

# ANALYSIS ON SALES PREDICTION USING DIFFERENT REGRESSION MODELS.

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## INTRODUCTION:

**This presentation will provide an overview of regression analysis, a powerful statistical technique used to understand the relationship between advertising channels and product sales. We'll explore the key steps involved, from data collection to model interpretation, equipping you with the tools to make data-driven decisions.**



# Table of Content:

- 1) Introduction.**
- 2) Objectives.**
- 3) Data collection and Preprocessing.**
- 4) Exploratory Data Analysis.**
- 5) Fitting Of Multiple Linear Regression Model.**
- 6) Prediction of Response Variable.**
- 7) Conclusion.**
- 8) Key Findings and Recommendation.**
- 9) Future Scope and Limitations.**

# OBJECTIVES:

- 1) Investigate the impact of advertising expenditure levels across different channels on product sales**
- 2) To check the multicollinearity among predictor variables.**
- 3) To check which regressor variable is most significant.**
- 4) To understand relationship of variables.**
- 5) Identify the most influential subset of features that significantly contribute to predict the response variable.**
- 6) To find which model can predict the response variable more accurately.**

# Data Collection and Preprocessing

## 1 Gather Relevant Data

Collected data on advertising spend and product sales from Kaggle.

<https://www.kaggle.com/datasets/singhnavjot/2062001/product-advertising-data>

## 3 Dataset Overview

Number of observations: 300

Number of variables: 7

## 2 Data Cleaning

Ensure data quality by identifying and addressing missing values, outliers, and inconsistencies.

## 4 Data Distribution

Data is Normally Distributed.



# Description Of Variables

Independent Variable

**1)Product Sales**

1

2

Dependent Variable

**1)TV**

**2)BillBoards**

**3) Google\_Ads**

**4) Social\_Media**

**5) Influencer\_Marketing**

**6) Affiliate\_Marketing**



# Model Selection and Fitting

## Multiple Linear Regression Model

- 1) Fit a Linear Regression Model.
- 2) Check Assumptions of Model.
- 3) Visual Inspection.
- 4) Statistical Test

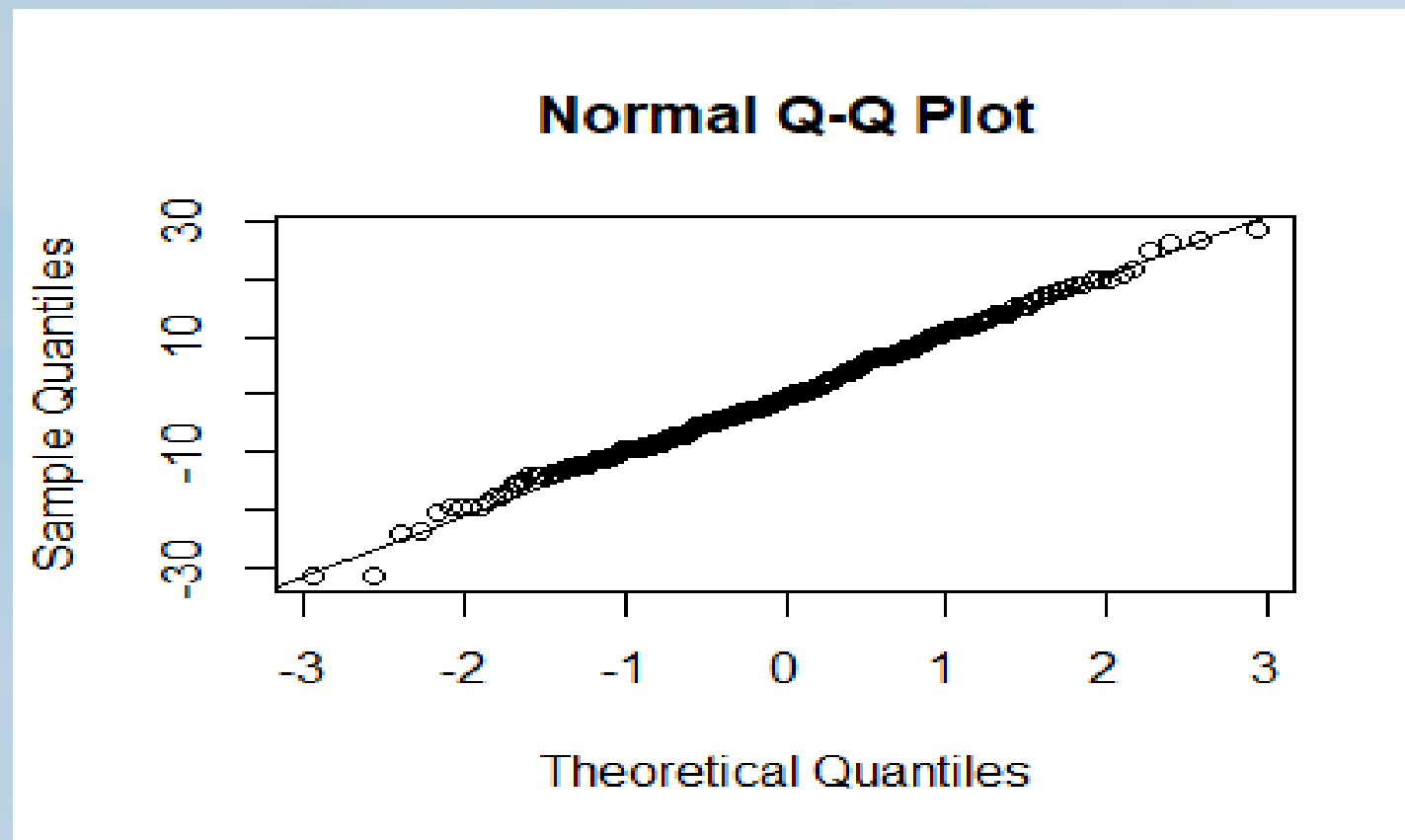
1

2

## Multicollinearity

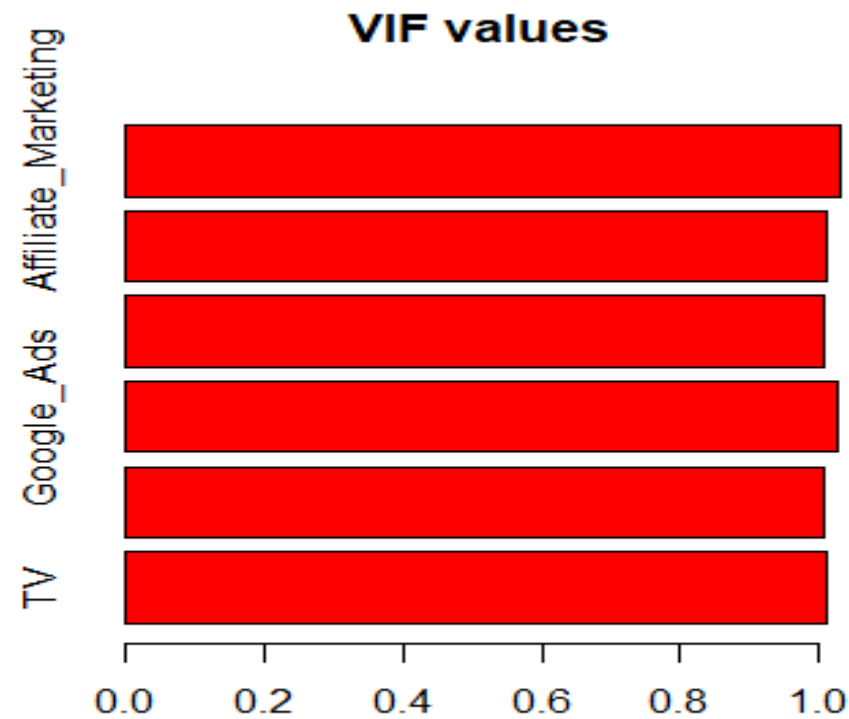
- 1) Test for Multicollinearity with VIF.
- 2) Data Visualisation.

Q) What is combined effect of advertising across all channels on Product Sales?





**Q) Does Multicollinearity among variables affect the accuracy and stability of regression model?**



**OUTPUT:**

TV	Billboards	Google_Ads	Social_Media	Influencer_Marketing
1.011536	1.006814	1.027308	1.007057	1.008573
Affiliate_Marketing				
1.030827				

# Interpreting Regression Coefficients

- Perform t-test to check the contribution of each regressor to overall prediction of Response Variable.

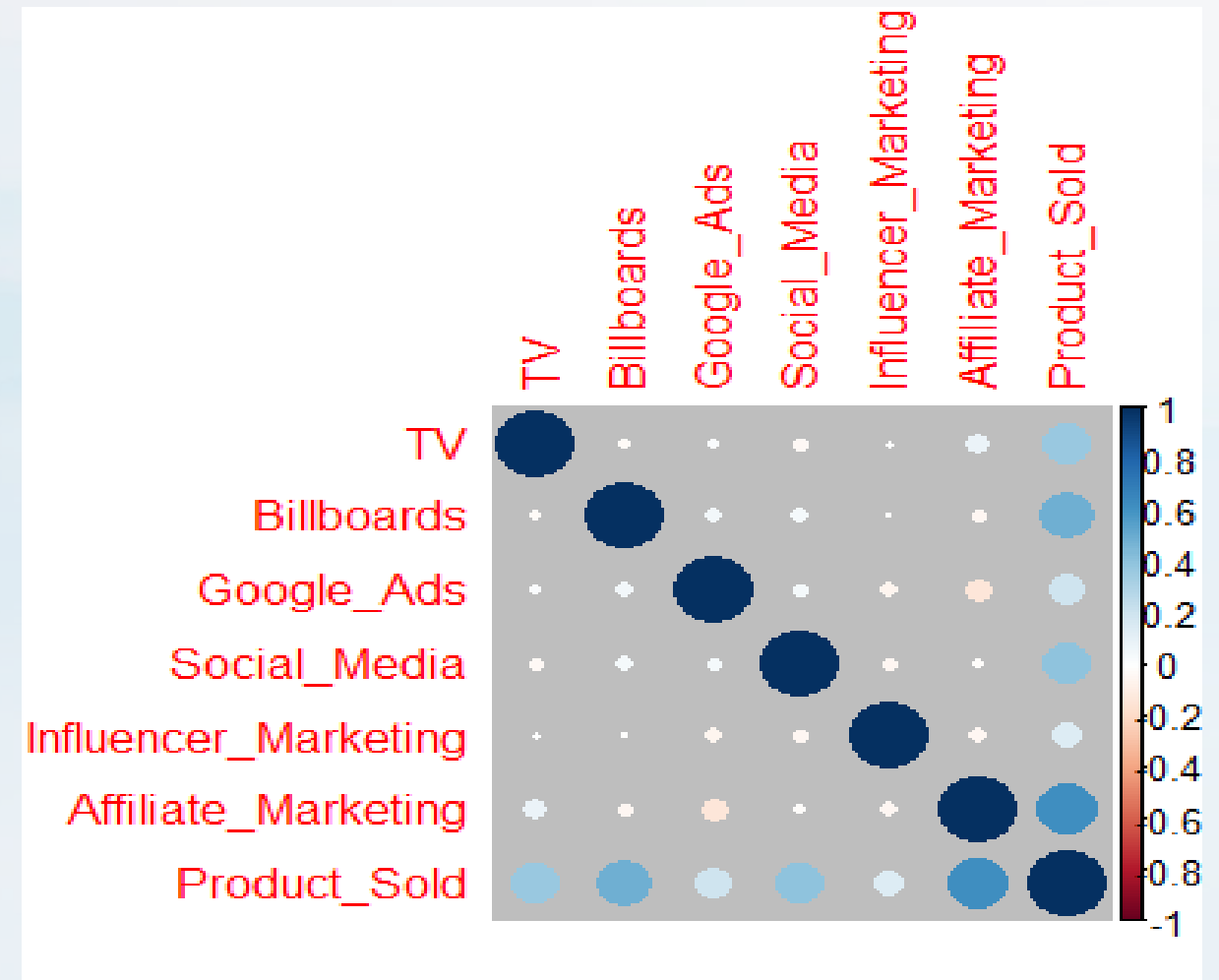
- Output:
- Coefficients:

	$\text{Pr}( >  t  )$
• (Intercept)	0.974
• TV	$<2e-16$
• Billboards	$<2e-16$
• Google_Ads	$<2e-16$
• Social_Media	$<2e-16$
• Influencer_Marketing	$<2e-16$
• Affiliate_Marketing	$<2e-16$



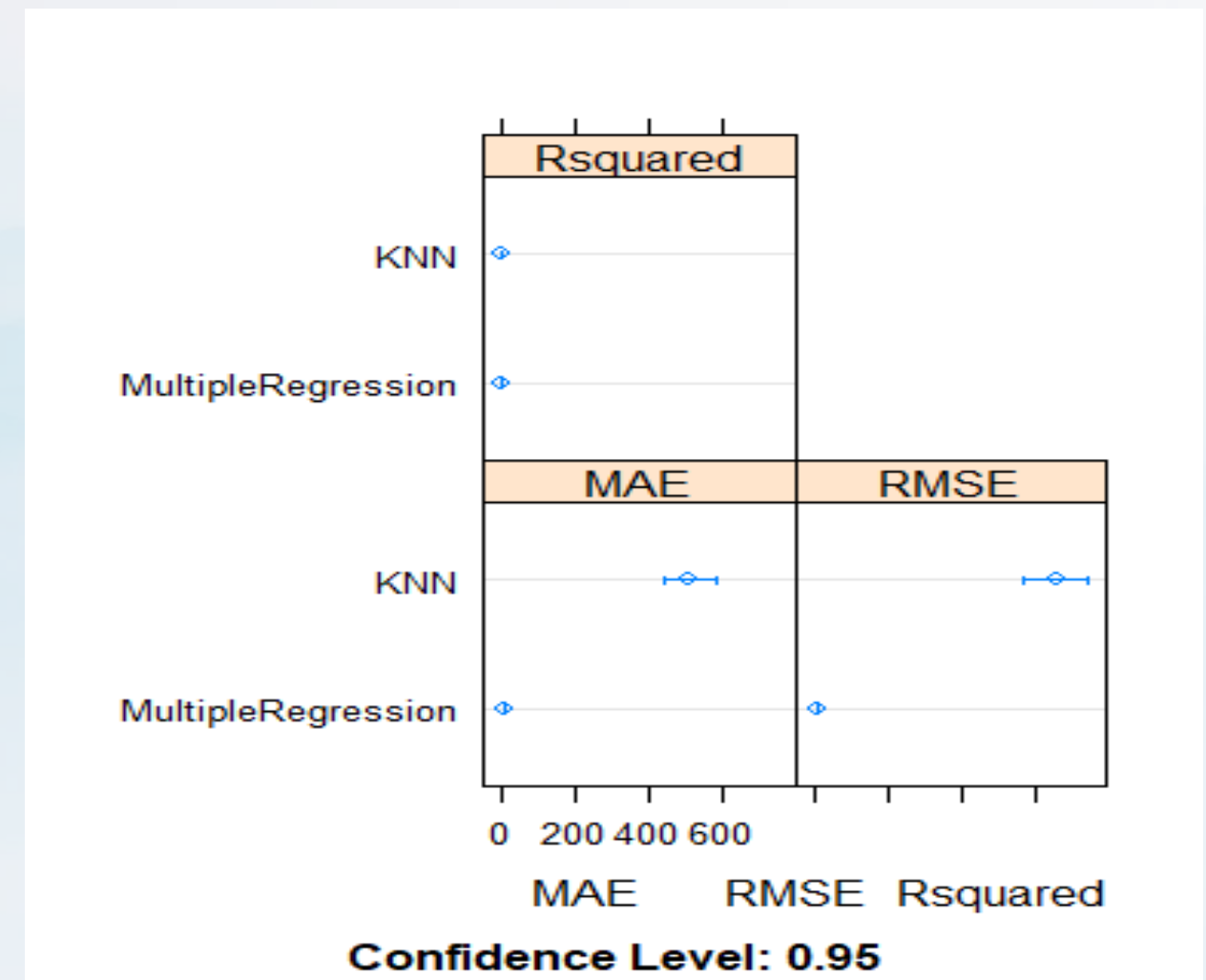
**Q)Does the level of advertising expenditure across different Channels have a significant impact on Product Sales?**

Regressors	Correlation Coefficient
TV	0.374227
Billboards	0.4805929
Google_Ads	0.2036781
Social_Media	0.4010880
Influencer_Marketing	0.1408437
Affiliate_Marketing	0.6116025

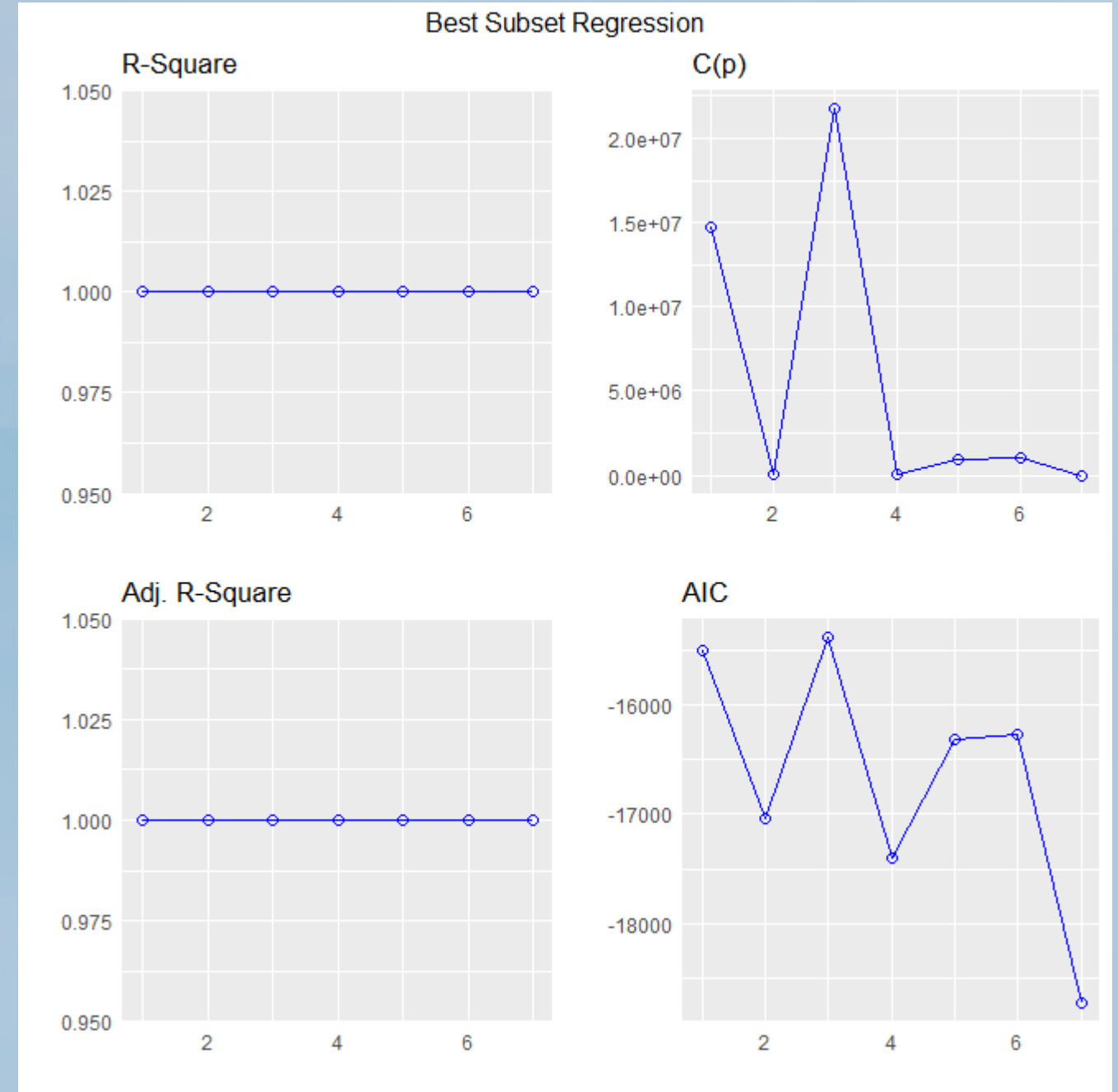


Q)Can the KNN Model accurately predict the given observation based on the dataset?

- knn\_model
- k-Nearest Neighbors
- 212 samples 6 predictor
- Resampling results:
- RMSE Rsquared MAE
- 658.1077 0.8809018 511.6202
- 
- lm\_model
- Linear Regression
- 212 samples 6 predictor
- Resampling results:
- RMSE Rsquared MAE
- 9.715039 0.999968 7.767454



**Q)Which is the most influential subset of features that significantly contribute to predict Product Sales?**



Q)Which is the most influential subset of features that significantly contribute to predict Product Sales ?

	rsquare	adjr	predrsq	cp	aic	sbic
TV	1	1	1	14690285.27	-15497.24	-16354.60
BillBoards	1	1	1	84738.27	-17040.82	-17900.15
Google Ads	1	1	1	21751960.59	-15375.49	-16236.85
Social Media	1	1	1	24671.45	-17404.53	-18267.72
Influencer Marketing	1	1	1	923796.27	-16319.09	-17184.44
Affiliate Marketing	1	1	1	1073937.95	-16271.92	-17139.28



## Q) How well do Advertising Budgets predict actual Product Values?

TV	Billboards	Google_Ads	Social_Media	Influencer_Marketing	Affiliate_Marketing
890	350	650	230	500	150

Predicted value of Product sold : 5579.988

# Evaluating Model Performance



## Model Fit

Assess the overall goodness of fit of the regression model using metrics like R-squared and adjusted R-squared.



## Prediction Accuracy

Evaluate the model's ability to accurately predict product sales based on advertising data.



## Residual Analysis

Examine the residuals of the model to identify any patterns or violations of assumptions.



## Model Refinement

Iteratively refine the model by adding or removing variables, trying different techniques, or addressing any issues identified.

# Conclusion and Recommendations

## Key Findings

- Identify the most effective advertising channels in driving product sales -
- Quantify the impact of each channel on sales through the regression coefficients -
- Highlight any unexpected or counterintuitive relationships between advertising and sales

## Recommendations

- Suggest optimal advertising budget allocation across channels to maximize sales -
- Recommend strategies for integrating advertising efforts across digital and traditional channels -
- Provide guidance on further data collection or analysis to refine the model and insights

# Future Scope and Limitation

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## FUTURE SCOPE:

**Personalized Marketing and Product Recommendations:** As companies gather more data on consumer preferences and behavior, they may be able to personalize marketing messages and product recommendations to individual customers. This could help drive demand for specific products or services and reduce the risk of stockouts by matching supply with highly targeted demand.

**Collaborative Supply Chain Management:** Collaboration among supply chain partners, facilitated by platforms and technologies, could become more prevalent. This could lead to greater transparency, coordination, and responsiveness across the supply chain, enabling companies to better anticipate and meet demand fluctuations.

## LIMITATION:

The dataset provided has several limitations that could impact its analysis:

**Limited Context:** The dataset lacks contextual information about the market, industry, or specific products being advertised. Without this context, it's challenging to draw meaningful insights or make accurate predictions.

**Single Time Frame:** It appears that the dataset covers a single time frame. Analyzing data from only one period may not capture trends or seasonal variations, limiting the depth of analysis.

**No Target Variable Explanation:** The dataset contains various advertising channels and corresponding amounts spent on each, but it's unclear what specific outcome or target variable these expenditures are intended to influence (e.g., sales, brand awareness, website traffic). Understanding the target variable is crucial for designing effective analytical models.

Thank  
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