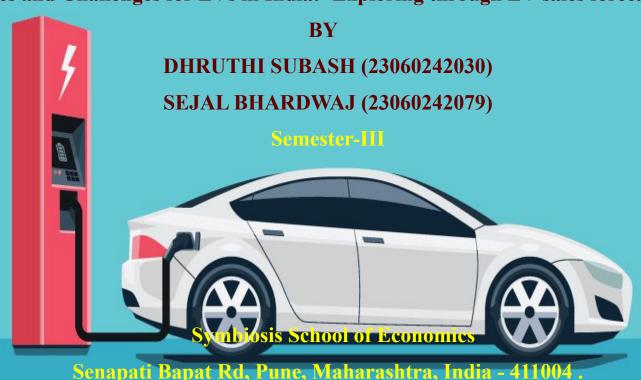


### **Msc Economics Python mini Project**

On

### Opportunities and Challenges for EVs in India:- Exploring through EV sales forecasting in India



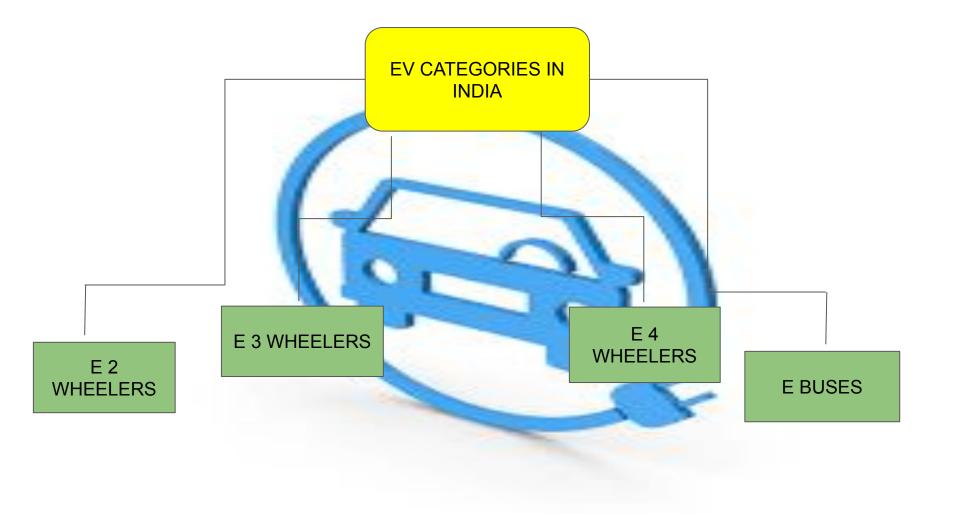
# PROBLEM STATEMENT

### **EV Sales Forecasting**

This project will leverage machine learning techniques and time series analysis to create an accurate and reliable sales forecasting model that can predict EV sales over a defined period, providing valuable insights into the future of electric mobility in India.

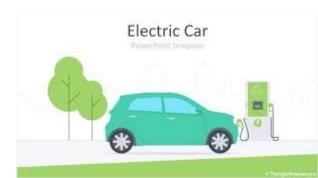
### **RESEARCH QUESTIONS**

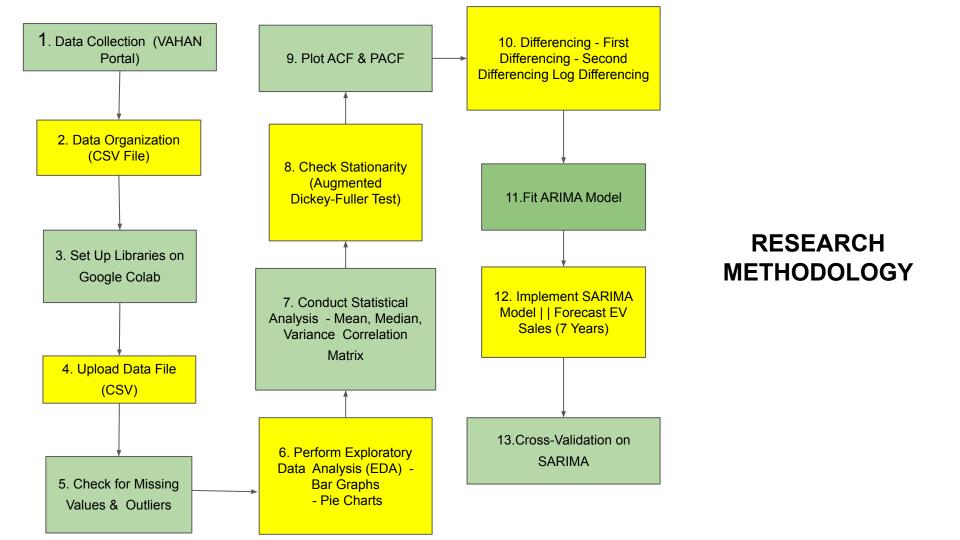
- 1) What is the scope of the EV industry in India?
- 2) How has the demand for EV been in the recent years (2017-2023)?
- 3) Which time of the year has experienced the most number of EV sales?
- 4) How can predictive models (such as SARIMA or machine learning models) be leveraged to improve the accuracy of long-term EV sales forecasts in India?



### **DATA COLLECTION**

The data has been retrieved from the **VAHAN PORTAL**, The data on the monthly sales of EVs - 2 wheeler, 3 wheeler, 4 wheeler and bus from the years 2017- 2023 has been collected and made into a CSV file which was then uploaded on google colab to perform the further analysis.



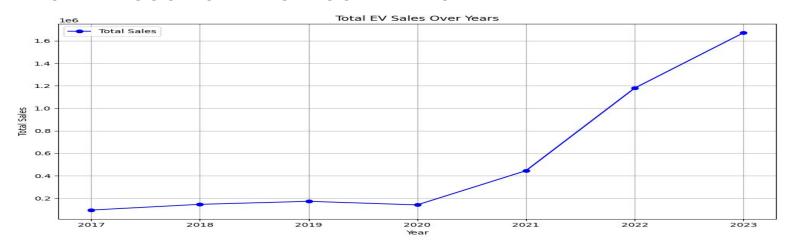


# SET UP THE DEVELOPMENT ENVIRONMENT

```
import warnings
warnings.filterwarnings('ignore')
# Import packages
import pandas as pd
import numpy as np
import itertools
from statsmodels.tsa.seasonal import seasonal decompose
import math
from sklearn.preprocessing import OneHotEncoder
from sklearn.model selection import train test split, cross val score, KFold, TimeSeriesSplit, GridSearchCV
from sklearn.base import BaseEstimator, RegressorMixin
from sklearn.svm import SVR
from sklearn.linear model import LinearRegression as LR
from sklearn.linear model import Ridge
from sklearn.metrics import mean squared error, mean absolute percentage error, mean absolute error
import statsmodels.api as sm
from statsmodels.tsa.stattools import acf, pacf
from statsmodels.tsa.statespace.sarimax import SARIMAX
from statsmodels.graphics.tsaplots import plot acf, plot pacf
!pip install pmdarima
from pmdarima import auto arima, ARIMA
from pmdarima.model selection import SlidingWindowForecastCV
from pmdarima.model selection import cross val score as SARIMACV
```

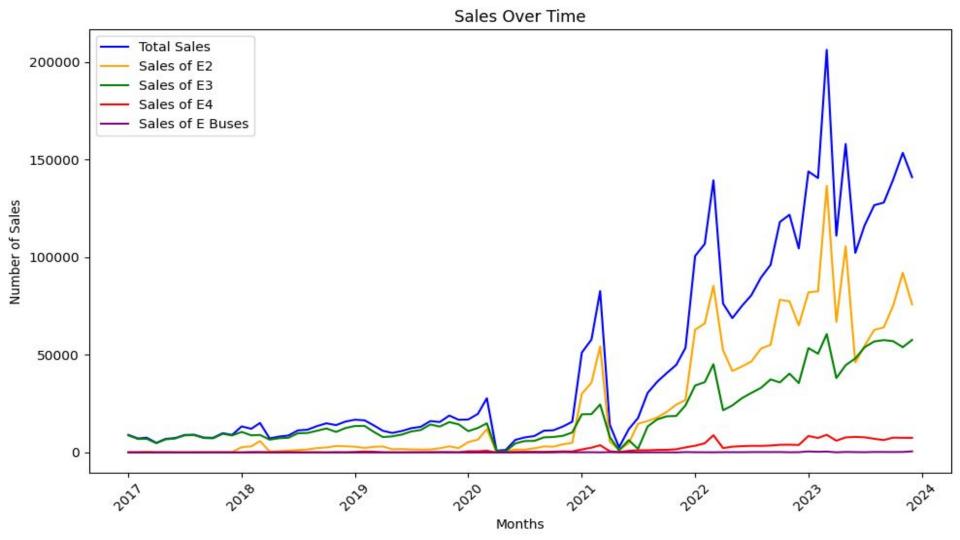
# **EXPLORATORY DATA ANALYSIS**

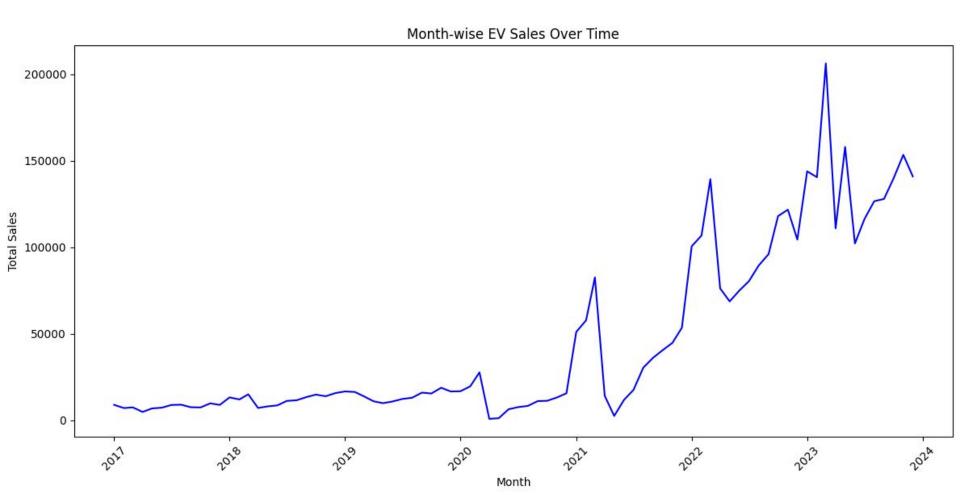
### YEAR - WISE ANALYSIS OF TOTAL EV SALES OVER YEARS



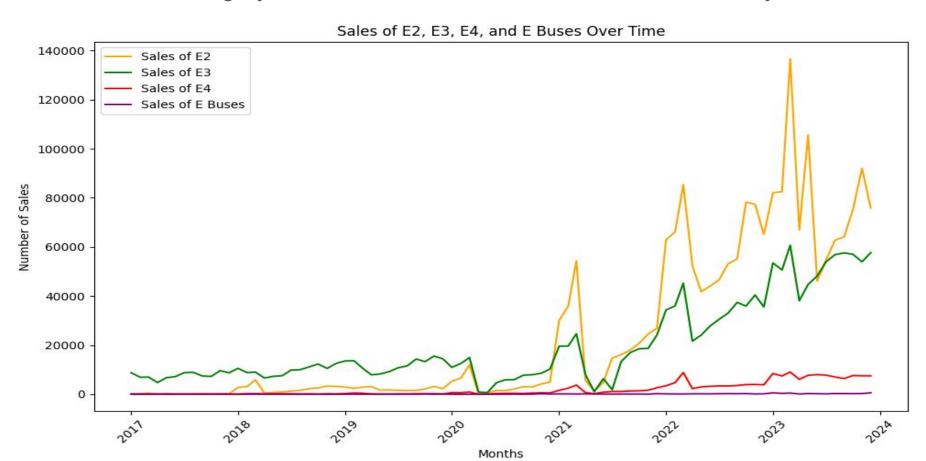
#### **Overall Trend**

- Significant upward trend with rapid acceleration, particularly from 2021 onward.
- Reflects a strong shift toward EV adoption in India, fueled by policy support and environmental awareness.





# Category wise sales of E2,E3,E4 and E-Buses over the years



## **CATEGORY WISE EV SALES (2017-2023)**

# **Key Insights from Category wise EV Sales** (2017-2023)

#### Exponential Growth Post-2021

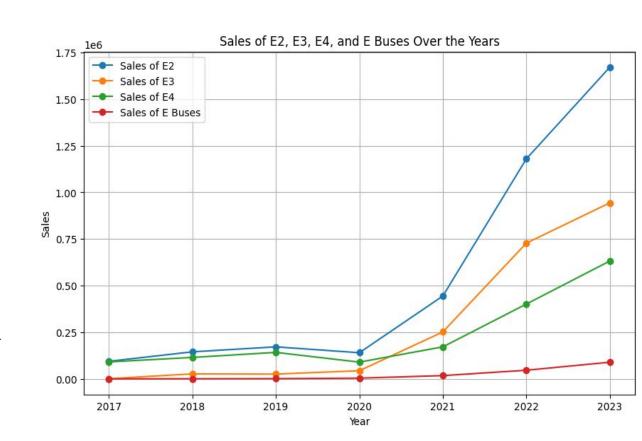
 Sales surged significantly from 2021 onwards.

### 2. Category Highlights

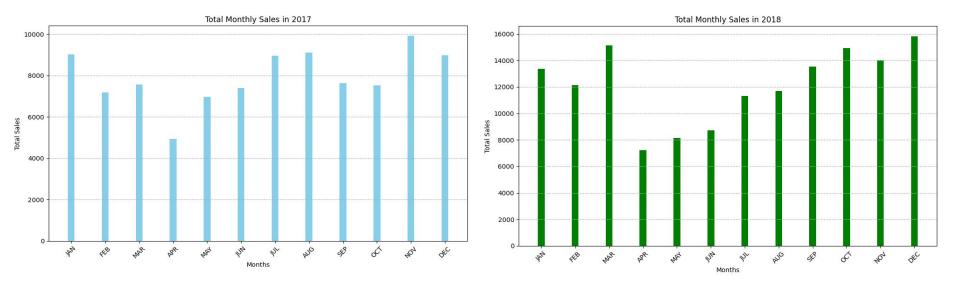
- **E2 & E3 Vehicles:** Saw the largest sales increases after 2020.
- E4 Vehicles: Steady growth, similar upward trend.
- E Buses: Notable growth in 2021-2022, driven by public transport electrification.

### 3. Key Drivers

- Government Incentives and Improved EV Infrastructure.
- Increased Consumer Demand for eco-friendly options.

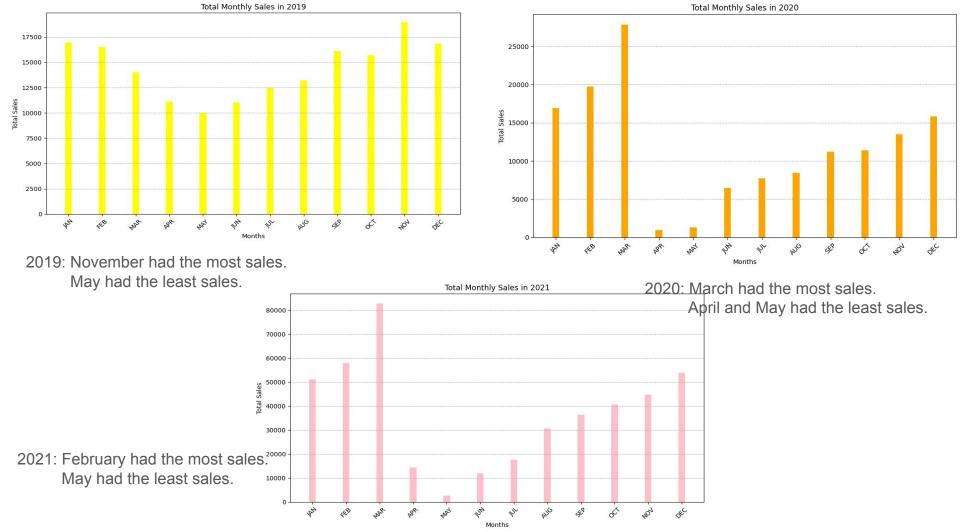


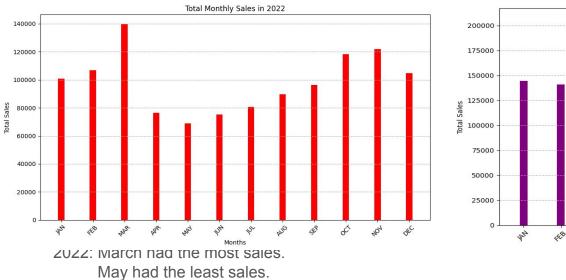
# **TOTAL MONTHLY SALES OF EVs FROM 2017 TO 2023**

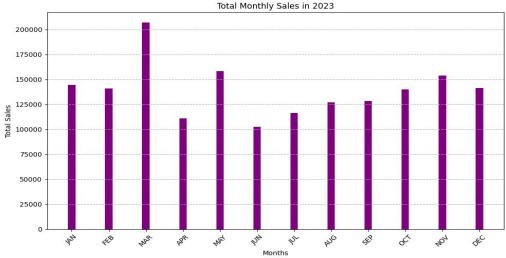


2017: November and December had the most sales.
April and May had the least sales.

2018: December had the most sales. April and May had the least sales.







Breakdown of EV Sales over the years (2017-2023)

2023: March had the most sales. June had the least sales.

• So we can see from the graphs that November and March have had the most number of sales and the sales have been the least in the month of May.

# REASON FOR THE SPIKE OF SALES DURING NOVEMBER AND MARCH

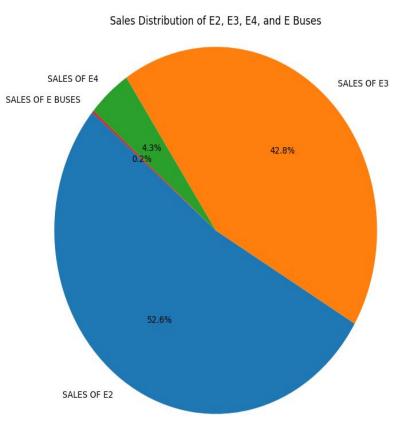
### Festive Season and Year-End Offers (November):

- Diwall: November often sees a spike in consumer purchases due to Diwali, one of India's biggest festivals. People tend to buy new items, including cars, as a sign of prosperity during this time.
- Discounts and Promotions: Car manufacturers and dealerships typically offer substantial discounts, promotional
  offers, and year-end clearances in November to boost sales and clear out inventory before the end of the financial
  veat

### Financial Year-End (March):

- Tax Planning: March marks the end of the financial year in India. Many buyers—especially businesses and professionals—lend to make large purchases, such as vehicles, to take advantage of tax-saving incentives, depreciation benefits, and fiscal planning.
- Bonus and Incentives: March is also a common month for employees to receive annual bonuses or incentives, which
  often drives higher consumer spending, including on vehicles.

# **E2,E3,E4 AND E BUS OVERALL SALES DISTRIBUTION**



### **EV Sales Distribution by Category**

- E3 Vehicles:
  - 52.63% of total sales
  - Most popular category among consumers.
- E2 Vehicles:
  - 42.82% of total sales
  - Strong market presence, close behind E3.
- E4 Vehicles:
  - 4.34% of total sales
  - Smaller market share with potential for growth.
- E Buses:
  - 0.20% of total sales
  - Minimal share, but potential to expand with targeted efforts.

Percentage\_sales = (total\_sales\_per\_category / total\_sales) \* 100

# Electric two-wheelers (E-2Ws) and three-wheelers (E-3Ws) are popular in India for a number of reasons, including

Lower cost of ownership

Electric 3Ws are cheaper to own than their gasoline counterparts, even without subsidies.

Government subsidies

The Faster Adoption and Manufacturing of Electric Vehicles (FAME II) scheme has reduced the cost of ownership of electric 3Ws.

Favorable total cost of ownership

The total cost of ownership of electric 2Ws and 3Ws is favorable in India.

Less developed road infrastructure

Two-wheeled vehicles are more popular in developing countries like India because they are often used for short distances around cities.

Economical option

Three-wheelers are promoted as an economical option for short- to medium-distance public transportation.

Local manufacturing

Local manufacturing of batteries, critical components, and charging infrastructure can reduce costs and improve the acceptability of EVs.

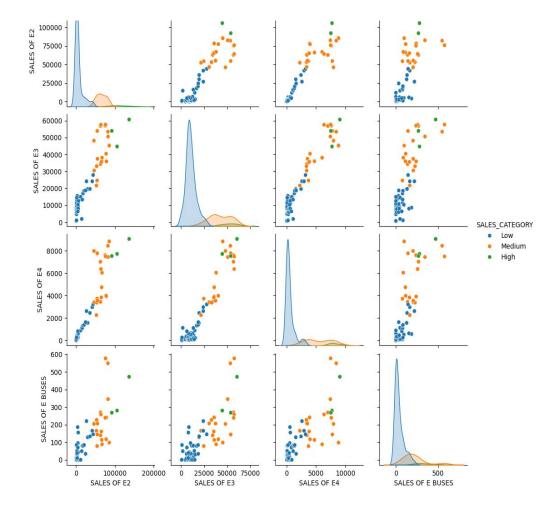
Renewable energy

India plans to install 500 gigawatts of clean energy by the end of the decade and reach net zero emissions by 2070.

India is the world's largest electric 3W market, and the second largest electric 2W market globally.

#### Key Insight:

E3 & E2 dominate the market, while E4 and E Buses showroom for expansion, requiring focused marketing and awareness campaigns.



# Pair Plot Analysis: EV Sales (E2, E3, E4, E-Buses)

#### Sales Distribution:

- E2 (Two-Wheelers): Skewed right, with most sales at lower values and a few high outliers.
- E3 (Three-Wheelers): Similar to E2, with lower sales concentrated and an upward trend.
- E4 (Four-Wheelers): Lower sales overall, with a strong right skew.
- E-Buses: Lowest sales, with most data concentrated at very low values.

### 2. Key Relationships:

- E2 & E3: Strong positive correlation, indicating higher E2 sales are associated with higher E3 sales.
- E2 & E4 / E3 & E4: Positive but less concentrated correlation.
- E2/E3 & E-Buses: Weaker, more scattered relationships.
- E4 & E-Buses: Weakest correlation among all, with little connection.

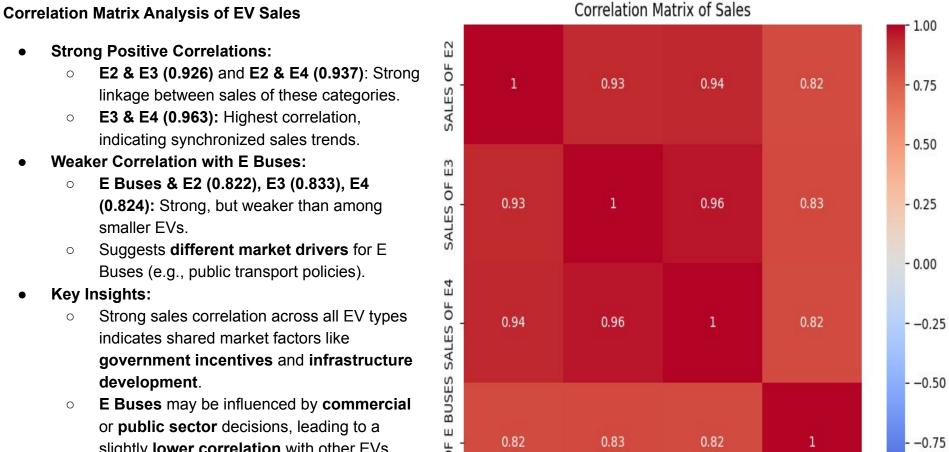
### 3. **Summary**:

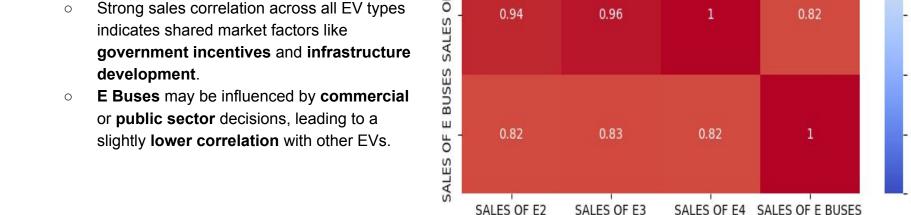
- E2 and E3 dominate the market, showing strong interdependence.
- E4 and E-Buses have lower sales and weaker correlations with other categories.

### STATISTICAL ANALYSIS

```
import pandas as pd
    # For the columns: 'SALES OF E2', 'SALES OF E3', 'SALES OF E4', 'SALES OF E BUSES'
    # Calculate mean, median, and variance for each sales column
    mean sales = data[['SALES OF E2', 'SALES OF E3', 'SALES OF E4', 'SALES OF E BUSES']].mean()
    median_sales = data[['SALES OF E2', 'SALES OF E3', 'SALES OF E4', 'SALES OF E BUSES']].median()
    variance sales = data[['SALES OF E2', 'SALES OF E3', 'SALES OF E4', 'SALES OF E BUSES']].var()
    std_dev_sales = data[['SALES OF E2', 'SALES OF E3', 'SALES OF E4', 'SALES OF E BUSES']].std()
    # Display the results
    print("Mean Sales:\n", mean sales)
    print("\nMedian Sales:\n", median sales)
    print("\nVariance of Sales:\n", variance sales)
    print("\nStandard Deviation of Sales:\n", std dev sales)
→ ▼ Mean Sales:
     SALES OF E2
                         24124.642857
    SALES OF E3
                        19629.285714
    SALES OF E4
                         1990.154762
    SALES OF E BUSES
                           92.380952
    dtype: float64
    Median Sales:
     SALES OF E2
                          3241.0
    SALES OF E3
                        11916.0
    SALES OF E4
                          443.5
    SALES OF E BUSES
                           38.5
    dtype: float64
    Variance of Sales:
     SALES OF E2
                         1.042468e+09
    SALES OF E3
                        2.788771e+08
    SALES OF E4
                        7.434483e+06
    SALES OF E BUSES
                       1.500988e+04
    dtvpe: float64
    Standard Deviation of Sales:
     SALES OF E2
                         32287.277239
    SALES OF E3
                        16699.614667
    SALES OF E4
                         2726.624770
    SALES OF E BUSES
                         122.514804
    dtype: float64
```

# MEAN, MEDIAN, VARIANCE AND STANDARD DEVIATION





-1.00

SALES OF E4 SALES OF E BUSES

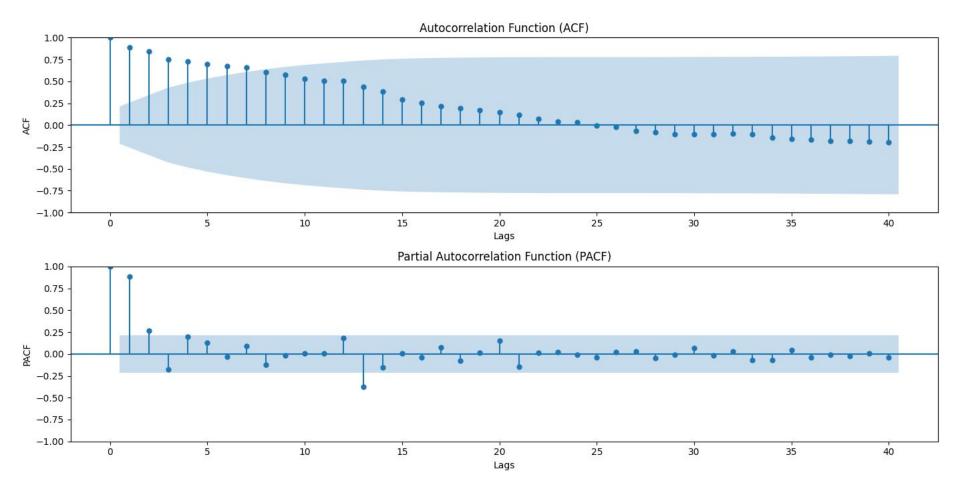
### LIMITATIONS AND SCOPE

- 1. Policy Impact:
  - Analyze the influence of government initiatives (subsidies, tax incentives, infrastructure) on EV adoption.
- 2. Technological Advancements:
  - Explore innovations in battery technology (energy density, charging speed, cost reduction).
- 3. Smart Grids & Renewable Energy:
  - Examine the integration of EVs with smart grids and renewable energy sources for efficiency and sustainability.
- 4. Consumer Behavior:
  - Investigate factors influencing EV purchasing decisions (environmental awareness, economic considerations).
- 5. Environmental Benefits:
  - Quantify EV impact on CO2 reduction and air quality improvements.
- 6. Economic Impact:
  - Study job creation, GDP effects, and industry shifts driven by EV transition.

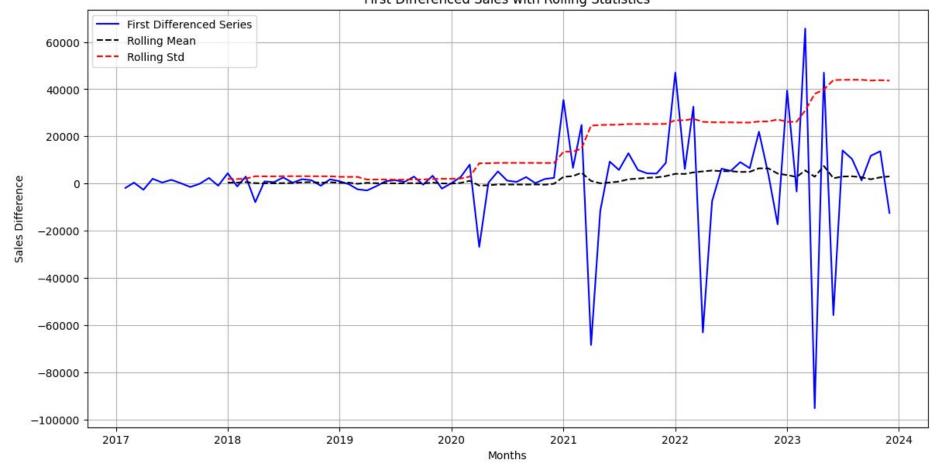
This concise version is ready for your slide, keeping the core points for clarity and impact.

Rolling Mean & Standard Deviation of Month-wise Sales Original Series Rolling Mean 200000 -Rolling Std S 100000 

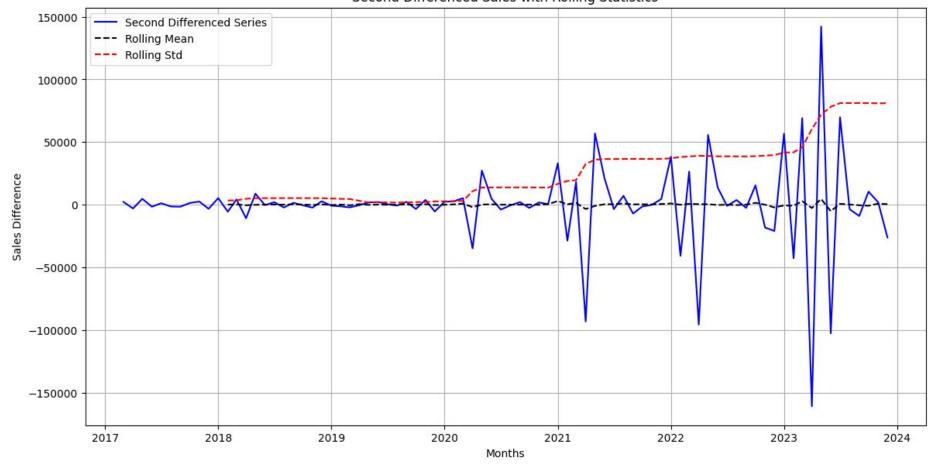
Month



First Differenced Sales with Rolling Statistics

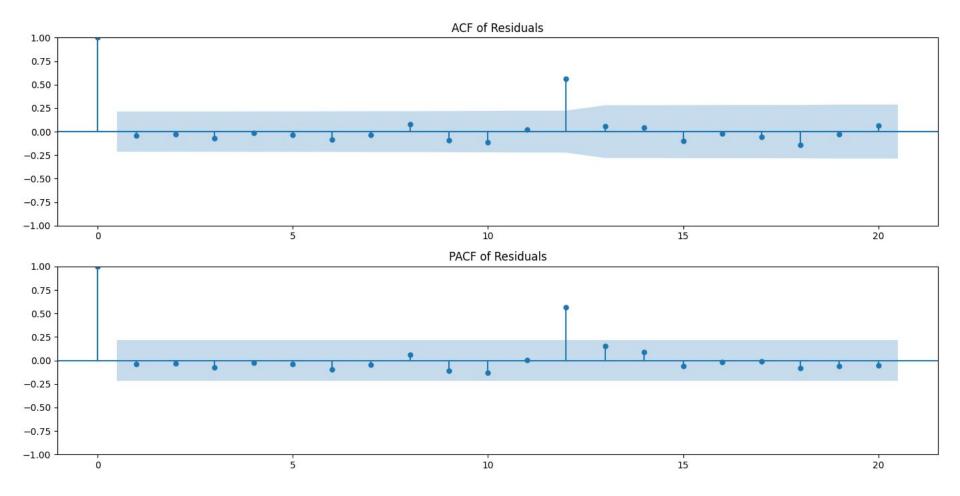


Second Differenced Sales with Rolling Statistics



# SARIMAX Results

Dep. Variable: Total Some Model: ARIMA(6, 1) Date: Fri, 18 Oct of the model: 18:28		Total S	ales No.	Observations:		84
		, 1) Log	g Likelihood		-930.386 1876.771 1896.122	
		2024 AIC				
		8:30 BIC				
Sample: 01-01-		2017 HQIC			1884.545	
		- 12-01-	2023			
Covariance Type:			opg			
y=======	coef	std err	z	P> z	[0.025	0.975]
ar.L1	0.2263	1.010	0.224	0.823	-1.754	2.206
ar.L2			0.794	0.427	-0.468	
ar.L3	-0.4001	0.139	-2.875	0.004	-0.673	-0.127
ar.L4	-0.0499	0.360	-0.139	0.890	-0.755	0.656
ar.L5	0.1461	0.281	0.520	0.603	-0.405	0.697
ar.L6	-0.0873	0.172	-0.507	0.612	-0.424	0.250
ma.L1	-0.5931	1.021	-0.581	0.561	-2.594	1.407
sigma2	3.472e+08	1.17e-08	2.98e+16	0.000	3.47e+08	3.47e+08
Ljung-Box (L1) (Q):			0.14	Jarque-Bera	(JB):	30.07
Prob(Q):			0.70	Prob(JB):	()	0.00
Heteroskedasticity (H):			152.40	Skew:		-0.09
Prob(H) (two-sided):			0.00	Kurtosis:		5.94

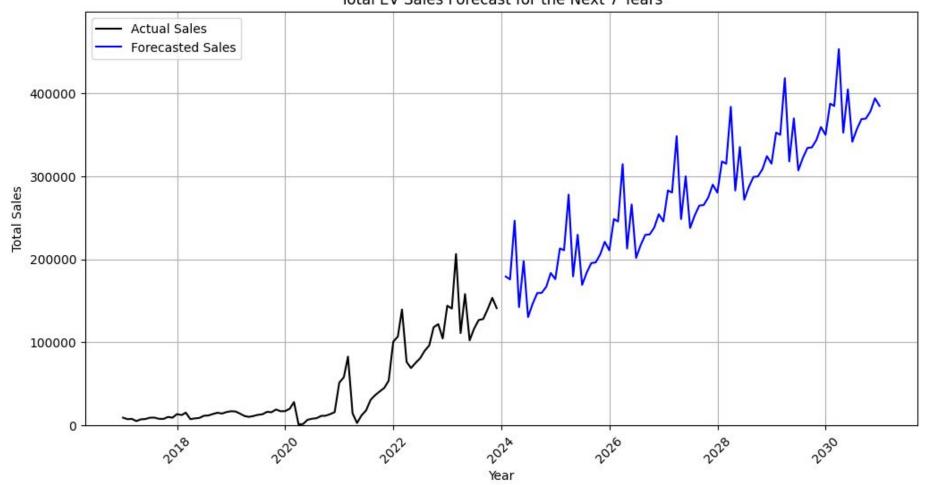


```
SARIMAX Results
______
Dep. Variable:
                                 Total Sales No. Observations:
               SARIMAX(1, 1, 1)x(1, 1, 1, 12) Log Likelihood
Model:
                                                                       -772.784
                            Fri, 18 Oct 2024 AIC
                                                                       1555.567
Date:
Time:
                                   18:39:20 BIC
                                                                       1566.880
Sample:
                                 01-01-2017 HOIC
                                                                       1560.066
                                - 12-01-2023
Covariance Type:
                                        opg
              coef
                                          P> | z |
                     std err
                                                    [0.025
                                                              0.975]
          -0.6609 0.197 -3.347 0.001 -1.048 -0.274
ar.L1
        0.2684 0.282 0.953 0.341 -0.284 0.821

-0.5701 0.921 -0.619 0.536 -2.374 1.234

0.7726 0.904 0.854 0.393 -1.000 2.545
ma.L1
ar.5.L12
ma.S.L12
sigma2 1.805e+08 2.19e-08 8.23e+15 0.000 1.81e+08 1.81e+08
______
Ljung-Box (L1) (Q):
                                 0.03 Jarque-Bera (JB):
                                                                    9.96
Prob(0):
                               0.86 Prob(JB):
                                                                    0.01
Heteroskedasticity (H):
                             43.82
                                      Skew:
                                                                    -0.18
Prob(H) (two-sided):
                                0.00 Kurtosis:
                                                                    4.80
Warnings:
[1] Covariance matrix calculated using the outer product of gradients (complex-step).
[2] Covariance matrix is singular or near-singular, with condition number 2.8e+31. Standard errors may be unstable.
          Forecasted Total Sales
2024-01-31
                 179169.213272
2024-02-29
                 175702.073054
2024-03-31
                 246376.804092
2024-04-30
                 142302.145561
2024-05-31
                 197581.295038
. . . . . . .
               368865.540468
2030-08-31
2030-09-30
                 369491.742224
2030-10-31
                 378412.971901
2030-11-30
                 393959.556917
2030-12-31
                 384786.394157
```

Total EV Sales Forecast for the Next 7 Years



```
12s
        Performing stepwise search to minimize aic
   <del>_</del>_<del>+</del>
         ARIMA(2,0,2)(1,1,1)[12] intercept : AIC=979.699, Time=0.77 sec
         ARIMA(0,0,0)(0,1,0)[12] intercept : AIC=1001.845, Time=0.01 sec
         ARIMA(1,0,0)(1,1,0)[12] intercept : AIC=973.646, Time=0.15 sec
         ARIMA(0,0,1)(0,1,1)[12] intercept
                                             : AIC=987.070, Time=0.12 sec
         ARIMA(0,0,0)(0,1,0)[12]
                                             : AIC=1009.735, Time=0.02 sec
         ARIMA(1,0,0)(0,1,0)[12] intercept : AIC=976.772, Time=0.03 sec
         ARIMA(1,0,0)(2,1,0)[12] intercept : AIC=975.199, Time=0.25 sec
         ARIMA(1,0,0)(1,1,1)[12] intercept
                                             : AIC=974.938, Time=0.16 sec
         ARIMA(1,0,0)(0,1,1)[12] intercept
                                            : AIC=973.421, Time=0.13 sec
         ARIMA(1,0,0)(0,1,2)[12] intercept : AIC=975.242, Time=0.31 sec
         ARIMA(1,0,0)(1,1,2)[12] intercept : AIC=976.932, Time=0.42 sec
         ARIMA(0,0,0)(0,1,1)[12] intercept
                                             : AIC=1003.480, Time=0.07 sec
         ARIMA(2,0,0)(0,1,1)[12] intercept
                                            : AIC=974.622, Time=0.14 sec
         ARIMA(1,0,1)(0,1,1)[12] intercept
                                            : AIC=975.344, Time=0.16 sec
         ARIMA(2,0,1)(0,1,1)[12] intercept
                                            : AIC=976.388, Time=0.37 sec
         ARIMA(1,0,0)(0,1,1)[12]
                                             : AIC=973.113, Time=0.08 sec
         ARIMA(1,0,0)(0,1,0)[12]
                                             : AIC=976.785, Time=0.02 sec
         ARIMA(1,0,0)(1,1,1)[12]
                                             : AIC=974.306, Time=1.02 sec
         ARIMA(1,0,0)(0,1,2)[12]
                                             : AIC=974.767, Time=1.99 sec
         ARIMA(1,0,0)(1,1,0)[12]
                                             : AIC=972.442, Time=0.99 sec
         ARIMA(1,0,0)(2,1,0)[12]
                                             : AIC=974.339, Time=0.42 sec
         ARIMA(1,0,0)(2,1,1)[12]
                                             : AIC=976.198, Time=0.48 sec
                                             : AIC=1011.915, Time=0.05 sec
         ARIMA(0,0,0)(1,1,0)[12]
         ARIMA(2,0,0)(1,1,0)[12]
                                             : AIC=974.212, Time=0.16 sec
         ARIMA(1,0,1)(1,1,0)[12]
                                             : AIC=973.436, Time=0.34 sec
         ARIMA(0,0,1)(1,1,0)[12]
                                             : AIC=990.091, Time=0.09 sec
         ARIMA(2,0,1)(1,1,0)[12]
                                             : AIC=976.250, Time=0.39 sec
```



Best model: ARIMA(1,0,0)(1,1,0)[12]

Total fit time: 9.213 seconds

Mean Absolute Error (Set 1): 44554.05720416259

Forecast for the next 84 months: [178306.88154851 175873.0951826 241166.58712275 146098.52413031 193014.78758505 137283.41946636 151298.69386203 161686.63412514

163029.57114257 174793.92294467 188497.32634678 176039.58964645 213348.35499154 210914.61991029 276208.08921671 181140.03621334

228056.29525955 172324.9290865 186340.20262349 196728.14326557

198071.08011574 209835.43199166 223538.8353612 211081.09867523 248389.86401399 245956.12893553 311249.59824072 216181.54523789

263097.80428386 207366.43811092 221381.71164786 231769.65228996

233112.58914012 244876.94101604 258580.34438558 246122.60769962

283431.37303837 280997.63795992 346291.10726511 251223.05426227 298139.31330825 242407.9471353 256423.22067225 266811.16131435

268154.09816451 279918.45004043 293621.85340997 281164.11672401 318472.88206276 316039.1469843 381332.61628949 286264.56328666

333180.82233263 277449.45615969 291464.72969663 301852.67033873

303195.60718889 314959.95906482 328663.36243435 316205.62574839 353514.39108714 351080.65600869 416374.12531388 321306.07231105

368222.33135702 312490.96518407 326506.23872102 336894.17936312

338237.11621328 350001.4680892 363704.87145874 351247.13477278 388555.90011153 386122.16503308 451415.63433827 356347.58133543

403263.84038141 347532.47420846 361547.7477454 371935.6883875

373278.62523767 385042.97711359 398746.38048312 386288.64379716]

# FORECASTS OF EV SALES DONE WITH AUTHENTIC **SOURCES**

- Electric vehicle sales in India are projected to reach 10 million units by 2030 Economic times
- The EV market is expected to grow at a CAGR of 49% from 2022 to 2030, with major contributions from the e-2W and e-3W segments - Clean Mobility shift
- Union Minister Nitin Gadkari forecasts that India's EV sales will reach 10 million units per year by 2030. Business Standard projects that EV sales will reach 27.2 million units by 2032. - Business standard

EV sales in India to reach 10 mn units by 2030, to create 5 mn jobs: Nitin Gadkari

By Sharmistha Mukherjee, ET Bureau + Last Updated: Sep 10, 2024, 11:35:00 PM IST



S ≪ AA □

Electric vehicle sales in India are projected to reach 10 million units by 2030, creating 5 million jobs, according to Union Minister Nitin Gadkari. The market potential is expected to touch Rs 20 lakh crore, with the EV finance market growing to Rs 4 lakh crore. Battery



Inaugural Session of the 64th SIAM Annual Convention 2024, in New Delhi, Union Minister Nitin Gadkari also urged automakers to take

Electric vehicle sales in the local market will grow to 10 million units by the turn of the decade, generating employment for 5 million people, Union Minister for Road, Transport & Highways Nitin Gadkari said on Tuesday.

Gadkari, who was speaking at the 64th Annual Convention of the Society of Indian Automobile Manufacturers (CIAM) said the market notantial for

INDUSTRY POLICY & REGULATION PRODUCTS & TECHNOLOGY MARKET & TRENDS ECOSYSTEM EV VAHAN DASHBOAR

instrumental in managing lithium-ion batteries, extending their lifespan, improving energy efficiency, and reducing waste - crucial factors for sustainable mobility. However, EVs require up to three times more semiconductor content than conventional vehicles.

Given their huge role in the digitisation drive of the country, as per PwC, the semiconductor market in India is expected to reach \$32 billion by 2025. The EV market accounts for a significant share of this growth. The Economic Survey of India 2023 forecasts a 49% compound annual growth rate (CAGR) in India's domestic EV market between 2022 to 2030, with an estimated 10 million annual sales by 2030. On the demand side, therefore, semiconductor technology is on the upswing.

# Conclusion - Opportunities and Challenges for EVs in India

### 1. Economic Opportunities

- Lower Operating Costs: Electricity is cheaper than gasoline, and government subsidies can reduce the cost of EVs over time.
- **Job Creation:** Expansion in EV manufacturing and infrastructure development can generate employment opportunities in India.

### 2. Technological Opportunities

- Advancements in Battery Tech: As battery costs decrease, EVs will become more affordable, leading to wider adoption.
- Innovation in Charging Infrastructure: Development of faster charging stations and renewable energy-based grids can enhance EV convenience.

### 3. Environmental Opportunities

- Reduced Emissions: EVs contribute to a significant reduction in emissions compared to traditional gasoline vehicles, helping combat global warming.
- Sustainable Energy Integration: EVs can integrate with renewable energy sources, reducing dependency on fossil fuels.

### 4. Social Opportunities

- **Health & Safety:** Reduction in air pollution due to fewer emissions from EVs can improve public health outcomes.
- Growing Acceptance: As more people adopt EVs, social acceptance and infrastructure will improve, fostering more widespread
  use.

### **Challenges of Electric Vehicles (EVs)**

### 1. Economic Challenges

- High Initial Cost: EVs have a high upfront cost due to expensive components and manual assembly.
- Battery Costs: Lithium-ion batteries, which make up a significant part of the EV cost, are expensive and linked to ethical concerns such as child labor in cobalt mining.
- Infrastructure Costs: Charging infrastructure, both public and private, requires significant investment. Profitability is low in developing countries like India.

### 2. Technological Challenges

- Battery Safety & Cost: Li-ion batteries are prone to safety risks like overheating and require costly maintenance.
- Energy Storage Systems: One-third of the EV cost comes from energy storage, which includes expensive materials and labor.
- Charging Technology: Slow charging times and reliance on renewable energy sources can limit adoption.

### 3. Environmental Challenges

- Greenhouse Gas Emissions: EV charging still relies on power grids that emit greenhouse gases, contributing to global warming.
- Battery Disposal: Improper disposal of EV batteries can lead to health risks and environmental harm.

### 4. Social Challenges

Consumer Attitudes: Range anxiety and reluctance to adopt new technologies are major hurdles in EV adoption.

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### **DATASET**

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