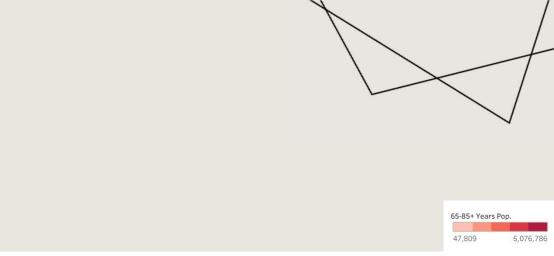
#### Preparing for Influenza Season

During influenza season there is high demand for additional medical staff to respond to patient needs across the United States. This analysis provides insight into when and where to send additional support to protect vulnerable populations (those 65+ years old) and minimize influenza's impact.

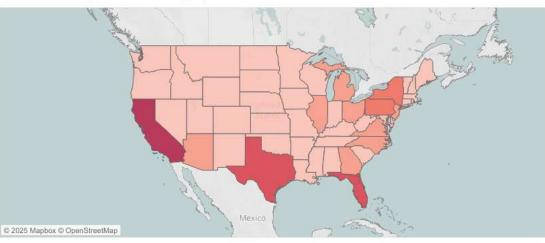
#### Deaths by Age Group 2009-2017







#### Vulnerable Population Demographics 2009-2017



The chart indicates the vulnerable population of 65+ years (as defined by the CDC) should be the priority for the upcoming flu season. Ages <64 are negligibly affected comparatively. The map indicates where this vulnerable population is predominantly located.

#### The Plan:

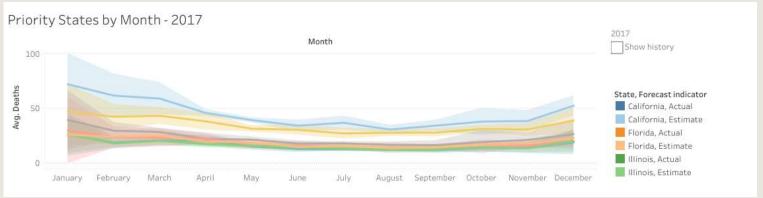
Based on 2017 populations, states can be prioritized based on vulnerable population size.

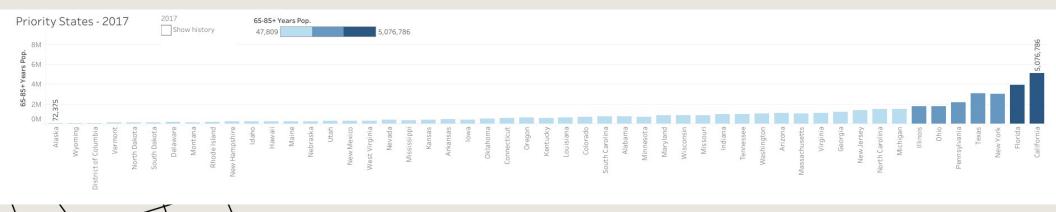
**High Priority** = California and Florida, **Medium Priority** = New York, Texas, Pennsylvania, Ohio, Illinois, **Low Priority** = All Others Priority months for all states are January, February, March, November, and December.

#### Limitations:

Due to data limitations, the exact amount of staff to send per patient is prevented at this time.



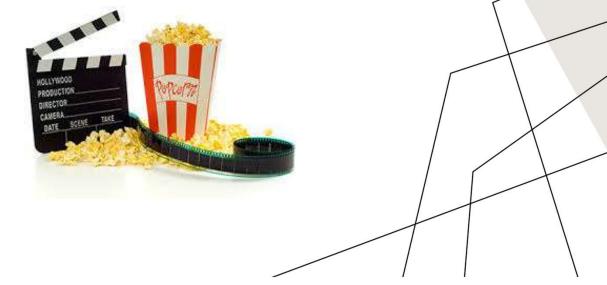




# Project #3 - Rockbuster Stealth LLC SQL/ Tableau/Data Warehousing

Finding Data-Driven Answers to Business Questions for an Online Video Rental Company

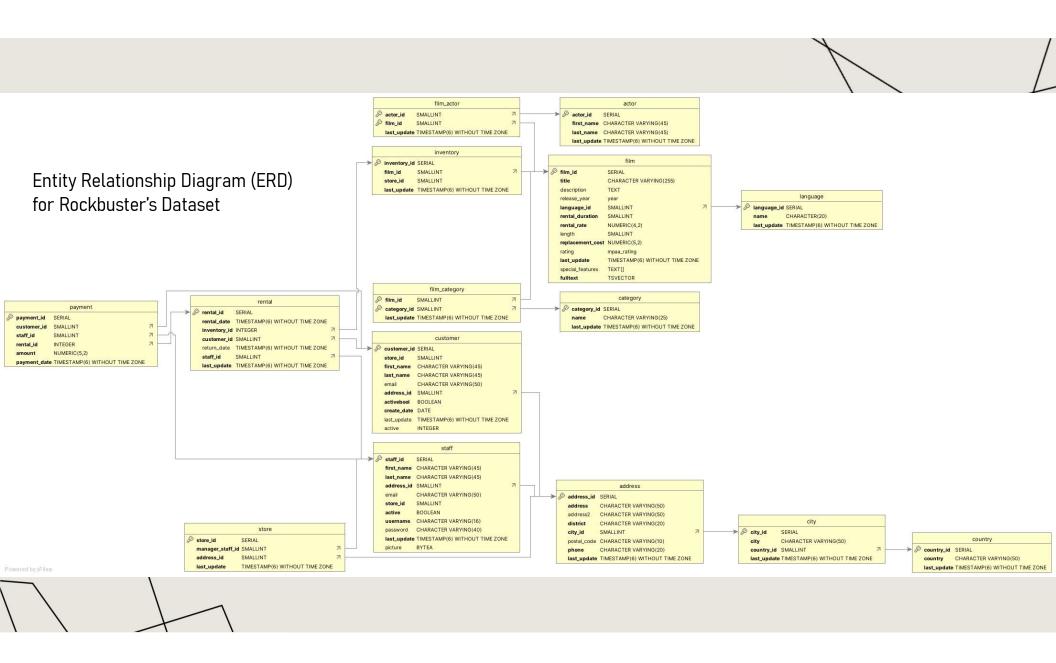
Link to GitHub Repository



### Goals

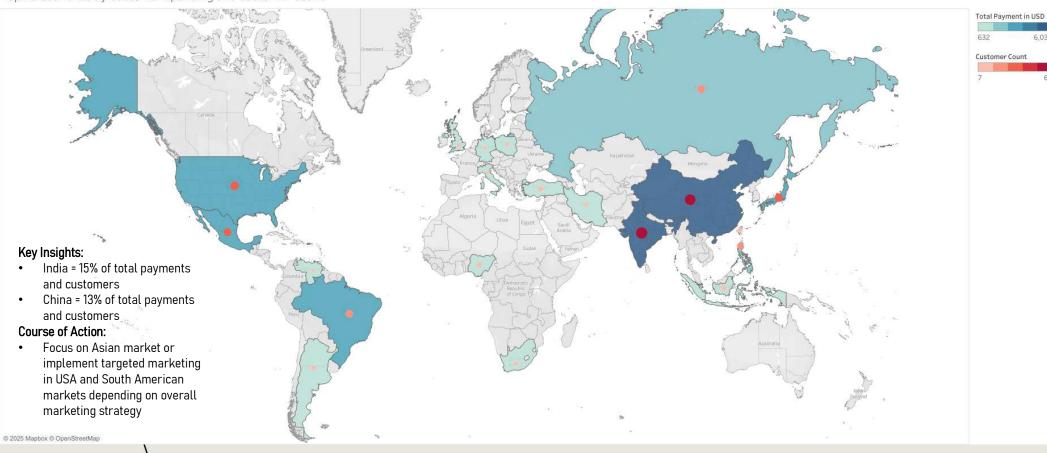
- Assess regional differences in sales and customers
- Assess most and least profitable movies
- Provide insight into customer rental trends

- Sales and customer data from Rockbuster
- Data was cleaned, formatted, and joined for statistical analysis and pattern identification
- Excel, Tableau, and SQL queries were used to clean, analyze, and provide insights into the data through visualizations





Top 20 Countries by Customer Spending and Customer Count





# Project #4 – Instacart

# Python/Anaconda/Jupyter

Finding Insights for a Data-Driven Marketing Strategy for an Online Grocery Store

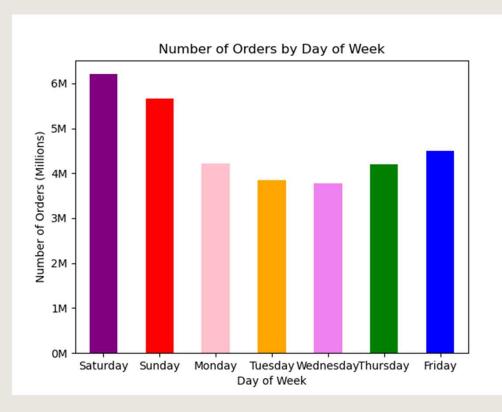
Link to GitHub Repository



#### Goals

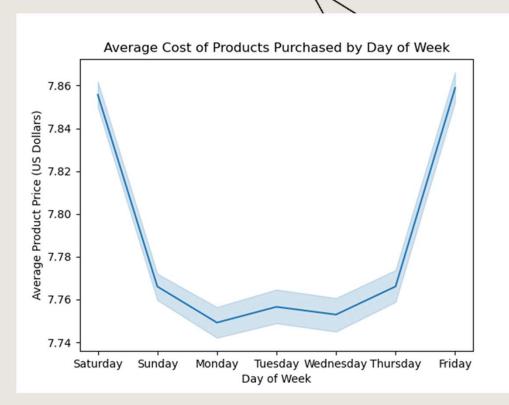
- Assess busy times of days and weeks based on traffic and money spent
- Assess effective price range groupings
- Provide insights into customer demographics

- Customer data set containing identifiable personal consumer information
- Order, products, and sales data from Instacart's 2017 sales year
- Excel and Anaconda/Jupyter were used to clean, merge, analyze, and provide insights into the data through visualizations
  - Python libraries including pandas, NumPy, os, matplotlib, scipy, and seaborn used to visualize and conduct statistical analysis on the data





- The highest volume of sales occurs on weekends. Course of Action:
- Initiate sales on weekdays to drive traffic

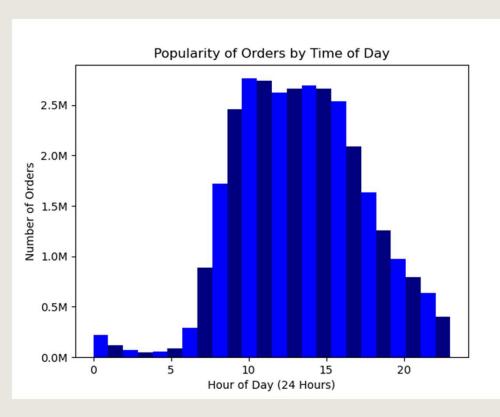


#### Key Insight:

 The ~12 cent difference from the low to the high end of the chart would indicate a noticeable difference in cost by day of the week, but may not be significant depending on financial goals.

#### Course of Action:

 May indicate people are more willing to pay for convenience on weekends, retain full prices on Fridays and Saturdays

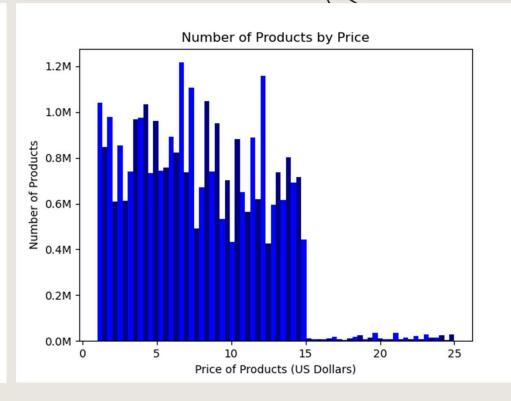




 9am to 5pm are when most orders are placed on Instacart.

#### Course of Action:

- Consider staffing more during busy times
- Push advertising on the shoulders of busy times to increase impulse buys

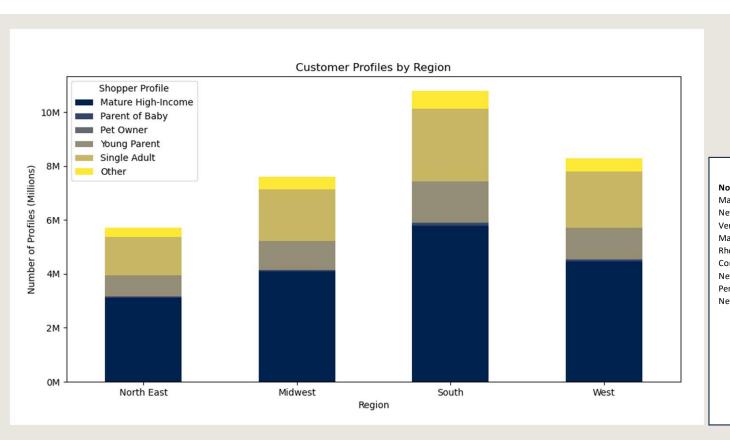


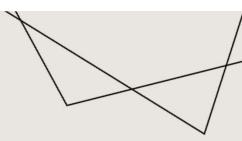
#### Key Insight:

 Products priced lower than \$15 sell in much higher volumes than those priced greater than or equal to \$15.

#### Course of Action:

 Products on cusp of \$15 dollars can be lowered below the threshold





orth East:	Midwest:		
aine	Wisconsin		
ew Hampshire	Michigan		
ermont	Illinois		
assachusetts	Indiana		
node Island	Ohio		
onnecticut	North Dakota		
ew York	South Dakota		
ennsylvania	Nebraska		
ew Jersey	Kansas		
	Minnesota		
	Iowa		
	Missouri		

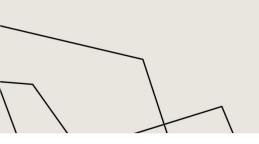
South: West: Delaware Idaho Maryland Montana Wyoming District of Columbia Virginia Nevada West Virginia Utah North Carolina Colorado South Carolina Arizona Georgia New Mexico Florida Alaska Washington Kentucky Tennessee Oregon Mississippi California Alabama Hawaii Oklahoma Texas Arkansas Louisiana

### Key Insights:

- Chart indicates shopper profiles by region.
- Shopper profile concentrations are consistent across regions.

#### Course of Action:

Regionally based marketing may not be necessary based on customer profile



# Project #5 - Pig E. Bank Excel

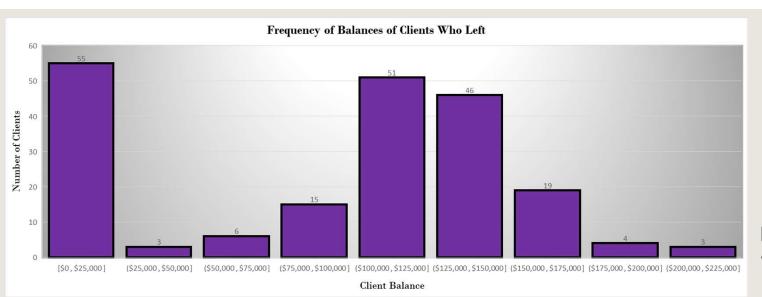
Assisting with Customer Projects at a Global Bank

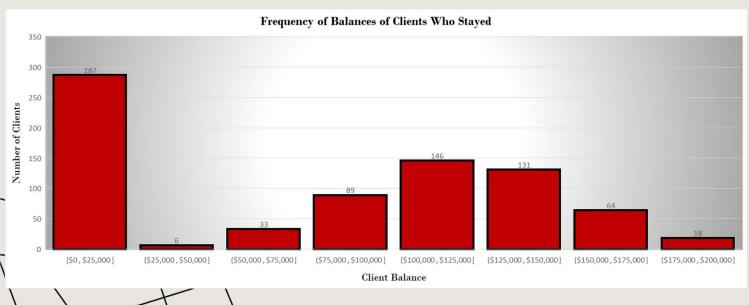


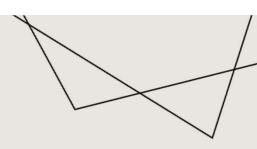
### Goals

 Identifying reasons why clients might leave or remain with the bank

- Data set provided by Pig E. Bank including client information of credit scores, country, gender, age, tenure, salary, and bank usage
- Excel was used to clean, anonymize, analyze, and provide insights into the data through visualizations, pivot tables, and statistical analysis





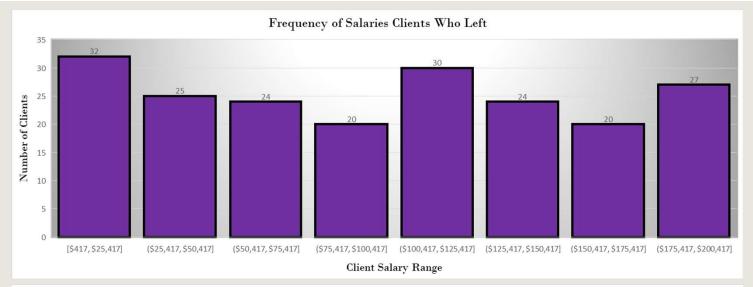


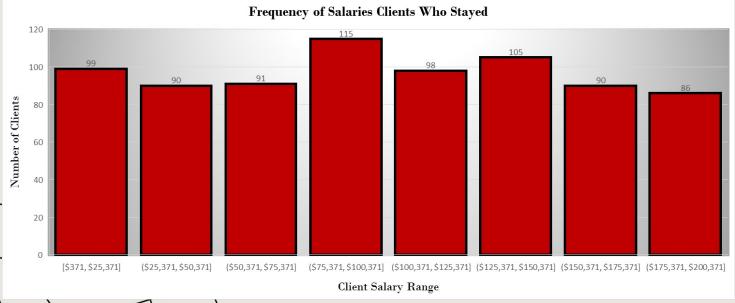
### Key Insights:

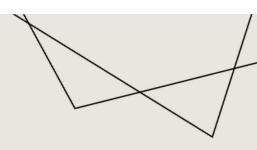
- These charts identify if client balances were an indication of staying with or leaving the bank.
- The similar shape of the data indicates it was not a factor.

#### Course of Action:

Other factors must be evaluated





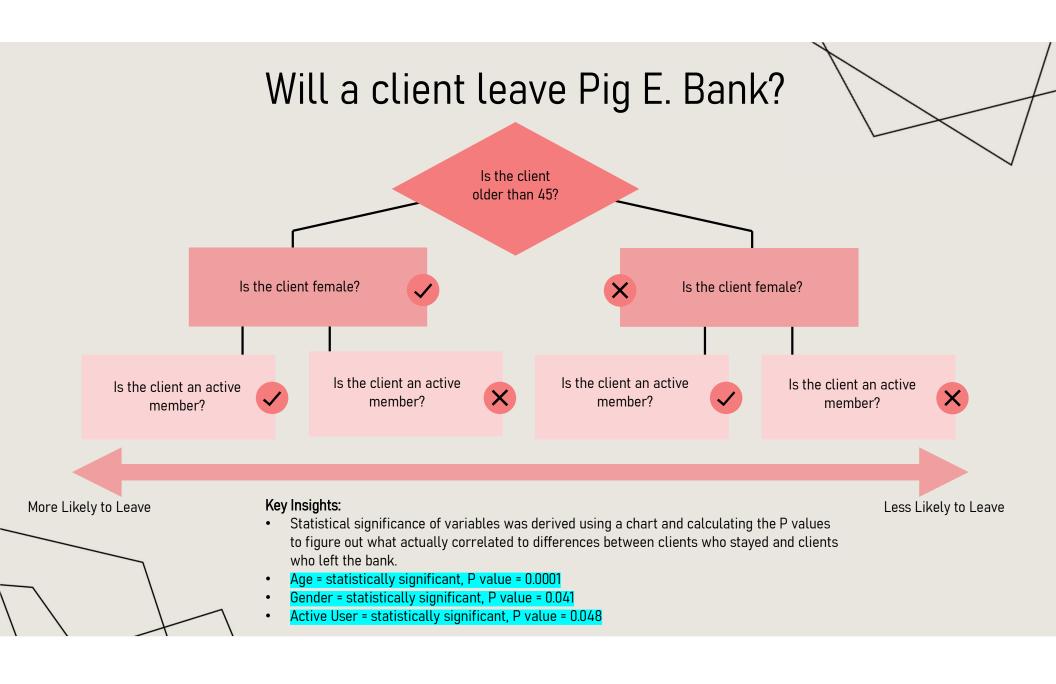


### Key Insights:

- These charts identify if client salaries were an indication of staying with or leaving the bank.
- The proportions are different enough to warrant need for further exploration.

### Course of Action:

 Widen the search to include more clients and their salaries, if possible, to evaluate if this is a significant factor



# Project #6 – USA Real Estate Market

# Python and Excel

Assessing Factors that Affect USA Housing Prices

Link to GitHub Repository

Link to Tableau Story

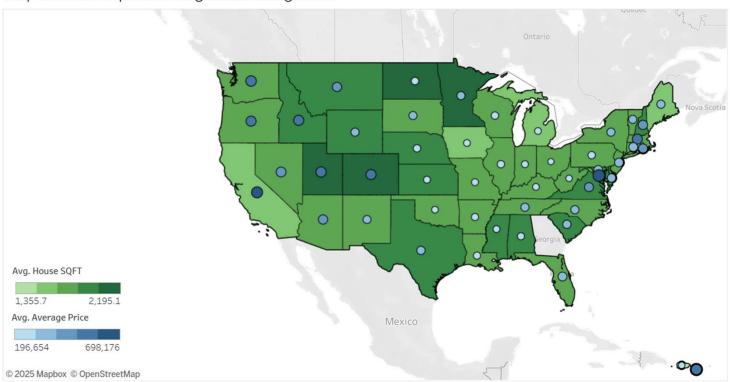


### Goals

- Identifying reasons why certain houses are more expensive than others
- Provide insights for sellers and buyers into the factors affecting their housing costs

- 2,226,382 data entries from across the USA and its territories including house price, # of beds, # of baths, acreage, square footage, city, state, and sell date
- Python was used to conduct cleaning, wrangling, exploratory analysis through visualizations, geospatial analysis, regression analysis, cluster analysis, time-series analysis, and analysis narrative and final results presented in a Tableau dashboard

### Map of House Square Footage and Average Price

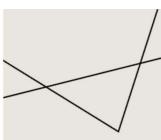


**Location, location.** We all know the age-old adage that location is the biggest factor driving real estate prices.

#### What if there is more to it?

This analysis attempts to find out if house features can be an equally important factor in predicting housing prices. Or is there another factor entirely worth looking into?

Using data from realtor.com (a real estate listing website that is the second most visited of its kind in the United States as of 2024, with over 100 million monthly active users) this analysis will dive into factors that could potentially drive housing prices.

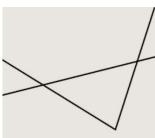




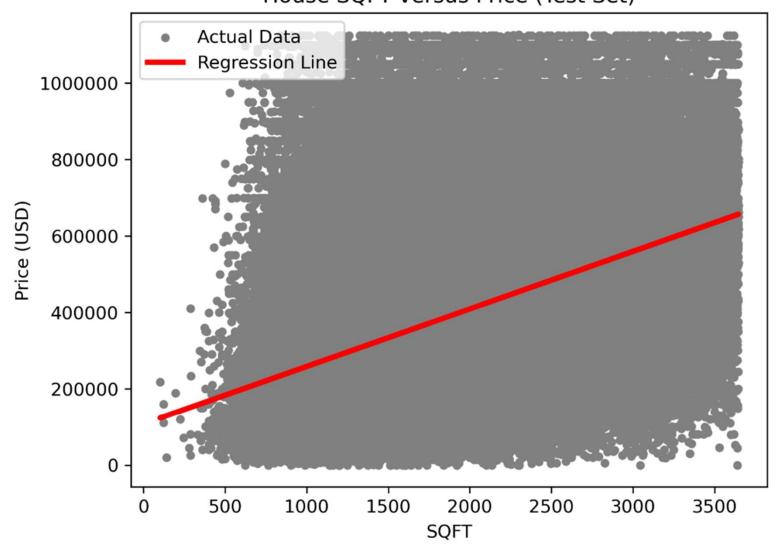
To start the exploration, we started by looking for any **linear relationship** between variables.

As this chart shows **there is not any linear dependence** between number of baths or bedrooms or size of the house or land. You can see that the median cost of the house is not significantly affected by any of these factors.

This led us to conclude there must be another factor at work here.



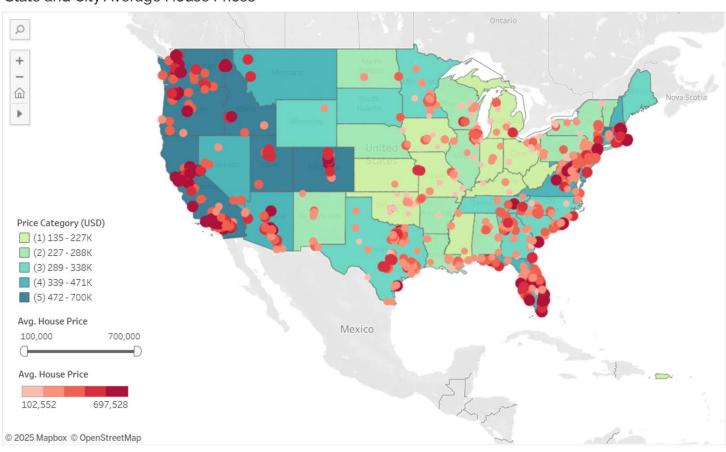
### House SQFT versus Price (Test Set)



To confirm the hypothesis that house features (house square footage being the example here) did not significantly impact house price a linear regression was conducted.

The results indicated that only 19% of house cost could be attributed to square footage. Therefore a linear regression was not enough to explain house costs.





The age-old adage appears to still be accurate. The data reveals that location is in fact the largest factor when navigating house prices.

As you can see by the map; large, dense cities and popular places to live (coasts, islands, and metropolitan areas) are still the most expensive parts of the USA real estate market.

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