

DATA-DRIVEN INSIGHTS FOR UTILITY SERVICE MANAGEMENT: A STATISTICAL APPROACH



**Project Report
Submitted to
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Certificate

This is to certify that **Ms. A S PRAISIE JEMIMAH** and **Ms. Y Sri NIKITHA** have completed their project work as partial fulfillment of the course work of Stat 414: R-Programming (Lab based) submitted to the Department of Statistics, Pondicherry University in November 2024. They have carried out all the stages of project work right from the data collection to report making, on their own. The data that they have collected is from a real time context. No part of this work was carried out earlier in any format by any one for M.Sc./ MBA/ MTech/ etc. dissertation works or Ph.D. thesis.

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(P. Tirupathi Rao)

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Abstract

Effective utility management is integral to ensuring consistent and efficient power distribution in regions with diverse consumer demographics. This project focuses on analyzing utility management data from the Mahbubnagar Operation Circle of the Southern Power Distribution Company of Telangana Ltd. The dataset includes a variety of attributes such as consumer information, nature of supply, load, meter characteristics, and financial metrics, providing a detailed snapshot of the operational dynamics in the region. The primary objective of this study is to identify patterns, relationships, and insights that can guide better decision-making and operational optimization.

The analysis employs advanced statistical techniques, including Multiple Linear Regression (MLR) for exploring predictors of load, and Generalized Linear Models (GLM) to understand categorical variations. One-way and two-way Analysis of Variance (ANOVA) tests are used to assess the effects of categorical variables on load and financial parameters. Factor Analysis and Principal Component Analysis (PCA) simplify the complexity of the data by identifying latent dimensions, while hierarchical clustering aids in segmenting consumers based on similar attributes.

Significant findings reveal that factors such as the make and capacity of meters, along with the nature of supply, strongly influence the load and related financial metrics. The study uncovers variations among supply categories—such as agricultural, domestic, and industrial—that have a direct impact on load distribution and revenue generation. These insights are instrumental in tailoring strategies for targeted service delivery and resource allocation.

The project also highlights the importance of prescriptive statistics in modeling operational scenarios, offering actionable recommendations for improving efficiency. Predictive models for load estimation and segmentation of consumer groups enable utility managers to prioritize infrastructure upgrades and enhance customer service. The hierarchical clustering results further validate the effectiveness of data-driven approaches for identifying distinct consumer needs.

Overall, this study underscores the role of statistical and machine learning techniques in addressing key challenges in utility management. By applying these methodologies, the project demonstrates how data-driven strategies can enhance load forecasting, operational planning, and service reliability. The results are not only applicable to the study region but also serve as a model for similar utility management scenarios across other regions.

In conclusion, this project combines robust statistical methodologies with domain-specific insights to offer practical solutions for improving power distribution efficiency. The findings pave the way for further research into integrating advanced analytics with real-time operational data, ensuring sustainable and efficient utility management in the long term.

C H A P T E R 1

PREFACE

1.1. Introduction and Description of the Problem

The utility management data of Southern Power Distribution Company (SPDCL) for the Mahabubnagar operation circle, specifically the Mohamodabad section, represents a critical component of effective energy distribution and management. In today's era, ensuring uninterrupted and equitable power supply is a significant challenge for utility companies. Problems such as data inconsistency, lack of integration, and inefficient data usage in decision-making can lead to operational inefficiencies, financial losses, and consumer dissatisfaction. This study addresses these issues by analyzing the collected utility management data to enhance transparency, streamline operations, and optimize resource allocation.

1.2. Significance of the Problem Under Study

Efficient power distribution is the backbone of societal and economic development. The collected dataset has the potential to uncover patterns, improve forecasting accuracy, and identify bottlenecks in the distribution process. Addressing these problems not only benefits the utility provider but also ensures a reliable supply of electricity to consumers, fostering industrial, agricultural, and residential growth.

1.3. Relevance and Importance of the Study to Society

The insights gained from this study can significantly improve power supply reliability, reduce wastage, and enhance consumer satisfaction. In rural and semi-urban regions like Mahabubnagar, uninterrupted electricity is essential for agricultural productivity, education, and household comfort. An optimized system also reduces operational costs, making electricity affordable for the community.

1.4. Glossary

- **Nature of Supply:** Refers to the type of electricity supply, such as single-phase or three-phase.
- **Category:** Classification of consumers (e.g., domestic, industrial).
- **MF (Multiplying Factor):** A constant used to convert readings from the meter to actual usage.
- **Sec Dep (Security Deposit):** The refundable deposit paid by consumers.
- **Dev Chg (Development Charges):** One-time charges for infrastructure setup.
- **PR (Payment Receipt):** A document indicating payment confirmation.

1.5. Motivation for Selecting the Specific Topic

The increasing focus on data-driven decision-making in the power sector motivated this study. The opportunity to explore real-world data and its impact on improving electricity distribution efficiency aligns with both personal and societal interests.

1.6. Data Collection Sources

The dataset was obtained from the automated systems of the Mohammadabad section under the Mahaboobnagar operation circle of SPDCL. It includes detailed consumer information, billing details, and associated metadata from their internal system.

1.7. Data Collection Methodology

Data was systematically extracted from the SPDCL's automated system using pre-defined templates to ensure uniformity and completeness. Each variable was verified for accuracy and relevance before inclusion in the dataset.

1.8. Data Preparation Steps

- **Cleaning:** Removal of duplicate or irrelevant records.
- **Validation:** Cross-checking variables like dates and consumer numbers for consistency.
- **Coding:** Assigning numeric codes to categorical variables for analysis.
- **Integration:** Merging data from multiple sources into a unified format.

1.9. Specimen Data

Specimen data template shown here

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z	AA	AB
1	SL.NO	DATE	REGNO	CONSUMI	CONSUMER AREA	NAM	MANDAL	NATURE C	CATEGORY	LOAD	METER NO	MAKE	PHASE	CAPACITY MF	SEC	DEP	DEV CHG	APP FEE	DO NO	DO DATE	PR NO	PR DATE	DATE OF F	AREA COC	STRUCTURE	SERVICE NO	
2	1	24/06/2024	NR913244	VAJRA 5+9-16	DESI DESAI PAL	NON-DOR				9000	7795120	HPL - ELEC	3 10-40A	1	7200	10800	59	2.36E+09	24/06/2022	9.13E+10	24/06/2022	7/7/2024		1203	1.53E+09	1.2E+08	
3	3	26/07/2024	NR913244	KOTHA A1-29/1	M-MOHAMIN	MOHAMIN	NON-DOR			9000	7786410	HPL - ELEC	3 10-40A	1	7200	10800	59	2.4E+09	26/07/2022	9.13E+10	26/07/2022	31/07/2024		1207	1.53E+09	1.21E+08	
4	4	28/12/2024	NR913233	P BASAIA 15/10/03	KANCHAN	KANCHAN	NON-DOR			30000	5198813	HPL - ELEC	3 10-40A	1	24000	36000	59	9.6152	28/12/2022	9.13E+10	28/12/2022	07/6/2024		1207	1.53E+09	1.21E+08	
5	5	23/12/2024	NR913233	KALAI BA1-11/5	DESI DESAI PAL	DESAIPAL DOMESTIC				5000	8317588	HPL - ELEC	3 100/3A	1	4000	6000	59	9.6029	22/12/2022	9.13E+10	23/12/2022	27/12/2024		1238	1.53E+09	1.24E+08	
6	8	17/01/2024	NR913233	KALAI BA1-71	ANN ANNAREC	ANNAREC	AGRICULT			500	HE230921	HIMACHA	1 10-40A	1	200	600	29.5	5614	17/01/2024	9.13E+10	17/01/2024	17/01/2024		302	1.53E+09	3.02E+08	
7	9	26/03/2024	NR913243	VISLAVATI 5-1/36/1	ANN ANNAREC	ANNAREC	AGRICULT			1000	HE231221	HIMACHA	1 10-40A	1	200	1200	29.5	2.28E+09	26/03/2022	9.13E+10	26/03/2022	27/03/2024		302	1.53E+09	3.02E+08	
8	14	7/5/2024	NR913244	KATRAVA 1-61/1	ANN ANNAREC	ANNAREC	NON-DOR			500	HP240135	HPL - ELEC	1 10-40A	1	200	600	29.5	2.32E+09	7/5/2024	9.13E+10	7/5/2024	13/05/2024		302	1.53E+09	3.02E+08	
9	16	19/07/2024	NR913244	KATRAVA 1-75/1	ANN ANNAREC	ANNAREC	DOMESTIC			2000	HP240369	HPL - ELEC	1 5-30A	1	1600	2400	59	674431	18/07/2022	9.13E+10	19/07/2022	26/07/2024		302	1.53E+09	3.02E+08	
10	17	17/08/2024	NR913244	KATRAVA 1-52	ANN ANNAREC	ANNAREC	DOMESTIC			1000	HP240369	HPL - ELEC	1 5-30A	1	200	1200	29.5	305247	17/08/2022	9.13E+10	17/08/2022	19/09/2024		302	1.53E+09	3.02E+08	
11	18	20/09/2024	NR913244	SABAVATI 73	ANN ANNAREC	ANNAREC	AGRICULT			500	HP240369	HPL - ELEC	1 5-30A	1	200	600	29.5	2.44E+09	20/09/2022	9.13E+10	20/09/2022	25/09/2024		302	1.53E+09	3.02E+08	
12	19	16/10/2024	NR913244	KADAVAT 1-97/1	ANN ANNAREC	ANNAREC	AGRICULT			500	HP240369	HPL - ELEC	1 5-30A	1	200	600	29.5	2.46E+09	16/10/2022	9.13E+10	16/10/2022	24/10/2024		302	1.53E+09	3.02E+08	
13	25	23/09/2024	NR913233	MD HARSHI 1-81	CHO CHOWDAL	CHOWDAL	DOMESTIC			500	HP230535	HPL - ELEC	1 5-30A	1	200	700	29.5	5290	22/09/2022	9.13E+10	23/09/2022	30/09/2024		304	1.53E+09	3.08E+08	
14	26	21/11/2024	NR913233	PURI JI 1-19	CHO CHOWDAL	CHOWDAL	CHOWDAL	AGRICULT		2000	HE230921	HIMACHA	1 5-30A	1	1600	2400	59	5703	21/11/2022	9.13E+10	21/11/2022	26/11/2024		304	1.53E+09	3.08E+08	
15	27	21/11/2024	NR913233	PURI JI 1-19	CHO CHOWDAL	CHOWDAL	CHOWDAL	AGRICULT		500	HE230921	HIMACHA	1 5-30A	1	200	600	29.5	5685	20/11/2022	9.13E+10	21/11/2022	26/11/2024		304	1.53E+09	3.08E+08	
16	31	13/02/2024	NR913243	SABHAVA 4-19	CHO CHOWDAL	CHOWDAL	CHOWDAL	AGRICULT		1000	HE231221	HIMACHA	1 5-30A	1	200	1200	29.5	287173	13/02/2022	9.13E+10	13/02/2022	20/02/2024		304	1.53E+09	3.08E+08	
17	32	22/02/2024	NR913243	KATRAVA 4-30	CHO CHOWDAL	CHOWDAL	CHOWDAL	AGRICULT		1000	HE231221	HIMACHA	1 5-30A	1	200	1200	29.5	96105	13/02/2022	9.13E+10	22/02/2022	26/02/2024		304	1.26E+09	3.08E+08	
18	43	6/3/2024	NR913243	MUBINA E-43/AA	CHO CHOWDAL	CHOWDAL	DOMESTIC			1000	HE231221	HIMACHA	1 5-30A	1	200	1200	29.5	2.26E+09	6/3/2024	9.13E+10	6/3/2024	15/03/2024		308	1.53E+09	3.08E+08	
19	44	6/3/2024	NR913243	JOGU ALI 1-37	CHO CHOWDAL	CHOWDAL	CHOWDAL	DOMESTIC		500	HE231221	HIMACHA	1 5-30A	1	200	600	29.5	2.26E+09	6/3/2024	9.13E+10	6/3/2024	14/03/2024		308	1.53E+09	3.08E+08	
20	45	6/3/2024	NR913243	JOGU CHI 1-55	CHO CHOWDAL	CHOWDAL	CHOWDAL	AGRICULT		500	HE231221	HIMACHA	1 5-30A	1	200	600	29.5	2.26E+09	6/3/2024	9.13E+10	6/3/2024	14/03/2024		308	1.53E+09	3.08E+08	
21	46	7/5/2024	NR913244	KAVALI KI 1-10/B	CHO CHOWDAL	CHOWDAL	CHOWDAL	DOMESTIC		500	HP240135	HPL - ELEC	1 5-30A	1	200	600	29.5	3.22E+09	7/5/2024	9.13E+10	7/5/2024	13/05/2024		308	1.53E+09	3.08E+08	
22	48	14/08/2024	NR913244	VADDE RI 2-61/A	CI CHO CHOWDAL	CHOWDAL	CHOWDAL	AGRICULT		1000	HP240369	HPL - ELEC	1 5-30A	1	200	1200	29.5	2.41E+09	14/08/2022	9.13E+10	14/08/2022	23/08/2024		308	1.53E+09	3.08E+08	
23	49	14/08/2024	NR913244	MUDHAMA 3-39	CHO CHOWDAL	CHOWDAL	CHOWDAL	AGRICULT		500	HP240369	HPL - ELEC	1 5-30A	1	200	600	29.5	3.19E+09	14/08/2022	9.13E+10	14/08/2022	23/08/2024		308	1.53E+09	3.08E+08	

1.10. Preparation of Data Matrix

The data matrix was created by arranging variables as columns and consumer records as rows, ensuring every column represents a single variable, such as consumer name or service number.

1.11. Data Formatting on Excel Sheet

- Column headers were standardized to ensure clarity and brevity.
- Dates were formatted as YYYY-MM-DD.
- Numeric variables like security deposits and development charges were aligned to two decimal places.
- Categorical variables were encoded for ease of analysis.

C H A P T E R 2

Exploratory Data Analysis and Descriptive Statistics

2.1. Importance of Exploratory Data Analysis (EDA)

Exploratory Data Analysis (EDA) is a crucial step in understanding the structure, content, and patterns within a dataset. It involves summarizing the main characteristics of data using statistical and graphical techniques. The significance of EDA lies in:

- **Data Familiarization:** Gaining insights into the dataset's structure, relationships, and anomalies.
- **Identifying Trends and Patterns:** Unveiling hidden trends and associations.
- **Data Validation:** Detecting inconsistencies, missing values, and outliers.
- **Guiding Further Analysis:** Providing direction for hypothesis formulation and model building.

EDA bridges the gap between raw data and actionable insights, making it indispensable in data-driven decision-making processes.

2.2. Meaning and Scope of Descriptive Statistics

Descriptive statistics involves the quantitative summarization of data to describe its main features. It encompasses measures of central tendency, dispersion, and distribution.

Scope:

- **Summarization:** Simplifying complex datasets.
- **Comparisons:** Enabling meaningful data comparisons.
- **Visualization:** Supporting effective communication of data through charts and graphs.

2.3. Need & Significance of Descriptive Statistics

Descriptive statistics is vital for:

- **Decision Support:** Providing a solid foundation for informed decisions.
- **Data Simplification:** Reducing large datasets into comprehensible formats.
- **Performance Metrics:** Offering insights into operational efficiency.
- **Baseline for Analysis:** Acting as the starting point for inferential statistics.

2.4. Brief Statistical Description of Descriptive Statistics

Descriptive statistics can be broadly categorized into:

- Measures of Central Tendency:
 - Mean: The average value of the dataset.
 - Median: The middle value when data is ordered.
 - Mode: The most frequently occurring value.
- Measures of Dispersion:
 - Range: The difference between the maximum and minimum values.
 - Variance: The average squared deviation from the mean.
 - Standard Deviation: A measure of data spread relative to the mean.
- Shape of Distribution:
 - Skewness: Indicates asymmetry in the data distribution.
 - Kurtosis: Describes the "peakedness" or flatness of the data distribution.

2.5. Frequency Tables

Frequency tables are a structured way to represent the distribution of data across various categories or intervals. They allow for quick identification of patterns and trends in the data.

2.6. Statistical Diagrams

Statistical diagrams visually represent data, making it easier to interpret trends, relationships, and distributions.

2.7. Results

1. Summary Statistics

- Purpose: Provides an overview of each column in the dataset.
- Output Example:
 - LOAD: Numeric column with Min: 3.0, Mean: 1092, Max: 30000.
 - Represents the serial number of records.
 - Central tendency (mean and median) and spread (min, max) are provided.
 - DATE, REGI.NO, etc.: Character columns, described by their Length, Class, and Mode.
 - No numeric summary is computed for non-numeric columns.

2. Data Structure (str())

- Purpose: Displays the structure of the dataset.

Key Points:

- 'data.frame': 322 obs. of 26 variables: Indicates that the dataset has 322 rows and 26 columns.
- Shows data types (e.g., int, chr) for each column.

Examples:

- LOAD is an integer column representing some numerical metric (e.g., power load).
- CONSUMER.NAME is a character column containing consumer names.

3. Interquartile Range (IQR)

- Purpose: Measures the spread of the middle 50% of data.
- Formula: $IQR = Q3 - Q1$ $IQR = Q3 - Q1$ (difference between the third quartile and first quartile).

Example:

- LOAD: IQR is 500, meaning the middle 50% of load values lie within a range of 500 units.
- Columns with IQR = 0 indicate no variability in the middle 50% of the data (e.g., MF).

4. Quartile Deviation

- Purpose: Also known as Semi-Interquartile Range. It measures half of the IQR.
- Formula: $QD = \text{IQR}/2$

Example:

- LOAD: Quartile Deviation is 250, showing that the central data varies by 250 units from the median.

5. Range

- Purpose: Describes the difference between the maximum and minimum values.
- Formula: $\text{Range} = \text{Max} - \text{Min}$

Example:

- LOAD: Range is $30,000 - 3 = 29,997$, indicating a large variation in load values.
- CATEGORY: Range is 3, as values vary between 1 and 4.

6. Sum

- Purpose: The total sum of all numeric values in a column.

Example:

- LOAD: Sum is 351,503, representing the cumulative load across all records.

7. Mean

- Purpose: Provides the average value.
- Formula: $\text{Mean} = \text{Sum of values} / \text{Number of values}$

Example:

- LOAD: Mean is 1,091.624, indicating the average load per consumer.

8. Median

- Purpose: Middle value when data is sorted in ascending order.

Example:

- LOAD: Median is 1,000, showing the central load value across all consumers.

9. Mean Deviation

- Purpose: Measures the average deviation from the mean.

Example:

- SL.NO: Mean Deviation is 230.5443, showing a considerable spread around the average.

10. Standard Deviation

- Purpose: Indicates the dispersion of data around the mean.
- Formula: $SD = \sqrt{Var(x)}$

Example:

- LOAD: SD is 1,852.035, indicating high variability in load values.

11. Deciles

- Purpose: Divides data into 10 equal parts, each representing 10% of the data.

Example:

LOAD:

- 1st Decile (10%): 3
- 9th Decile (90%): 1,000
- Indicates that 90% of loads are $\leq 1,000$.

12. Octiles

Purpose: Divides data into 8 equal parts, each representing 12.5% of the data.

Example:

LOAD:

- 2nd Octile: 300
- 5th Octile: 1000

Represents finer division compared to deciles

13. Mode

Purpose: Most frequently occurring value(s).

Example:

- DATE: Mode is 6/3/2024, indicating the most common record date.
- CONSUMER.NAME: Mode is C PADMAMMA.

14. Covariance

Definition: Measures the degree to which two variables vary together. Unlike correlation, covariance is not standardized and depends on the scales of the variables.

Value Range:

- Positive: Variables tend to increase together.
- Negative: As one variable increases, the other tends to decrease.
- Near Zero: No relationship.

Formula:

- $\text{Cov}(X,Y) = \sum(X_i - \bar{X})(Y_i - \bar{Y}) / n - 1$

Example Analysis for NSR_M:

- Covariance between LOAD and CATEGORY:
- A large positive covariance indicates that as the CATEGORY value increases, the LOAD also tends to increase.
- Covariance magnitude depends on the units, making it less interpretable than correlation.

15. Univariate Tables

These are univariate frequency tables, where each table presents the frequency distribution of a single variable (e.g., MANDAL.NAME, NATURE.OF.SPPLY, etc.). Univariate analysis focuses on understanding the distribution and characteristics of a single variable without considering relationships between multiple variables.

1. MANDAL.NAME

- Description: This table shows the distribution of various locations (Mandals or villages) and their corresponding frequencies (number of occurrences).
- Example Interpretation:
 - "AMUDALA GADDA THANDA" appears 2 times.
 - "ANNAREDDYPALLY" appears 7 times.
 - "BAPANI KUNTA THANDA" appears 1 time, and so on.
- Insight: Some areas (e.g., "CHOWDARPALLY" with 24 occurrences) might have more records, indicating more activity or relevance in the dataset compared to others.

2. NATURE.OF.SPPLY

- Description: This table shows the type of supply for various categories such as AGRICULTURAL, DOMESTIC, etc., and the frequency with which each type occurs.
- Example Interpretation:
 - "AGRICULTURAL" is noted 63 times.
 - "DOMESTIC" is noted 205 times.
 - "GENERAL PURPOSE" appears 3 times, and so on.
- Insight: The dataset likely contains a heavy focus on domestic supply, followed by agricultural.

3. CATEGORY

- Description: The table here shows the distribution of the variable CATEGORY, which might be a classification or segmentation variable.
- Example Interpretation:
 - Category 1 appears 264 times.
 - Category 2 appears 57 times.
 - Category 4 appears once.
- Insight: The majority of the entries belong to Category 1, with Category 4 being the least frequent.

4. LOAD

- Description: This table represents the frequency distribution of different LOAD values, which likely represent energy or electrical load.
- Example Interpretation:
 - A load of 3 occurs 1 time.
 - A load of 500 occurs 91 times.
 - A load of 30000 occurs 1 time.
- Insight: The dataset contains various load values, with most occurrences at 500 and 1000.

5. MAKE

- Description: The table shows the frequency distribution of equipment or device manufacturers.
- Example Interpretation:
 - "HIMACHAL ENERGY" appears 147 times.
 - "HPL - ELECTRIC" appears 175 times.
- Insight: Both manufacturers appear frequently, with similar frequencies.

6. PHASE

- Description: This table shows the distribution of phases (likely related to some electrical or operational phase).
- Example Interpretation:
 - Phase 1 occurs 316 times.
 - Phase 3 occurs 6 times.
- Insight: Phase 1 is significantly more frequent than Phase 3.

7. CAPACITY

- Description: This table shows the distribution of the capacity of equipment, such as power rating (e.g., 10-40A, 100/5A).
- Example Interpretation:
 - 10-40A appears 6 times.
 - 5-30A appears 315 times.
- Insight: A significant number of records have a capacity of 5-30A.

8. MF

- Description: The table displays the frequency of a certain category (possibly related to some model or feature).
- Example Interpretation:
 - 1 appears 322 times.
- Insight: Only a single category (1) is present in this variable.

9. SEC.DEP

- Description: This table shows the frequency of different security deposit amounts.
- Example Interpretation:
 - 200 appears 258 times.
 - 400 appears 2 times, and so on.
- Insight: A majority of the records have a security deposit of 200, with other values appearing far less frequently.

10. DEV.CHG

- Description: This table shows the frequency of development charges (likely related to infrastructure or utility charges).
- Example Interpretation:
 - 600 appears 80 times.
 - 708 appears 11 times, and so on.
- Insight: The development charge 600 is the most frequent, with other charges appearing less often.

11. APP.FEE

- Description: This table shows the application fee amounts.
- Example Interpretation:
 - 29.5 appears 264 times.
 - 59 appears 58 times.
- Insight: The application fee of 29.5 is more common than 59.

12. PR.DATE

- Description: This table shows the frequency of various PR.DATE (possibly the date of registration or application).
- Example Interpretation:
 - 1/11/2023 occurs 8 times.
 - 1/3/2024 occurs 2 times, and so on.
- Insight: Most records are from specific dates like 1/11/2023.

13. AREA.CODE

- Description: This table shows the frequency of various area codes (likely indicating geographical regions).
- Example Interpretation:
 - Area code 302 appears 7 times.
 - Area code 308 appears 11 times, and so on.
- Insight: The most frequent area codes are 308 and 309.

14. STRCUTURE.CODE

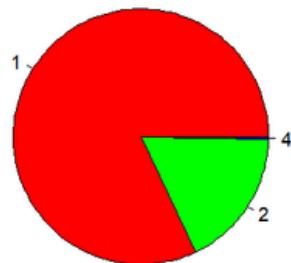
- Description: This table shows the frequency of different structure codes, possibly relating to infrastructure or building types.
- Example Interpretation:
 - 1257340010 appears 1 time.
 - 1527002006 appears 6 times, and so on.
- Insight: Different structure codes appear at varying frequencies, with some being quite rare.

16. STATISTICAL DIAGRAMS

PIE CHARTS

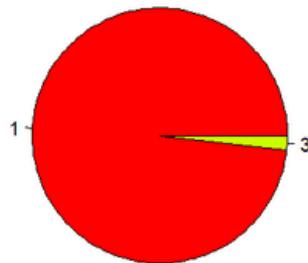
- Below is the pie chart based on the distribution of the CATEGORY column in the dataset NSR_M. (*refer annexure for code*)

PIE CHART OF CATEGORY

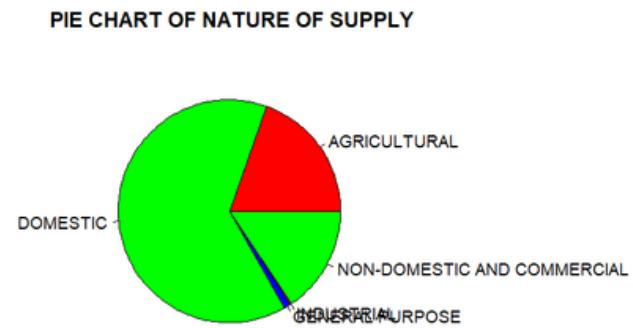


- Below is a pie chart for the PHASE column in the dataset. (*refer annexure for code*)

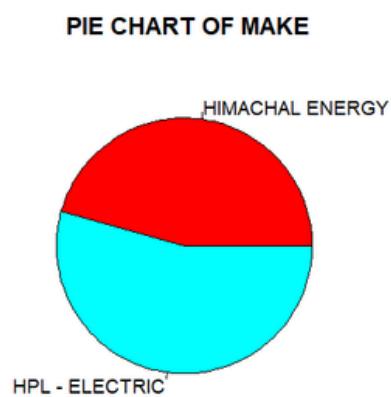
PIE CHART OF PHASE



- Below is the pie chart for the NATURE.OF.SPPLY column. (*refer annexure for code*)

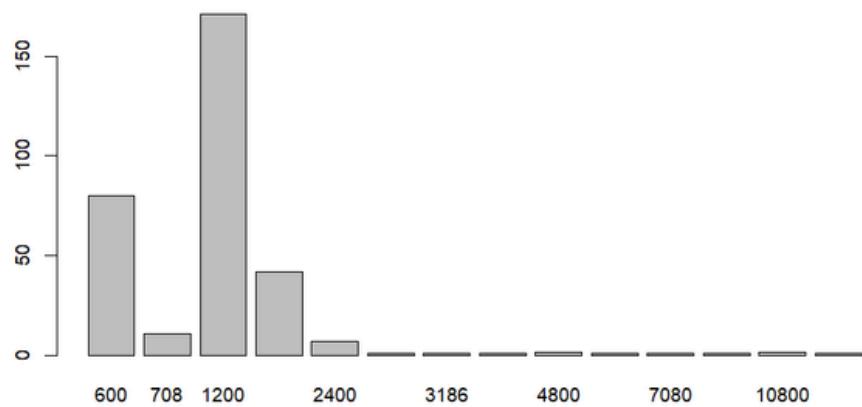


- Below is the pie chart for the MAKE column. (*refer annexure for code*)

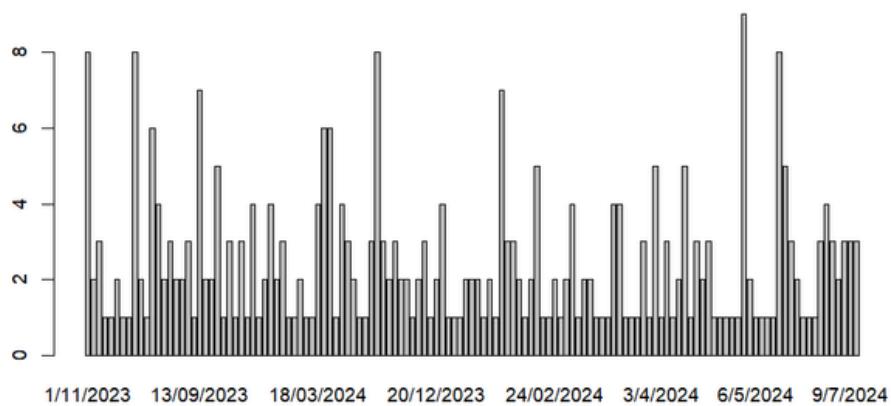


BAR PLOTS

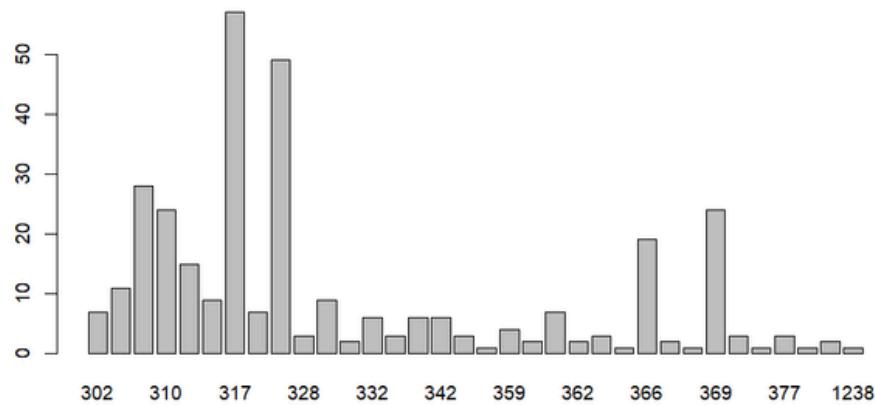
- Purpose: This bar plot shows the frequency of each unique value in the DEV.CHG column.
- Example: If DEV.CHG contains categories like "Improved," "Unchanged," and "Worsened," the bar plot will show the count of rows for each of these values as bars. (*Refer annexure for code*)



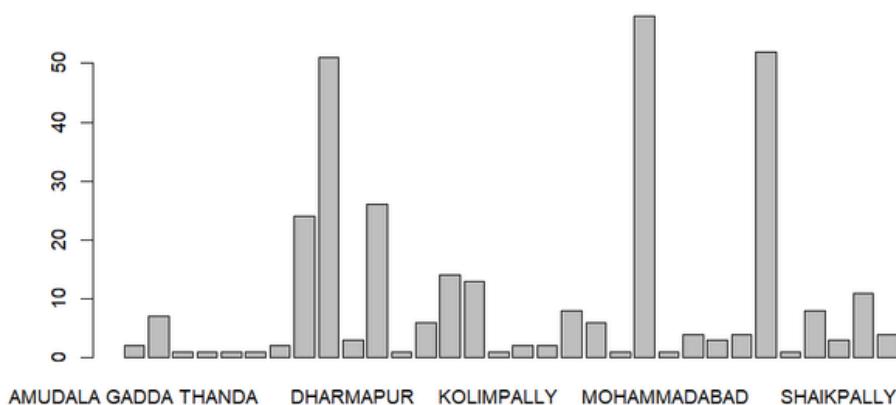
- Purpose: This bar plot visualizes the frequency of records for each unique date in the DATE column.
- Example: If DATE includes values like "2024-01-01," "2024-01-02," etc., the bar plot will have bars corresponding to how many records fall on each date.
- Caution: If the dataset contains many unique dates, the bar plot may become cluttered. (*Refer annexure for code*)



- Purpose: This bar plot represents the frequency of each unique area code in the AREA.CODE column.
- Example: If AREA.CODE includes values like "123," "456," and "789," the bar plot will show how often each area code appears. (*Refer annexure for code*)



- Purpose: This bar plot displays the frequency of occurrences of each unique mandal name in the MANDAL.NAME column.
- Example: If MANDAL.NAME contains names like "Mandal1," "Mandal2," and "Mandal3," the bar plot will depict how often each name is found in the dataset. (*Refer annexure for code*)



C H A P T E R 3

Inferential & Relational Data Analysis

3.1. Predictive Statistics

Predictive statistics is a branch of statistical analysis that focuses on using historical data to predict future outcomes. It employs mathematical models, algorithms, and statistical techniques to identify patterns in data.

Relevance in Utility Management:

- Helps forecast electricity demand based on past consumption patterns.
- Assists in load management by predicting peak and off-peak hours.
- Supports decision-making in resource allocation and infrastructure planning.

3.2. Inferential Data Analysis

Inferential statistics allow us to draw conclusions about a population based on sample data. By analyzing data collected from a subset of consumers, you can generalize findings to the entire population.

Importance in Utility Management:

- Decision Support: Provides evidence-based decisions, e.g., whether new infrastructure is required.
- Significance Testing: Determines if differences in energy consumption across categories or areas are statistically significant.

3.3. Relational Data Analysis : Correlation and Regression

Correlation Analysis: Measures the strength and direction of relationships between variables.

- Example: Load vs. Category or Nature of Supply vs. Capacity.
- Use correlation coefficients (e.g., Pearson's r) to quantify relationships.

Regression Analysis: Explores how dependent variables (e.g., energy consumption) are influenced by independent variables (e.g., category, load, phase).

- Simple Regression: Predicting load based on category.
- Multiple Regression: Using multiple factors like area, mandal, and capacity to predict load.

3.4. Parameter Estimation in Regression

In regression models, parameter estimation involves calculating the coefficients (e.g., slope, intercept) that best describe the relationship between variables.

Steps for Estimation:

1. Define the regression equation
2. Use methods like Ordinary Least Squares (OLS) to minimize error and find the coefficients

Utility Management Application:

- Estimating the effect of phase or load on consumption to recommend infrastructure upgrades.

3.5. Estimation of Correlation Parameters

Correlation Coefficient Estimation: Calculates how strongly two variables are related.

- Pearson Correlation (r): Measures linear relationships between variables like load and capacity.
- Spearman's Rank Correlation: Useful for non-linear relationships, e.g., between mandal name and load.

Applications:

- Understanding if high-capacity meters are correlated with specific areas or categories.
- Identifying key drivers of energy usage.

3.6. Parametric Hypothesis Tests

Parametric tests assume data follows a specific distribution (e.g., normal distribution). Common tests include:

- Single Sample t-Test: Compare the mean load of a single category to a hypothesized value.
- Two-Sample t-Test: Compare average loads between two categories or mandals.

Examples in Dataset:

- Test if the average load differs between single-phase and three-phase connections.
- Compare service charges across different mandals.

3.7. Non-Parametric Hypothesis Tests

Non-parametric tests do not require data to follow a specific distribution. Common tests include:

- Mann-Whitney U Test: Compare two independent groups, e.g., load in rural vs. urban areas.
- Wilcoxon Signed-Rank Test: Compare paired data, e.g., before and after introducing a new tariff.

Utility Management Example:

- Analyze whether consumer satisfaction levels differ across mandals without assuming normality.

3.8. Results

Predictive Statistics using Arima model

This table represents the point forecasts and the confidence intervals for future values of the "category" variable based on an ARIMA model. Here's an explanation of each column:

1. Point Forecast:

- This is the predicted value for each month, based on the ARIMA model. For example, the forecast for Nov 27 is 1.166008. This is the expected value of the "category" variable for that specific time period.

2. Lo 80 and Hi 80:

- These columns represent the 80% confidence interval for the point forecast. The Lo 80 value is the lower bound, and the Hi 80 value is the upper bound of the confidence interval, with 80% confidence that the true value will fall within this range.
- For Nov 27, the 80% confidence interval is from 0.6435076 (Lo 80) to 1.688508 (Hi 80).

3. Lo 95 and Hi 95:

- These columns represent the 95% confidence interval for the point forecast. The Lo 95 value is the lower bound, and the Hi 95 value is the upper bound of the confidence interval, with 95% confidence that the true value will fall within this range.
- For Nov 27, the 95% confidence interval is from 0.3669126 (Lo 95) to 1.965103 (Hi 95).

Example for November 2027:

- The forecast value is 1.166008 for Nov 27.
- The 80% confidence interval suggests the true value will be between 0.6435076 and 1.688508.
- The 95% confidence interval suggests the true value will be between 0.3669126 and 1.965103.

Simple linear regression model (lm) to predict the "LOAD" variable based on the "MAKE" variable.

Model Equation:

The formula for the regression model is:

$$\text{LOAD} = \beta_0 + \beta_1 \times \text{MAKE}$$

Where:

- Intercept (750): The predicted load when "MAKE" is not specified (i.e., the baseline value when "MAKE" is at the reference category). This means the model predicts a load of 750 when the "MAKE" is not "HPL - ELECTRIC" (the reference level).
- MAKEHPL - ELECTRIC (12,500): This coefficient represents the change in the predicted load when "MAKE" is "HPL - ELECTRIC", compared to the baseline category. In this case, the predicted load for "HPL - ELECTRIC" is 12,500 more than the baseline load of 750.

Coefficients:

- (Intercept): 750 (baseline load).
- MAKEHPL - ELECTRIC: 12,500 (increase in load for the "HPL - ELECTRIC" make compared to the reference category).

2. Predicted Values:

You used the model to make predictions for two different values of the "MAKE" variable:

- For "MAKE" = "HPL - ELECTRIC": The predicted load is 13,250. This is calculated as:
$$750 + 12,500 = 13,250$$
- So when the "MAKE" is "HPL - ELECTRIC", the model predicts a load of 13,250.
- For "MAKE" = "HIMACHAL ENERGY": The predicted load is 750. Since "HIMACHAL ENERGY" is likely the reference category for "MAKE", the predicted load is simply the intercept value of 750.

Simple linear regression model (lm) without intercept

The formula for the regression model is:

$$\text{LOAD} = \beta_1 \times \text{MAKEHIMACHAL ENERGY} + \beta_2 \times \text{MAKEHPL - ELECTRIC}$$

Where:

- The model is forced through the origin (no intercept) since you used $0 + \text{MAKE}$. This means that the regression will estimate the load directly based on the levels of "MAKE", treating each level as a separate variable.

2. Coefficients:

- **MAKEHIMACHAL ENERGY (750)**: The predicted load when "MAKE" is "HIMACHAL ENERGY". Since there is no intercept, this coefficient represents the exact predicted value for this category.
- **MAKEHPL - ELECTRIC (13,250)**: The predicted load when "MAKE" is "HPL - ELECTRIC". This is the coefficient directly associated with this category, showing how much load corresponds to this make.

3. Model Interpretation:

- "**MAKEHIMACHAL ENERGY- "**MAKEHPL - ELECTRIC****

In this model, both "MAKEHIMACHAL ENERGY" and "MAKEHPL - ELECTRIC" are treated as separate predictors, without a baseline category, so the predicted load is equal to the coefficient value for each specific "MAKE".

4. Why " $0 + \text{MAKE}$ " was used:

- Using $0 + \text{MAKE}$ forces the model to exclude the intercept. Normally, a model includes an intercept, and one category would act as the baseline (reference category), with the other coefficients representing the difference from that reference.
- When you specify $0 + \text{MAKE}$, you effectively treat each category of "MAKE" as an independent variable with its own coefficient. Therefore, there's no "reference" category anymore, and the coefficients represent the actual predicted value for each "MAKE".

5. Practical Use:

- This model is useful when you want to treat each level of the "MAKE" variable as an individual predictor, without comparing them to a baseline category.
- It provides the direct predicted values for each category, but with no overall baseline, which may not always be appropriate depending on your analysis needs.

Conclusion:

- MAKEHIMACHAL ENERGY leads to a predicted load of 750.
- MAKEHPL - ELECTRIC leads to a predicted load of 13,250.

The key difference between this model and one with an intercept is that here, each "MAKE" category directly has its own estimated load without any reference point.

Linear regression model (lm) to predict DEV.CHG (likely the development charge) based on the AREA.CODE variable

The model equation for DEV.CHG based on AREA.CODE is:

$$\text{DEV.CHG} = \beta_0 + \beta_1 \times \text{AREA.CODE1207} + \beta_2 \times \text{AREA.CODE1238} + \beta_3 \times \text{AREA.CODE302}$$

- Intercept (10,800): This is the baseline value for DEV.CHG when the AREA.CODE is not one of the specific categories included in the model (i.e., the default or reference category).
- AREA.CODE1207 (12,600): This coefficient represents the change in DEV.CHG when AREA.CODE is 1207, compared to the baseline reference category. In other words, if the AREA.CODE is 1207, the development charge will be 12,600.
- AREA.CODE1238 (-4,800): This coefficient represents the change in DEV.CHG when AREA.CODE is 1238, compared to the reference category. If the AREA.CODE is 1238, the development charge will be -4,800.
- AREA.CODE302 (-9,900): This coefficient represents the change in DEV.CHG when AREA.CODE is 302, compared to the reference category. If the AREA.CODE is 302, the development charge will be -9,900.

2. Predicted Values:

You used the model to predict DEV.CHG for a new data point where AREA.CODE = 302.

- For AREA.CODE = 302, the predicted value of DEV.CHG is 900. This is calculated as the intercept (10,800) minus the coefficient for AREA.CODE302 (-9,900): Predicted $\text{DEV.CHG} = 10,800 + (-9,900) = 900$

Parametric Hypothesis Tests

1. One Sample t-tests

These tests are comparing the mean of a sample to a hypothesized value.

- LOAD: The mean of LOAD (1,091.62) is significantly different from 10, with a t-statistic of 10.48 and a p-value < 2.2e-16. The confidence interval for the true mean is between 888.57 and 1,294.68.
- SEC.DEP: The mean of SEC.DEP (510.25) is significantly different from 12, with a t-statistic of 5.80 and a p-value of 1.56e-08. The confidence interval for the true mean is between 341.34 and 679.15.
- APP.FEE: The mean of SEC.DEP is significantly different from 0, with a t-statistic of 5.94 and a p-value of 7.28e-09. The confidence interval for the true mean is between 341.34 and 679.15.

2. Two Sample t-tests

These tests compare the means of two groups.

- SEC.DEP vs DEV.CHG: The means of SEC.DEP (510.25) and DEV.CHG (1,356.39) are significantly different with a t-statistic of -5.60 and a p-value of 3.33e-08. The confidence interval for the difference in means is between -1,142.89 and -549.40.
- LOAD by PHASE: The mean LOAD for Phase 1 (903.49) is significantly different from Phase 3 (11,000.00), with a t-statistic of -2.61 and a p-value of 0.04785. The confidence interval for the difference is between -20,052.30 and -140.72.
- CATEGORY by PHASE: The mean CATEGORY for Phase 1 (1.17) is significantly different from Phase 3 (2.00), with a t-statistic of -36.71 and a p-value < 2.2e-16. The confidence interval for the difference is between -0.87 and -0.78.

3. Proportion Tests

These tests assess if the proportion in a given category is different from a hypothesized value.

- NATURE.OF.SPPLY = "Domestic": The proportion of "Domestic" nature of supply is 0 (none in the sample), significantly different from 0.3, with a p-value < 2.2e-16.
- NATURE.OF.SPPLY = "AGRICULTURAL": The proportion of "Agricultural" nature of supply (0.1957) is not significantly different from 0.2, with a p-value of 0.9002. The 95% confidence interval is between 0.1546 and 0.2441.
- MAKE = "HPL - ELECTRIC": The proportion of "HPL - ELECTRIC" make (0.5435) is significantly different from 0.2, with a p-value < 2.2e-16. The 95% confidence interval is between 0.4873 and 0.5986.
- MAKE = "HIMACHAL": The proportion of "Himachal" make is 0 (none in the sample), significantly different from 0.7, with a p-value < 2.2e-16.
- MANDAL.NAME: The proportion of "MOHAMMADABAD" vs "MOKARLABAD" is compared, yielding a p-value of 0.4017. There is no significant difference between the two mandals.

4. Variance Tests

These tests compare the variances of two groups.

- LOAD vs SEC.DEP: The ratio of variances (1.45) is significantly different from 1 with a p-value of 0.001012.
- LOAD vs DEV.CHG: The ratio of variances (0.69) is significantly different from 1 with a p-value of 0.000895.
- DEV.CHG vs SEC.DEP: The ratio of variances (2.10) is significantly different from 1 with a p-value of 5.60e-11.
- DEV.CHG vs APP.FEE: The ratio of variances (38,601.74) is significantly different from 1 with a p-value < 2.2e-16.
- LOAD by PHASE: The ratio of variances (0.0018) is significantly different from 1 with a p-value < 2.2e-16.
- SEC.DEP by PHASE: The ratio of variances (0.00296) is significantly different from 1 with a p-value < 2.2e-16.

5. Correlation and Correlation Test

Correlation

- Definition: Measures the strength and direction of a linear relationship between two numeric variables.
- Value Range:
 - +1+1+1: Perfect positive correlation (as one variable increases, the other also increases).
 - -1-1-1: Perfect negative correlation (as one variable increases, the other decreases).
 - 000: No linear relationship.
- Formula:
- $r = \text{Cov}(X, Y) / \sigma_X \sigma_Y$
- Where:
 - Cov(X, Y): Covariance between X and Y.
 - σ_X, σ_Y : Standard deviations of X and Y.
- Example Analysis for NSR_M:
 - Correlation between LOAD and CATEGORY:
 - If $r \approx 0.8$ or $r \approx 0.8$: Strong positive relationship (e.g., higher categories might correspond to higher loads).
 - If $r \approx -0.3$ or $r \approx -0.3$: Weak negative relationship.

Correlation tests examine the relationship between two continuous variables.

- SEC.DEP and DEV.CHG: The correlation is 0.98, with a p-value < 2.2e-16, indicating a solid positive relationship.
- SEC.DEP and APP.FEE: The correlation is 0.42, with a p-value of 3.53e-15, indicating a moderate positive relationship.
- APP.FEE and DEV.CHG: The correlation is 0.31, with a p-value of 1.94e-08, indicating a moderate positive relationship.

Non -Parametric Hypothesis Tests

1. Runs Test for Randomness

- Purpose: Tests whether a sequence of data points is random or follows a pattern.
- How It Works: Counts the number of runs (sequences of increasing or decreasing values) in the data. Deviations from the expected number of runs suggest non-randomness.
- NSR_M\$LOAD (*refer annexure for code*):
- Statistic = -2.8373, p-value = 0.004549
- Conclusion: Evidence of non-randomness.
- NSR_M\$SEC.DEP and NSR_M\$APP.FEE (*refer annexure for code*):
- Statistic = NaN, p-value = NA (implies all observations are identical or there is no variation in the data).
- NSR_M\$DEV.CHG (*refer annexure for code*):
- Statistic = -6.0231, p-value = 1.711e-09
- Conclusion: Evidence of non-randomness.

2. Wilcoxon Rank Sum Test

- Purpose: Compares the distributions of two independent groups when normality cannot be assumed.
- How It Works: Ranks all observations, then compares the sums of the ranks between groups.
- LOAD ~ PHASE (*refer annexure for code*):
- W = 0, p-value = 4.266e-07
- Conclusion: Significant difference in LOAD across phases.
- SEC.DEP ~ PHASE (*refer annexure for code*):
- W = 0, p-value = 1.633e-09
- Conclusion: Significant difference in SEC.DEP across phases.

3. Sign Test for Median

- Purpose: Tests whether the median of a dataset is equal to a specified value.
- How It Works: Counts the number of observations above and below the hypothesized median. The test is based on the binomial distribution.

NSR_M\$SEC.DEP (Tested median = 15) (*refer annexure for code*):

- $s = 322$, p-value < 2.2e-16
- 95% CI: [200, 200]
- Conclusion: The true median is significantly different from 15, estimated median = 200.
- NSR_M\$APP.FEE (Tested median = 1000) (*refer annexure for code*):
- $s = 0$, p-value < 2.2e-16
- 95% CI: [29.5, 29.5]
- Conclusion: The true median is significantly different from 1000, estimated median = 29.5.

Regression Coefficients

Regression coefficients of a simple linear regression model where LOAD is the dependent variable and MAKE is the independent variable.

Explanation of Findings:

- Intercept ((Intercept)):
- Estimate: 750
- This is the baseline value of LOAD when MAKE is at its reference category (assumed to be the default level of MAKE in the dataset, such as the first alphabetical level).
- Standard Error: 6936.092
- Indicates the variability of the intercept estimate.
- t-value: 0.1081
- Measures how far the intercept is from 0 in terms of its standard error.
- p-value: 0.9191
- Not statistically significant ($p > 0.05$), meaning the intercept does not strongly suggest a baseline difference.

- Effect of MAKE (MAKEHPL - ELECTRIC):
- Estimate: 12,500
- This shows that the LOAD increases by 12,500 units when MAKE is "HPL - ELECTRIC" compared to the reference category.
- Standard Error: 8494.943
- Represents the variability of this coefficient estimate.
- t-value: 1.4715
- Measures how far the coefficient is from 0 in terms of its standard error.
- p-value: 0.2151
- Not statistically significant ($p > 0.05$), suggesting that the difference in LOAD associated with MAKE = "HPL - ELECTRIC" is not strong enough to rule out randomness.

C H A P T E R 4

Advanced Data Analysis

4.1. Tests for More Than Two Samples

These tests are used when comparing more than two groups to determine if there are statistically significant differences among them.

Types of Tests:

- Parametric Tests:
 - One-way ANOVA: Analyzes differences among group means for a single factor (e.g., comparing loads across multiple areas).
 - Two-way ANOVA: Analyzes the effect of two independent factors (e.g., category and phase) and their interaction on the dependent variable.
- Non-parametric Tests:
 - Kruskal-Wallis Test: Non-parametric alternative to one-way ANOVA, useful when data doesn't meet normality assumptions.
 - Friedman Test: Non-parametric alternative for repeated measures data.

Applications in Utility Management:

- Comparing mean energy consumption across multiple mandals.
- Analyzing the impact of different phases and categories on service charges.

4.2. One-Way and Two-Way Analysis of Variance

One-Way ANOVA:

- Purpose: Compare the means of more than two groups categorized by a single factor.
- Example: Analyzing if average load differs across three categories (e.g., domestic, commercial, industrial).

Steps:

1. Define null hypothesis (H_0): All group means are equal.
2. Calculate the F-statistic: Ratio of variance between groups to variance within groups.
3. Evaluate significance: Use p-values to reject or fail to reject H_0 .

Two-Way ANOVA:

- Purpose: Determine the effect of two factors and their interaction on the dependent variable.
- Example: Assessing how phase (single or three) and area (urban or rural) affect energy consumption.

Steps:

- Include two factors and interaction terms in the model.
- Analyze individual and combined effects on the outcome variable.
- Use visual tools like interaction plots for interpretation.

4.3. Multiple Linear Regression (MLR)

MLR models the relationship between one dependent variable and multiple independent variables. It helps to identify how changes in predictors influence the outcome.

General Form:

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_n X_n + \epsilon$$

Applications in Utility Management:

- Predicting load based on consumer category, phase, and capacity.
- Evaluating how various factors (e.g., mandal name, nature of supply) influence service charges.

Steps:

1. Identify dependent (e.g., load) and independent variables (e.g., category, phase).
2. Fit the regression model using least squares estimation.
3. Assess model fit using metrics like R², adjusted R², and p-values.

4.4. Estimation and Testing of Parameters in MLR

Parameter Estimation:

- Coefficients (β): Estimate the effect of each independent variable on the dependent variable.
- Use methods like Ordinary Least Squares (OLS) to minimize residuals and determine parameter values.

Testing Parameters:

- Hypothesis Tests:
 - Null Hypothesis (H_0): Parameter (β_i) is equal to zero (no effect).
 - Perform t-tests for each coefficient to check significance.
- Confidence Intervals: Provide a range of plausible values for each parameter estimate.

Diagnostics:

- Check for assumptions like linearity, normality, and homoscedasticity (equal variance).
- Use residual plots and variance inflation factors (VIF) to identify multicollinearity or violations.

Example in Utility Data:

- Estimate the impact of capacity and mandal name on service charges.
- Test if load significantly varies with consumer category after controlling for other factors.

4.5. Results

One-Way ANOVA

1. First ANOVA Model:

- Objective: This model tests whether the LOAD varies significantly based on the different categories in the NATURE.OF.SPPLY (Nature of Supply) variable.

ANOVA Output (*refer annexure for code*):

- Sum of Squares:
 - NATURE.OF.SPPLY: 46,834,598
 - Residuals (Error): 1,054,206,221
- Degrees of Freedom:
 - NATURE.OF.SPPLY: 4
 - Residuals: 317
- Residual Standard Error: 1,823.615
- F-statistic: 3.521
- p-value: 0.00788

Summary Interpretation:

- F value: The F-value of 3.521 indicates that the variation in LOAD due to different NATURE.OF.SPPLY categories is greater than the variation due to residuals (errors).
- p-value: The p-value of 0.00788 is less than 0.05, which means the difference in LOAD across different categories of NATURE.OF.SPPLY is statistically significant. Thus, we can reject the null hypothesis that the means are equal and conclude that the nature of supply affects the LOAD.

2. Second ANOVA Model:

- Objective: This model tests whether LOAD varies significantly based on the different categories in the MAKE (Manufacturer) variable.

ANOVA Output (*refer annexure for code*):

- Sum of Squares:
 - MAKE: 9,791,448
 - Residuals (Error): 1,091,249,371
- Degrees of Freedom:
 - MAKE: 1
 - Residuals: 320
- Residual Standard Error: 1,846.66
- F-statistic: 2.871
- p-value: 0.0911

Summary Interpretation:

- F value: The F-value of 2.871 indicates some difference in LOAD between the levels of MAKE, but it is not as strong as the first model.
- p-value: The p-value of 0.0911 is greater than 0.05, suggesting that the difference in LOAD based on the MAKE categories is not statistically significant at the 5% level. This means we fail to reject the null hypothesis and conclude that MAKE does not significantly affect LOAD.

One-Way ANOVA to compare the means of different categories of NATURE.OF.SPPLY (which represents different types of supply, such as Domestic, Industrial, Agricultural, etc.) on two dependent variables: LOAD and SEC.DEP.

Here's a breakdown of the output:

ANOVA Summary:

- Sum of Squares: Measures the variability. The between-group sum of squares is 1.528e+08, and the within-group sum of squares (Residuals) is 3.504e+09.
- Degrees of Freedom:
 - Between groups: 4 (there are 5 groups, hence $df = 5-1 = 4$).
 - Within groups: 317 (the total number of observations minus the number of groups).
- F-Statistic: The F-value is 3.456, which is the ratio of the variance between the groups to the variance within the groups.
- p-value: The p-value for the ANOVA test is 0.00879. Since this value is less than the significance level of 0.05, you can reject the null hypothesis and conclude that there is a significant difference in the means of LOAD and SEC.DEP across different NATURE.OF.SPPLY categories.

Tukey's Honest Significant Difference (HSD) Test:

This test is used to perform pairwise comparisons between the groups to identify which specific groups differ from each other.

- Key Columns in the output:
 - diff: The difference in means between the two groups being compared.
 - lwr: The lower bound of the confidence interval for the difference.
 - upr: The upper bound of the confidence interval for the difference.
 - p adj: The adjusted p-value for the pairwise comparison.

Interpretation of Tukey's HSD results:

- Significant Differences (p adj < 0.05):
 - NON-DOMESTIC AND COMMERCIAL-AGRICULTURAL vs. AGRICULTURAL: The mean difference is 2057.49, with a p-value of 0.0105, which is significant.
 - NON-DOMESTIC AND COMMERCIAL-DOMESTIC vs. DOMESTIC: The mean difference is 1827.64, with a p-value of 0.0051, which is also significant.
- Non-Significant Differences (p adj > 0.05):
 - Comparisons between DOMESTIC vs AGRICULTURAL, GENERAL PURPOSE vs AGRICULTURAL, and others show no significant difference based on their p-values being greater than 0.05.

Conclusion:

- The One-Way ANOVA has shown that there are significant differences in the means of LOAD and SEC.DEP across the different categories of NATURE.OF.SPPLY.
- The Tukey HSD test further highlights which pairs of categories are significantly different from each other. For example, Non-Domestic and Commercial supply is significantly different from Agricultural and Domestic supply types.

This analysis is helpful to understand how the LOAD and SEC.DEP values vary according to the type of supply (NATURE.OF.SPPLY).

Two-Way ANOVA

The model is also testing for an interaction effect between these two factors (i.e., whether the effect of one factor depends on the level of the other factor).

Key Results from the Output:

1. Sum of Squares:

- MAKE: 9,791,448
- NATURE.OF.SPPLY: 48,530,877
- MAKE:NATURE.OF.SPPLY (Interaction): 36,312,765
- Residuals (the unexplained variance): 1,006,405,730

2. Degrees of Freedom:

- MAKE: 1 degree of freedom (because it is a single categorical factor with two levels).
- NATURE.OF.SPPLY: 4 degrees of freedom (assuming there are five levels in this factor).
- MAKE:NATURE.OF.SPPLY (Interaction): 2 degrees of freedom.
- Residuals: 314 degrees of freedom (based on the sample size and number of factors).

3. Mean Squares:

- Calculated by dividing the Sum of Squares by the degrees of freedom for each factor.
- These values represent the average amount of variation attributable to each source.

4. F-Statistic:

- MAKE: 3.055
- NATURE.OF.SPPLY: 3.785
- MAKE:NATURE.OF.SPPLY (Interaction): 5.665
- F-statistics compare the variation between groups to the variation within groups.

5. P-Values:

- MAKE: 0.08147 (indicating that MAKE is not statistically significant at the 0.05 level, but is borderline significant at 0.1).
- NATURE.OF.SPPLY: 0.00506 (indicating that NATURE.OF.SPPLY is statistically significant at the 0.01 level).
- MAKE:NATURE.OF.SPPLY (Interaction): 0.00383 (indicating that the interaction effect between MAKE and NATURE.OF.SPPLY is statistically significant).

Interpretation of Results:

- Main Effect of MAKE: The MAKE factor has a p-value of 0.08147, which suggests it is not statistically significant at the 0.05 level, but it could be significant at the 0.1 level. This means that the different MAKE levels might have a small impact on LOAD, but this effect is not strong enough to reject the null hypothesis at the typical 5% significance level.
- Main Effect of NATURE.OF.SPPLY: The NATURE.OF.SPPLY factor has a p-value of 0.00506, which is less than 0.01. This indicates that the different NATURE.OF.SPPLY levels have a significant effect on LOAD, meaning that the type of supply category has a clear influence on the outcome.
- Interaction Effect (MAKE:NATURE.OF.SPPLY): The interaction term has a p-value of 0.00383, which is significant ($p < 0.05$). This suggests that the effect of MAKE on LOAD depends on the level of NATURE.OF.SPPLY. In other words, the influence of MAKE might vary depending on the type of supply, and there is a significant interaction between the two factors.

Logistic Regression

GLMs extend traditional linear regression to allow for a broader range of response variable distributions and relationships between the predictors and the response variable. They are used when the dependent variable does not follow a normal distribution or when the relationship between the predictors and the response is non-linear.

1. Model 1:

This model includes an intercept and the independent variable LOAD.

Key Results (*refer annexure for code*):

- Coefficients:
 - Intercept: 1.1213841 (This represents the baseline value of the log-odds of the dependent variable when LOAD = 0).
 - LOAD: 0.0000595 (This indicates the change in the log-odds of CATEGORY for each unit increase in LOAD).
- Deviance:
 - Null Deviance: 54.82 (The deviance of the model with only the intercept, representing the total variability in the data).
 - Residual Deviance: 50.92 (The deviance of the fitted model, representing the variability left unexplained by the model).
- AIC: 325.9 (The Akaike Information Criterion, used for model selection. Lower values indicate better model fit).

Interpretation:

- The model is suggesting that LOAD has a very small impact on the dependent variable CATEGORY, as indicated by the small coefficient value (0.0000595).
- A reduction in deviance from 54.82 (null model) to 50.92 (fitted model) suggests that including LOAD explains some variability in CATEGORY, but the effect is relatively weak.

2. Model 2:

This model excludes the intercept (specified by `0 + LOAD`), meaning the relationship is forced to pass through the origin.

Key Results (*refer annexure for code*):

- Coefficients:
 - `LOAD`: 0.000325 (This represents the relationship between `CATEGORY` and `LOAD` when no intercept is included).
- Deviance:
 - Null Deviance: 508 (Larger than Model 1, likely because excluding the intercept increases the overall variability in the null model).
 - Residual Deviance: 351.2 (Indicates the variability left unexplained by the model).
- AIC: 945.7 (Much higher than Model 1, suggesting poorer model fit compared to the first model).

Interpretation:

- The absence of the intercept forces the model to assume that when `LOAD` = 0, the predicted value of `CATEGORY` is also 0, which might not align well with the data.
- The higher AIC and residual deviance indicate that this model fits the data worse than Model 1.

Comparison of Models:

- Model 1 (with an intercept) performs better than Model 2 (no intercept), as indicated by lower AIC and residual deviance.
- The relationship between `LOAD` and `CATEGORY` appears weak in both models, given the small coefficients and minimal reduction in deviance.

Multiple Linear Regression

The model:

- LOAD: The dependent variable being modeled.
- MAKE, PHASE, CAPACITY: Independent variables explaining variations in LOAD.

Coefficients (*refer annexure for code*):

(Intercept): 750

MAKEHPL - ELECTRIC: 15250

PHASE: NA

CAPACITY100/5A: -11000

- Intercept (750): Baseline value of LOAD when all predictors are at their reference levels.
- MAKEHPL - ELECTRIC (15250): Adjusts the baseline by adding 15250 when the "MAKE" is "HPL - ELECTRIC."
- PHASE (NA): Indicates an issue, possibly due to perfect multicollinearity or lack of variation in the "PHASE" variable.
- CAPACITY100/5A (-11000): Adjusts LOAD by subtracting 11000 when "CAPACITY" is "100/5A."

3. Predictions

Using predict()*(Refer annexure for code)*:

- Prediction for MAKE="HPL" - ELECTRIC", PHASE=1, CAPACITY="100/5A" is 5000.
- Formula: $\text{LOAD}=750+15250-11000=5000$

For:

- Prediction for MAKE="HIMACHAL ENERGY", PHASE=1, CAPACITY="10-40A" is 750, as neither PHASE nor CAPACITY contribute significantly (or are missing). (*refer annexure for code*)
- Prediction from rank-deficient fit; attr(*, "non-estim") has doubtful cases
- This occurs due to collinearity or insufficient variation among predictors, causing some coefficients to become indeterminate (e.g., PHASE = NA).

5. Model Without Intercept

- Removes the intercept term, allowing coefficients to represent the absolute contribution of each level.
- Coefficients now include (*refer annexure for code*):

MAKEHIMACHAL ENERGY: 750

MAKEHPL - ELECTRIC: 16000

PHASE: NA

CAPACITY100/5A: -11000

This approach assigns absolute values to categories of "MAKE" while excluding a baseline reference level.

Factor Analysis, PCA, and Hierarchical Clustering

1. Factor Analysis

Purpose:

Factor Analysis identifies underlying latent variables (factors) that explain correlations among observed variables.

Results Interpretation:

- Key Variables: LOAD, SEC.DEP, DEV.CHG, and APP.FEE were analyzed.
- Loadings:
 - MR1 (Factor 1): Strongly loads on LOAD, SEC.DEP, and DEV.CHG (all above 0.96).
 - MR2 (Factor 2): Strongly loads on APP.FEE (0.98).
 - MR3 (Factor 3): Minimal contribution to variance.
- Proportion Variance Explained:
 - MR1 explains 73% of variance.
 - MR2 explains an additional 26%.
 - MR3 contributes insignificantly.
- Chi-Square Test:
 - Fit indices indicate good reliability (e.g., Tucker Lewis Index = 1.004).
 - Residuals are minimal, suggesting the factors sufficiently explain the correlations.

Issues and Warnings:

- Missing values in the variable MF caused an error. Excluding MF allowed the analysis to proceed.
- Some loadings or results are highly skewed (e.g., standard deviation of residuals = 0), indicating potential data imbalance or redundancy.

2. Principal Component Analysis (PCA)

Purpose:

PCA reduces dimensionality while retaining most variance.

Setup:

- Variables used: LOAD, SEC.DEP, DEV.CHG, and APP.FEE.
- Standardization ensures comparability among variables with different scales.

Results:

- Principal Components:
 - PC1 explains 78.33% of the variance.
 - PC2 explains an additional 21.20%, capturing a combined 99.53%.
 - PC3 and PC4 are negligible, contributing <1%.
- Scree Plot: Sharp drop after PC2 confirms only 2 significant components.
- Biplot: Visualizes the contribution of variables to PC1 and PC2, showing clusters or relationships.

3. Hierarchical Clustering

Purpose:

Clustering groups similar observations based on a distance metric (e.g., Euclidean).

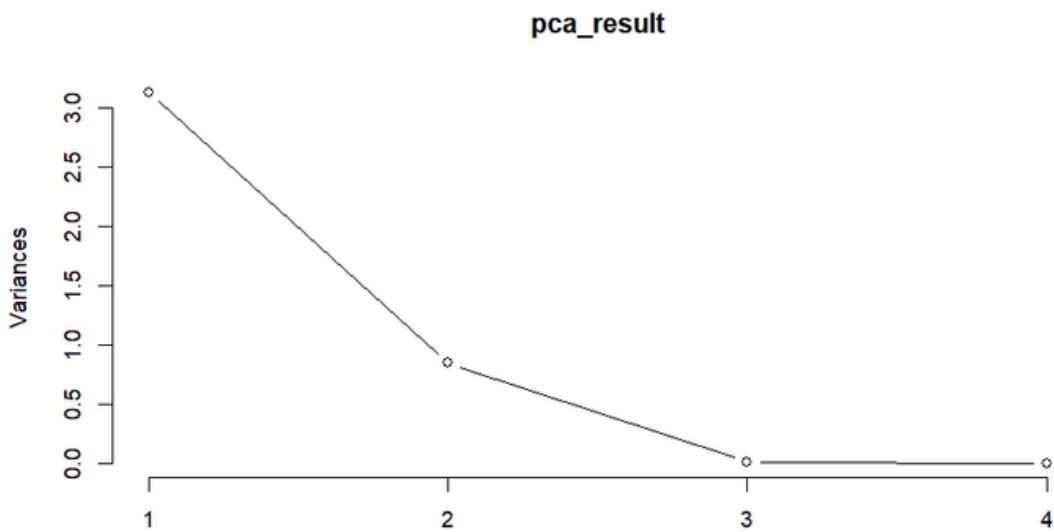
Setup:

- Distance matrix computed using Euclidean distance.
- Clustering method: Ward's method, which minimizes intra-cluster variance.

Dendrogram:

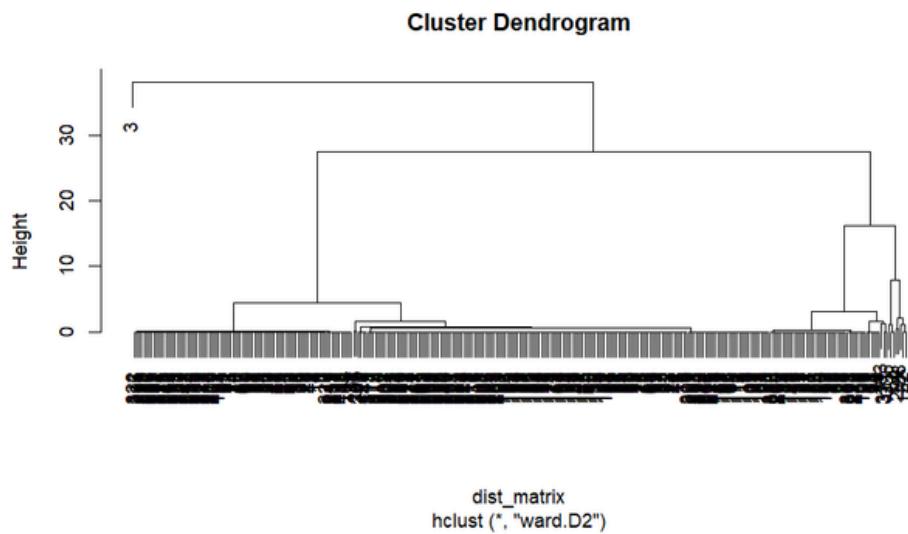
- Visualizes clusters hierarchically.
- Can be used to determine an optimal number of clusters by cutting the tree at an appropriate height.

Plot of PCA Result



The graph explains that the majority of variance is explained by first one or two components, as there is steep drop in variance after PC1 or PC2. This indicates that these components are most important for summarizing the data.

Cluster Dendrogram



C H A P T E R 5

Summary , Conclusions

&

Recommendations

Summary

The primary objective of the data collection was to comprehensively examine utility management practices in the Southern Power Distribution Company of Telangana Ltd. (TSSPDCL), specifically within the Mahbubnagar Operation Circle. The dataset focused on consumer behavior, electricity usage, and operational metrics, sourced from TSSPDCL's official records and databases. Information was gathered through data extraction from Management Information Systems (MIS), structured field surveys, and document reviews of scanned forms and agreements. This process ensured the inclusion of diverse variables, such as consumer identifiers, geographic details, operational attributes, meter data, financial records, and service details, providing a robust foundation for analysis.

Data preparation involved systematic cleaning, transformation, and integration to ensure quality and readiness for analysis. Missing data was handled through imputation or exclusion, and duplicates were removed using unique identifiers like SERVICE.NO. Variables were standardized, categorical fields encoded, and dates reformatted for consistency. Feature selection was employed to exclude irrelevant fields, and Principal Component Analysis (PCA) reduced redundancy and addressed multicollinearity. Validation steps ensured logical consistency and coherence between variables, enhancing the dataset's reliability.

The process also utilized advanced tools such as R and Microsoft Excel for data cleaning, transformation, and validation. Data integration merged information from multiple sources using unique keys, while feature engineering created new variables to enhance analytical insights. Consistency checks and statistical validations ensured accuracy, while tools like SQL databases supported efficient data querying and integration.

The final processed dataset was ready for statistical and predictive analyses, including regression, ANOVA, clustering, and PCA. Descriptive statistics summarized key variables, while inferential statistics and predictive models provided actionable insights. Ethical considerations, such as anonymization and regulatory compliance, safeguarded consumer privacy. Despite challenges like missing data and variability, the comprehensive methodology ensured a reliable and structured dataset for utility management analysis.

Data Exploration and Findings in Chapter -2

The dataset reveals several key patterns and insights. Variables such as MF, PHASE, and SEC.DEP show minimal or zero variance across a significant portion of the data, indicating homogeneity in these fields. In contrast, variables like LOAD and DEV.CHG exhibit large ranges, pointing to significant variability and potential outliers, with LOAD, for instance, reaching a maximum of 30,000 and displaying characteristics of a right-skewed distribution due to its high range, large IQR, and relatively low mean. The high frequency of specific values, such as modes in CATEGORY and LOAD, reflects recurring patterns that may signify typical operational scenarios or potential data duplication. Central tendencies, including mean, median, and mode, are consistent for many fields, indicating uniform or normal distributions. Relationships between variables also offer insights, such as the positive covariance between CATEGORY and LOAD, which suggests that higher categories (e.g., industrial or commercial consumers) are associated with increased electricity loads. Additionally, the dominance of certain values in frequency tables, such as Category 1 in CATEGORY and frequent values like 500 or 1000 in LOAD, highlights patterns in consumer behavior. Seasonal or temporal trends could also be explored using date fields like PR.DATE, which may reveal clustering of data entries on specific dates.

Data Exploration and Findings in Chapter -3

The point forecast represents the model's best estimate of future values for the "CATEGORY" variable, providing a precise predicted value. Confidence intervals complement this by accounting for uncertainty; the 80% confidence interval offers a narrower range with higher certainty about the predicted values, while the 95% confidence interval is broader, ensuring a more conservative estimate that increases the likelihood of capturing the true value. For example, the model predicts that "HPL - ELECTRIC" will have a significantly higher load (by 12,500 units) than "HIMACHAL ENERGY," with respective predicted loads of 13,250 and 750 units, demonstrating the relationship between "MAKE" and "LOAD."

In interpreting coefficients, the intercept of 10,800 indicates the baseline DEV.CHG for a reference AREA.CODE, while adjustments for other codes are reflected in their coefficients. For instance, an AREA.CODE of 1207 adds 12,600, resulting in a total DEV.CHG of 23,400, whereas an AREA.CODE of 302 decreases the charge by 9,900, leaving a predicted value of 900. These coefficients help estimate development charges for different areas relative to the baseline.

Statistical tests reveal significant findings across multiple dimensions. t-tests confirm meaningful differences in means between various groups, while proportion tests identify significant deviations in category proportions from hypothesized values. Variance tests highlight substantial differences in group variances, and correlation tests show strong positive relationships between SEC.DEP, DEV.CHG, and APP.FEE, suggesting interconnected financial patterns. This comprehensive analysis underscores the model's ability to uncover critical patterns and insights in the dataset.

Data Exploration and Findings in Chapter -4

The regression analysis reveals that while the coefficients for the intercept and the effect of MAKEHPL - ELECTRIC on LOAD are calculated, neither is statistically significant, indicating a weak or inconclusive relationship between LOAD and MAKE. This may result from high variability or an insufficient sample size. Similarly, analysis across models shows varying levels of significance. For instance, in the first model based on NATURE.OF.SPPLY, there is a significant difference in LOAD across supply categories (*p*-value = 0.00788). In contrast, the second model focusing on MAKE shows no significant differences in LOAD at the 5% significance level (*p*-value = 0.0911).

The One-Way ANOVA confirms significant differences in the means of LOAD and SEC.DEP across different NATURE.OF.SPPLY categories. The Tukey HSD test identifies which pairs of categories differ significantly; for example, Non-Domestic and Commercial supply types differ significantly from Agricultural and Domestic types. Further analysis reveals that the main effects of NATURE.OF.SPPLY and PHASE on LOAD are statistically significant when their respective *p*-values are below 0.05, indicating these factors strongly influence LOAD. Additionally, a significant interaction effect between NATURE.OF.SPPLY and PHASE suggests that the influence of supply type on LOAD varies by phase.

While MAKE might have a minor effect on its own, it is not strongly significant. However, the interaction between MAKE and NATURE.OF.SPPLY is significant, emphasizing the importance of considering their combined effect on LOAD. In multiple linear regression (MLR), Model 1 (with an intercept) performs better than Model 2 (without an intercept) based on lower AIC and residual deviance. However, the relationship between LOAD and CATEGORY remains weak in both models, reflected in small coefficients and minimal reductions in deviance.

Exploratory techniques like factor analysis and PCA identified two key latent factors (MR1 and MR2) that capture most of the variance, validating the dimensionality reduction approach. Clustering further supports segmentation tasks by uncovering group structures based on data similarity, which could be leveraged for targeted analysis or operational improvements. These findings collectively highlight the complex interplay of variables in explaining LOAD and guide further refinement of predictive models.

Conclusion

The analysis of utility management data from the Southern Power Distribution Company of Telangana Ltd. (TSSPDCL) revealed significant patterns in electricity consumption across various consumer categories such as Domestic, Agricultural, and Industrial. Key variables like LOAD, SEC.DEP, and DEV.CHG played a crucial role in explaining consumption behavior, with distinct financial and technical profiles observed across different groups. While statistical techniques like ANOVA, regression, and clustering provided valuable insights into energy usage and consumer behavior, challenges like missing data, multicollinearity, and limited predictive accuracy highlight the need for further refinement in data collection and modeling.

Recommenaditions

- Improve data quality by addressing missing values and ensuring consistency in key variables, such as MF and LOAD.
- Explore additional predictors (e.g., demographic factors, geographic information) to enhance model accuracy and predictive power.
- Investigate the impact of external factors, such as weather conditions and policy changes, on consumption patterns.
- Consider incorporating time-series data to better understand seasonal and trend-based consumption behaviors.
- Tailor pricing models and resource allocation strategies based on distinct consumer clusters identified in the analysis.
- Reduce multicollinearity by revisiting variable selection or using regularization techniques like ridge or lasso regression.
- Implement more advanced machine learning techniques for improved predictive modeling and accuracy in forecasting future consumption trends.

Annexure

R code used for analysis

```
NSR_M<-read.csv("C:\\Users\\pc\\OneDrive\\Documents\\project  
data\\NSR_MAHD.csv",header = TRUE)  
NSR_M  
  
attach(NSR_M)  
#descriptive statistics  
summary(NSR_M)  
str(NSR_M)  
names(NSR_M)  
dim(NSR_M)  
des_NSР_M<-NSR_M[,sapply(NSR_M,is.numeric)]  
des_NSР_M  
as.Date(DATE)  
factor(PHASE)  
#min  
minf<-function(x){  
  r<-min(x)  
  return(r)  
}  
for(i in 1:13){  
  res<-minf(des_NSР_M[i])  
  cat("Min of",names(des_NSР_M[i]),"is",res,"\\n")  
}  
  
#max  
maxf<-function(x){  
  r<-max(x)  
  return(r)  
}  
for(i in 1:13){  
  res<-maxf(des_NSР_M[i])  
  cat("Max of",names(des_NSР_M[i]),"is",res,"\\n")
```

```

#qunatile
quaf<-function(x){
  r<-quantile(x)
  return(r)
}
for(i in 1:13){
  res<-quaf(unlist(des_NS_R_M[i]))
  for(j in 1:4){
    cat(j,"th quantile of",names(des_NS_R_M[i]),"is",res[j],'\n')
  }
}
#idr
iqrdf<-function(x){
  r<-IQR(x)
  return(r)
}
for(i in 1:13){
  res<-iqrdf(unlist(des_NS_R_M[i]))
  cat("Interquartile Range of",names(des_NS_R_M[i]),"is",res,'n')
}
#qda
qdadf<-function(x){
  r<-IQR(x)/2
  return(r)
}
for(i in 1:13){
  res<-qdadf(unlist(des_NS_R_M[i]))
  cat("Quartile Deviation of",names(des_NS_R_M[i]),"is",res,'n')
}

```

```

#range
rangef<-function(x){
  r<-range(x)
  return(r)
}
for(i in 1:13){
  res<-rangef(s[i])
  cat("Range of",names(des_NS_R_M[i]),"is",res,'\n')
}
#sum
sumf<-function(x){
  r<-sum(x)
  return(r)
}
for(i in 1:13){
  res<-sumf(des_NS_R_M[i])
  cat("Sum of",names(des_NS_R_M[i]),"is",res,'\n')
}

```

```

#mean
meanf<-function(x){
  r<-mean(x)
  return(r)
}
for(i in 1:13){
  res<-meanf(unlist(des_NS_R_M[i]))
  cat("Mean of",names(des_NS_R_M[i]),"is",res,'\n')
}

```

```

mean deviation
madf<-function(x){
  r<-mad(x)
  return(r)
}
for(i in 1:13){
  res<-madf(unlist(des_NS_R_M[i]))
  cat("Mean deviation of",names(des_NS_R_M[i]),"is",res,'\n')
}

```

```

#sda
sdaf<-function(x){
  r<-sd(x)
  return(r)
}
for(i in 1:13){
  res<-sdaf(unlist(des_NS_R_M[i]))
  cat("Standard Deviation of",names(des_NS_R_M[i]),"is",res,"\n")
}

#deciles
decff<-function(x){
  r<-quantile(x,seq(0,1,by=0.1))
  return(r)
}
for(i in 1:13){
  res<-decff(unlist(des_NS_R_M[i]))
  for(j in 1:10){
    cat(j,"th Decile of",names(des_NS_R_M[i]),"is",res[j],"\n")
  }
}

#octiles
octf<-function(x){
  r<-quantile(x,seq(0,1,by=0.125))
  return(r)
}
for(i in 1:13){
  res<-octf(unlist(des_NS_R_M[i]))
  for(j in 1:8){
    cat(j,"th Octile of",names((des_NS_R_M[i])), "is",res[j],"\n")
  }
}

```

```

#median
medianf<-function(x){
  r<-median(x)
  return(r)
}
for(i in 1:13){
  res<-medianf(unlist(des_NSR_M[i]))
  cat("Median of",names(des_NSR_M[i]),"is",res,"\n")
}

#
#mode
NSR_MM<-NSR_M[c(-1,-3,-11,-19,-26)]
getmode<- function(v){
  x<-table(v)
  y<-max(x)
  a<-names(x)[which(x==y)]
  return(a)
}
for(i in 1:21){
  res<-getmode(unlist(NSR_MM[i]))
  cat("mode of",names((NSR_MM[i])), "is",res, '\n')
}

#
#cv
cvfun<-function(a,b){
  cv<-cov(a,b)
  return(cv)
}
for (i in 1:12) {
  for(j in 2:13){
    res<-cvfun(unlist(des_NSR_M[i]),unlist(des_NSR_M[j]))
    cat("covariance of",names(des_NSR_M[i]),"and",names(des_NSR_M[j]),"is",res, '\n')
  }
}

```

```

#cor
corfun<-function(a,b){
  cv<-cor(a,b)
  return(cv)
}
for (i in 1:12) {
  for(j in 2:13){
    res<-corfun(unlist(des_NSR_MM[i]),unlist(des_NSR_MM[j]))
    cat("correlation of",names(des_NSR_MM[i]),"and",names(des_NSR_MM[j]),"is",res,"\\n")
  }
}

#UNIVARIATE TABLE
NSR_UT<-NSR_MM[c(-1,-2,-3,-4,-16,-17)]
dim(NSR_UT)
utf<-function(x){
  r<-table(x)
  return(r)
}
for(i in 1:15){
  res<-utf(NSR_UT[i])
  print(res)
}

#BIVAR
NSR_BT<-NSR_MM[c(-1,-3,-7,-16,-17,-18,-19,-20,-21)]
dim(NSR_BT)
names(NSR_MM)
btf<-function(x,y){
  r<-table(x,y)
  return(r)
}
for(i in 1:11){
  for(j in 2:12)
    res<-btf(unlist(NSR_BT[i]),unlist(NSR_BT[j]))
    print(res)
}

```

```

#pie chart
pie(table(NSR_M$CATEGORY),main=" PIE CHART OF CATEGORY",col=rainbow(3))
pie(table(NSR_M$PHASE),main=" PIE CHART OF PHASE",col=rainbow(5))
pie(table(NSR_M$NATURE.OF.SPPPLY),main=" PIE CHART OF NATURE OF
SUPPLY",col=rainbow(3))
pie(table(NSR_M$MAKE),main=" PIE CHART OF MAKE",col=rainbow(2),labels =
c("HIMACHAL ENERGY","HPL - ELECTRIC"))

#hist
hist(NSR_M$APP.FEE)
hist(NSR_M$SEC.DEP)

#barplot
barplot(table(NSR_M$DEV.CHG))
barplot(table(NSR_M$DATE))
barplot(table(NSR_M$AREA.CODE))
barplot(table(NSR_M$MANDAL.NAME))

#Predictive statistics
install.packages("forecast")
library(forecast)
ts_data <- ts(NSR_M$CATEGORY, frequency = 12) # Assuming monthly data
fit <- auto.arima(ts_data)
forecast(fit, h = 12)

NSR_MH<-head(NSR_M)
NSR_MH

```

```

#load prediction
lm_model <- lm(LOAD ~ NATURE.OF.SPPLY + PHASE + AREA.CODE, data = NSR_MH)
lm_model
summary(lm_model)
test_data <- data.frame(
  NATURE.OF.SPPLY = c("DOMESTIC", "NON-DOMESTIC AND COMMERCIAL",
  "AGRICULTURAL", "AGRICULTURAL", "DOMESTIC",
  "NON-DOMESTIC AND COMMERCIAL"),
  PHASE = c(1, 3, 1, 2, 3, 1),
  AREA.CODE = c(101, 102, 103, 104, 101, 102))

)
predicted_load<-predict(lm_model, newdata = test_data)
predicted_load
test_data$Predicted_load <- predicted_load
print(test_data)

lm_model <- lm(LOAD ~ MAKE, data = NSR_MH)
lm_model
coef(summary(lm_model))
predict(lm_model, newdata = data.frame(MAKE="HPL - ELECTRIC"))
predict(lm_model, newdata = data.frame(MAKE="HIMACHAL ENERGY"))

lm_model <- lm(LOAD ~ 0+MAKE, data = NSR_MH)
lm_model

#dev.chg
lm_model <- lm(DEV.CHG ~ as.character(AREA.CODE) , data = NSR_MH)
lm_model
predict(lm_model, newdata = data.frame(AREA.CODE=302))

#PRESCRIPTIVE
t.test(NSR_M$LOAD, mu = 10)
t.test(NSR_M$SEC.DEP, mu = 12)
t.test(NSR_M$SEC.DEP, NSR_M$DEV.CHG)
t.test(NSR_M$SEC.DEP, NSR_M$APP.CHG)
t.test(LOAD ~ PHASE, data = NSR_M)
t.test(CATEGORY ~ PHASE, data = NSR_M)

```

```

#load prediction
lm_model <- lm(LOAD ~ NATURE.OF.SPPLY + PHASE + AREA.CODE, data = NSR_MH)
lm_model
summary(lm_model)
test_data <- data.frame(
  NATURE.OF.SPPLY = c("DOMESTIC", "NON-DOMESTIC AND COMMERCIAL",
  "AGRICULTURAL", "AGRICULTURAL", "DOMESTIC",
  "NON-DOMESTIC AND COMMERCIAL"),
  PHASE = c(1, 3, 1, 2, 3, 1),
  AREA.CODE = c(101, 102, 103, 104, 101, 102))

)
predicted_load<-predict(lm_model, newdata = test_data)
predicted_load
test_data$Predicted_load <- predicted_load
print(test_data)

lm_model <- lm(LOAD ~ MAKE, data = NSR_MH)
lm_model
coef(summary(lm_model))
predict(lm_model, newdata = data.frame(MAKE="HPL - ELECTRIC"))
predict(lm_model, newdata = data.frame(MAKE="HIMACHAL ENERGY"))

lm_model <- lm(LOAD ~ 0+MAKE, data = NSR_MH)
lm_model

#dev.chg
lm_model <- lm(DEV.CHG ~ as.character(AREA.CODE) , data = NSR_MH)
lm_model
predict(lm_model, newdata = data.frame(AREA.CODE=302))

#PRESCRIPTIVE
t.test(NSR_M$LOAD, mu = 10)
t.test(NSR_M$SEC.DEP, mu = 12)
t.test(NSR_M$SEC.DEP, NSR_M$DEV.CHG)
t.test(NSR_M$SEC.DEP, NSR_M$APP.CHG)
t.test(LOAD ~ PHASE, data = NSR_M)
t.test(CATEGORY ~ PHASE, data = NSR_M)

```

```

prop.test(x = sum(NSR_M$NATURE.OF.SPPLY == "Domestic"), n = nrow(NSR_M), p = 0.3)
prop.test(x = sum(NSR_M$NATURE.OF.SPPLY == "AGRICULTURAL"), n =
nrow(NSR_M), p = 0.2)
prop.test(x = sum(NSR_M$MAKE == "HPL - ELECTRIC"), n = nrow(NSR_M), p = 0.2)
prop.test(x = sum(NSR_M$MAKE == "HIMACHAL"), n = nrow(NSR_M), p = 0.7)

prop.test(c(sum(NSR_M$MANDAL.NAME == "MOHAMMADABAD "),
sum(NSR_M$MANDAL.NAME == " MOKARLABAD")),
c(nrow(NSR_M|NSR_M$MANDAL.NAME == "MOHAMMADABAD ", J),
nrow(NSR_M|NSR_M$MANDAL.NAME == "MOKARLABAD", J)))

var.test(NSR_M$LOAD,NSR_M$SEC.DEP, alternative = "two.sided", ratio = 1)
var.test(NSR_M$LOAD,NSR_M$DEV.CHG, alternative = "two.sided", ratio = 1)
var.test(NSR_M$DEV.CHG,NSR_M$SEC.DEP)
var.test(NSR_M$DEV.CHG,NSR_M$APP.FEE)
var.test(LOAD ~ PHASE, data = NSR_M)
var.test(SEC.DEP ~ PHASE, data = NSR_M)

cor.test(NSR_M$SEC.DEP, NSR_M$DEV.CHG)
cor.test(NSR_M$SEC.DEP, NSR_M$APP.FEE)
cor.test(NSR_M$APP.FEE, NSR_M$DEV.CHG)

install.packages("randtests")
library(randtests)
runs.test(NSR_M$LOAD)
runs.test(NSR_M$SEC.DEP)
runs.test(NSR_M$APP.FEE)
runs.test(NSR_M$DEV.CHG)

wilcox.test(LOAD ~ PHASE, data = NSR_M)
wilcox.test(SEC.DEP ~ PHASE, data = NSR_M)

library(BSDA)
SIGN.test(NSR_M$SEC.DEP, md = 15)
SIGN.test(NSR_M$APP.FEE, md = 1000)

```

```

#PRESCRIPTIVE STATISTICS
anova_model <- aov(LOAD ~ NATURE.OF.SPPLY, data = NSR_M)
#PRESCRIPTIVE STATISTICS
anova_model
summary(anova_model)
anova_model <- aov(LOAD+SEC.DEP ~ NATURE.OF.SPPLY, data = NSR_M)
#PRESCRIPTIVE STATISTICS
anova_model
summary(anova_model)
anova_model <- aov(LOAD ~ MAKE*NATURE.OF.SPPLY, data = NSR_M)
#PRESCRIPTIVE STATISTICS
anova_model
summary(anova_model)

```

```
TukeyHSD(anova_model)
```

```

#logistic
glmm<-glm(CATEGORY~LOAD,data=NSR_M)
glmm

```

```

glmm1<-glm(CATEGORY~0+LOAD,data=NSR_M)
glmm1

```

```

#mlr
lm_model <- lm(LOAD ~ MAKE+PHASE+CAPACITY, data = NSR_MH)
lm_model
predict(lm_model, newdata = data.frame(MAKE="HPL -
ELECTRIC",PHASE=1,CAPACITY="100/5A"))
predict(lm_model, newdata = data.frame(MAKE="HIMACHAL
ENERGY",PHASE=1,CAPACITY="10-40A"))

```

```

lm_model <- lm(LOAD ~ 0+MAKE+PHASE+CAPACITY, data = NSR_MH)
lm_model

```

```
install.packages("psych")
library(psych)

# Select numeric columns for factor analysis
numeric_data <- NSR_M[, sapply(NSR_M, is.numeric)] 

# Factor analysis with 2 factors (modify number as needed)
factor_analysis <- fa(numeric_data, nfactors = 3, rotate = "varimax")
print(factor_analysis)

names(NSR_M)
# Example: PCA on numerical data
numeric_data <- NSR_M[c(10,16,17,18)]

# Standardizing the data
scaled_data <- scale(numeric_data)

# PCA
pca_result <- prcomp(scaled_data)
summary(pca_result)

# Scree plot
plot(pca_result, type = "I")
biplot(pca_result)

dist_matrix <- dist(scaled_data, method = "euclidean")
hclust_model <- hclust(dist_matrix, method = "ward.D2")
plot(hclust_model)
```


82	14/08/2024	NR913244BALA VAR 7-84 ,DESIDESAIPALIDESAIPALAGRICULT	1	500 HP24036937740-HPL - ELECTRIC	15-30A	1	200	600	29.5	2413657026_14/08/2024 91303061166_14/08/2024 23/08/2024	309	1527002008	309000400
83	16/08/2024	NR913244K NARENT 3-38 ,DESIDESAIPALIDESAIPALAGRICULT	1	500 HP24036937740-HPL - ELECTRIC	15-30A	1	200	600	29.5	2415397011_16/08/2024 91303061166_16/08/2024 23/08/2024	309	1527002008	309000401
84	16/08/2024	NR913244K NARENT 3-38 ,DESIDESAIPALIDESAIPALAGRICULT	1	500 HP24036937740-HPL - ELECTRIC	15-30A	1	200	600	29.5	2425083213_12/09/2024 91303061166_16/08/2024 23/08/2024	309	1527002008	309000402
85	12/9/2024	NR913244K RAMES 7-86 ,DESIDESAIPALIDESAIPALAGRICULT	1	500 HP24036937740-HPL - ELECTRIC	15-30A	1	200	600	29.5	2425083213_12/09/2024 91303061166_16/08/2024 23/08/2024	309	1527002008	309000403
86	13/09/2024	NR913244Q-AZFEJA BE 8-37 ,DESIDESAIPALIDESAIPALNON-DOM	1	500 HP240369385194-HPL - ELECTRIC	15-30A	1	200	600	29.5	2435768608_13/09/2024 91303061166_13/09/2024	309	1527002008	309000404
87	26/10/2024	NR913244M KALASYN 383 DESAIPALIDESAIPALDOMESTIK	2	1000 HP24036937234-HPL - ELECTRIC	15-30A	1	800	1200	59	246602500_26/10/2024 913030615154_26/10/2024 28/10/2024	309	1527002008	309000405
88	3/11/2023	NR913235 G CHINA 7-5/11 ,GADDIRYAL GADIRYAL DOMESTIK	1	500 HE23082135645-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5617_2/11/2023 91304002191_3/11/2023 91117/2023	310	1527005015	310000565
89	3/11/2023	NR913233 VUAYA LA 2-5/1 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23082135645-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5613_2/11/2023 91304002190_3/11/2023 91117/2023	310	1527005015	310000566
90	19/12/2023	NR913233 GUPIA RAM 1-9/2 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23092138218-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5789_16/12/2023 913040003022_19/12/2023 26/12/2023	310	1527005015	310000567
91	19/12/2023	NR913233 NEFRATI 17-42 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23092138218-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5791_16/12/2023 913040003023_19/12/2023 26/12/2023	310	1527005015	310000568
92	19/12/2023	NR913233 ARUKALI 14-11 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23092138218-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5792_16/12/2023 913040003024_19/12/2023 26/12/2023	310	1527005015	310000569
93	19/12/2023	NR913233 KATIRE AN 3-30/1 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23092138218-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5790_16/12/2023 913040003025_19/12/2023 26/12/2023	310	1527005015	310000570
94	19/12/2023	NR913233 VARIA SRL 1-19 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23092138218-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5793_16/12/2023 913040003026_19/12/2023 26/12/2023	310	1527005015	310000571
95	19/12/2023	NR913233 PULLI AN 1-15 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23092138218-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5794_16/12/2023 913040003028_19/12/2023 26/12/2023	310	1527005015	310000572
96	9/1/2024	NR913245 VENKE 646 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23102150843-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5875_8/1/2024 913040003623_9/1/2024 16/01/2024	310	1527005015	310000573
97	12/1/2024	NR913245 ERARAKI 1-51 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23102150789-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5874_8/1/2024 913040003774_12/1/2024 17/01/2024	310	1527005015	310000574
98	12/1/2024	NR913245 MOGUL 2-31 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23102150789-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5872_16/12/2023 913040003025_19/12/2023 26/12/2023	310	1527005015	310000575
99	2/2/2024	NR913245 SALEM 3-26 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HE23122157875-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5793_16/12/2023 913040003026_19/12/2023 26/12/2023	310	1527005015	310000576
100	26/03/2024	NR913245 VUSLATI 4-14 ,GADDIRYAL GADIRYAL AGRICULT	1	1000 HE23122157875-HIMACHAL ENERGY	15-30A	1	200	600	29.5	5794_16/12/2023 913040003027_19/12/2023 26/12/2023	310	1527005015	310000577
101	16/04/2024	NR913245 SONAM 4-6 ,GADDIRYAL GADIRYAL AGRICULT	1	1000 HE24013567982-HPL - ELECTRIC	15-30A	1	200	600	29.5	305183_10/4/2024 9130400054022_16/04/2024 25/04/2024	310	1527005020	310000588
102	16/04/2024	NR913245 SONAM 4-6 ,GADDIRYAL GADIRYAL AGRICULT	1	1000 HE24013567982-HPL - ELECTRIC	15-30A	1	200	600	29.5	3200400536_16/04/2024 913040005402_16/04/2024 25/04/2024	310	1527005015	310000589
103	16/04/2024	NR913243 TANKARI 11-17 ,GADDIRYAL GADIRYAL AGRICULT	1	1000 HE24013567982-HPL - ELECTRIC	15-30A	1	200	600	29.5	3200400537_16/04/2024 913040005375_17/04/2024 25/04/2024	310	1527005015	310000590
104	21/05/2024	NR913245 Y GOPAL 7-46/1 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HP20193567917-HPL - ELECTRIC	15-30A	1	200	600	29.5	2334101684_21/05/2024 913030603383_21/05/2024 24/05/2024	310	1527005015	310000591
105	12/7/2024	NR913244 GUMDAU 7-35/3 ,GADDIRYAL GADIRYAL DOMESTIC	2	2000 HP24036936959-HPL - ELECTRIC	15-30A	1	200	600	29.5	305183_10/4/2024 913040005402_16/04/2024 25/04/2024	310	1527005020	310000592
110	11/7/2024	NR913244 MUCHIRL 6-29 ,GADDIRYAL GADIRYAL AGRICULT	2	1000 HP24036936959-HPL - ELECTRIC	15-30A	1	200	600	29.5	3200400536_16/04/2024 913040005402_16/04/2024 25/04/2024	310	1527005015	310000593
111	11/7/2024	NR913245 SANDRI 3-19/1 ,GADDIRYAL GADIRYAL AGRICULT	1	1000 HP24036936959-HPL - ELECTRIC	15-30A	1	200	600	29.5	3200400537_16/04/2024 913040005402_16/04/2024 25/04/2024	310	1527005015	310000594
112	17/08/2024	NR913244 KOTHAPALI 1-1/40 ,GADDIRYAL GADIRYAL AGRICULT	1	1000 HP2403693734-HPL - ELECTRIC	15-30A	1	200	600	29.5	2416258433_17/08/2024 91303061181_17/08/2024 20/08/2024	310	1527005015	310000595
131	22/10/2024	NR913244 ERAMOLL 7-49 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HP24036937989-HPL - ELECTRIC	15-30A	1	200	600	29.5	2463604265_22/10/2024 91303061191_22/10/2024 24/10/2024	310	1527005015	310000596
132	22/10/2024	NR913244 CHOURASI 3-31 ,GADDIRYAL GADIRYAL AGRICULT	1	1000 HP24036937989-HPL - ELECTRIC	15-30A	1	200	600	29.5	2463604266_22/10/2024 91303061192_22/10/2024 24/10/2024	310	1527005015	310000597
133	22/10/2024	NR913244 MARSARI 6-32 ,GADDIRYAL GADIRYAL AGRICULT	1	500 HP24036937989-HPL - ELECTRIC	15-30A	1	200	600	29.5	2463606880_22/10/2024 91303061193_22/10/2024 24/10/2024	310	1527005015	310000598
137	16/12/2023	NR913243 MUDAWA 4-5 ,JULAJULAPALLI MOHAMM DOMESTIC	1	1000 HE23092138195-HIMACHAL ENERGY	15-30A	1	200	600	29.5	3200400538_16/12/2023 913040002960_16/12/2023 23/12/2023	310	1527005015	310000599
138	2/2/2024	NR913245 GOLLA KE 1-164 ,JULAJULAPALLI JULAPALLI DOMESTIC	1	1000 HE23122158065-HIMACHAL ENERGY	15-30A	1	200	600	29.5	960749_29/01/2024 913040046468_2/2/2024_9/2/2024	312	1527011067	312000779
139	7/2/2024	NR913243 JUNIPALI 11-90 ,JULAJULAPALLI JULAPALLI DOMESTIC	1	1000 HE23122158012-HIMACHAL ENERGY	15-30A	1	200	600	29.5	875149_30/01/2024 913040044385_7/2/2024_11/2/2024	312	1527011067	312000780
140	7/2/2024	NR913243 GANJHUM 1-101 ,JULAJULAPALLI JULAPALLI DOMESTIC	1	1000 HE23122158012-HIMACHAL ENERGY	15-30A	1	200	600	29.5	2463606880_22/10/2024 91303061191_22/10/2024 24/10/2024	310	1527005015	310000591
141	24/02/2024	NR913245 GOLLA AN 1-78 ,JULAJULAPALLI JULAPALLI DOMESTIC	1	1000 HE23122158012-HIMACHAL ENERGY	15-30A	1	200	600	29.5	246102_12/07/2024 913040044386_7/7/2024_12/7/2024	312	1527011067	312000781
142	2/3/2024	NR913245 GOPAM 3-31 ,JULAJULAPALLI JULAPALLI NON-DOM	1	1000 HE2312215783-HIMACHAL ENERGY	15-30A	1	200	600	29.5	281735_2/3/2024 913040044387_7/7/2024_12/7/2024	312	1527011067	312000782
143	7/3/2024	NR913245 GANGARAS 148 ,JULAJULAPALLI JULAPALLI DOMESTIC	2	2000 HE23122157952-HIMACHAL ENERGY	15-30A	1	200	600	29.5	59_22/07/2024 7/3/2024 91304005063_7/7/2024_14/02/2024	312	1527011067	312000783
144	27/03/2024	NR913243 KOTHAMA 3-119 ,JULAJULAPALLI JULAPALLI AGRICULT	1	1000 HE24013567309-HPL - ELECTRIC	15-30A	1	200	600	29.5	281737_7/3/2024 91304005063_7/7/2024_27/03/2024	312	1527011067	312000784
148	20/05/2024	NR913244 GOLLA VE 1-78 ,JULAJULAPALLI JULAPALLI DOMESTIC	1	1000 HE24013567569-HPL - ELECTRIC	15-30A	1	200	600	29.5	2333173075_20/05/2024 913030605304_20/05/2024_24/05/2024	312	1527011067	312000785
149	20/05/2024	NR913244 KASHIFUR 2-81 ,JULAJULAPALLI JULAPALLI DOMESTIC	1	1000 HE24013567569-HPL - ELECTRIC	15-30A	1	200	600	29.5	2333173075_20/05/2024 913030605304_20/05/2024_24/05/2024	312	1527011067	312000786
150	19/06/2024	NR913244 SMTI SWA 1-61 ,JULAJULAPALLI JULAPALLI DOMESTIC	1	1000 HP24036937021-HPL - ELECTRIC	15-30A	1	200	600	29.5	235741291_19/06/2024 913030606093_19/06/2024_24/06/2024	312	1527011067	312000787
151	9/7/2024	NR913244 G RAM 3-2 ,JULAJULAPALLI JULAPALLI DOMESTIC	1	1000 HP24036937021-HPL - ELECTRIC	15-30A	1	200	600	29.5	2376206697_9/7/2024 913030606093_9/7/2024_18/07/2024	312	1527011067	312000788
152	27/07/2024	NR913244 SMTI R 5/5 ,JULAJULAPALLI JULAPALLI NON-DOM	1	1000 HP2403693712-HPL - ELECTRIC	15-30A	1	200	600	29.5	788377_26/07/2024 91303061049_27/07/2024_31/07/2024	312	1527011067	312000789
153	3/8/2024	NR913244 PIMJAR 1-36 ,JULAJULAPALLI JULAPALLI AGRICULT	2	1000 HP2403693712-HPL - ELECTRIC	15-30A	1	200	600	29.5	2402759201_3/8/2024 91303061049_3/8/2024_7/8/2024	312	1527011067	312000790
172	26/10/2024	NR913244 GANGARA 4-74 ,KOLIKOBA TH-JULAPALL NON-DOM	1	1000 HP24036938849-HPL - ELECTRIC	15-30A	1	200	600	29.5	610494_22/10/2024 91303061508_25/10/2024_29/10/2024	312	1527011067	312000791

9				X		✓		fx		NR913244094296																	
A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z	AA	
174	1/11/2023	NR913223 POTHULA 8-225 ,KAN KANCHAN KANCHAN DOMESTIC	2	1000	HE220821356121HIMACHAL ENERGY	1	5-30A	1	800	1200	59	109568	8/9/2023	91304002104	1/11/2023	9/11/2023	313	1527213004	313000540								
175	2/1/2024	NR913223 KURVA 9a 8-263 ,KAN KANCHAN KANCHAN DOMESTIC	1	500	HE23102151198HIMACHAL ENERGY	1	5-30A	1	200	600	29.5	96053	2/1/2023	91304003325	2/1/2024	6/1/2024	313	1527012015	313000541								
176	2/1/2024	NR913243 ANEM AN 8-74 ,KAN KANCHAN KANCHAN DOMESTIC	1	500	HE23102151210HIMACHAL ENERGY	1	5-30A	1	200	600	29.5	96052	2/1/2023	91304003327	2/1/2024	6/1/2024	313	1527012012	313000542								
177	2/2/2024	NR913243 A VENKAT 8-19 ,KAN KANCHAN KANCHAN DOMESTIC	1	500	HE23122158069HIMACHAL ENERGY	1	5-30A	1	200	600	29.5	96050	2/1/2023	91304004269	2/2/2024	8/2/2024	313	1527012015	313000543								
178	7/3/2024	NR913243 GOLKA GH 8-39 ,KAN KANCHAN KANCHAN NON-DOM	1	500	HE23122157871HIMACHAL ENERGY	1	5-30A	1	200	600	29.5	96115	7/3/2024	91304005659	25/05/2024	29/05/2024	313	1527012015	313000544								
179	25/05/2024	NR913243 GOILA GH 8-39 ,KAN KANCHAN KANCHAN DOMESTIC	2	2000	HP23093550841HPL - ELECTRIC	1	5-30A	1	1600	2400	59	2337187629	25/05/2024	91303066659	25/05/2024	29/05/2024	313	1527012002	313000545								
180	7/6/2024	NR913244 BOLLAVATI 8-91/2 ,KAN KANCHAN KANCHAN NON-DOM	1	1000	HP23093550397HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	96156	7/6/2024	91303066833	22/06/2024	24/06/2024	313	1527024012	313000546								
181	22/06/2024	NR913244 GANGULI SYNO 476 ,KAN KANCHAN KANCHAN DOMESTIC	2	1000	HP23093550467HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	96168	18/06/2024	9130306833	22/06/2024	24/06/2024	313	1527024012	313000547								
182	22/06/2024	NR913244 B NARAYA 8-91/2 ,KAN KANCHAN KANCHAN DOMESTIC	1	1000	HP23093550467HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	109571	8/9/2023	9130400982	13/09/2023	22/09/2023	317	1527007006	317003244								
183	13/09/2023	NR913244 APPAGALLI 7-181 ,KAN KANCHAN KANCHAN DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	109564	8/9/2023	91304009820	13/09/2023	22/09/2023	317	1527007006	317003249								
184	13/09/2023	NR913244 PADMA 3-39/2 ,M MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	109563	8/9/2023	91304009975	13/09/2023	22/09/2023	317	1527007011	317003250								
185	13/09/2023	NR913244 PADMA 3-90/2 ,M MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	109562	8/9/2023	91304009975	13/09/2023	22/09/2023	317	1527007006	317003252								
187	13/09/2023	NR913244 PADMA 3-90/2 ,M MOHAM MOHAM NON-DOM	2	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	800	1416	29.5	109574	12/9/2023	91304009983	13/09/2023	22/09/2023	317	1527007006	317003252								
188	13/09/2023	NR913244 KATRAVIA 5-170/5 ,M MOHAM MOHAM DOMESTIC	2	500	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	708	29.5	5182	13/09/2023	91304009983	13/09/2023	22/09/2023	317	1527007011	317003253								
189	13/09/2023	NR913244 APPAGALLI 7-181 ,M MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5180	13/09/2023	91304009982	13/09/2023	22/09/2023	317	1527007011	317003254								
190	13/09/2023	NR913244 VADLA YA 2-245 ,MC MOHAM MOHAM NON-DOM	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	800	1416	29.5	5197	14/09/2023	91304009982	14/09/2023	22/09/2023	317	1527007025	317003255								
191	15/09/2023	NR913244 P C HENNIBA 3-64/1 ,KAN KANCHAN KANCHAN DOMESTIC	2	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	109563	8/9/2023	91304009975	13/09/2023	22/09/2023	317	1527007011	317003251								
192	16/09/2023	NR913244 MEDIPALLI 1-24/2 ,M MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	800	1416	29.5	109572	12/9/2023	91304009983	13/09/2023	22/09/2023	317	1527007025	317003252								
193	16/09/2023	NR913244 SATHYA 3-231 ,MC MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5193	14/09/2023	91304009983	13/09/2023	22/09/2023	317	1527007025	317003252								
194	16/09/2023	NR913244 SATHYA 3-231 ,MC MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5191	14/09/2023	91304009983	13/09/2023	22/09/2023	317	1527007011	317003254								
195	16/09/2023	NR913244 SRILATHA 1-158/1 ,KAN KANCHAN KANCHAN DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5192	14/09/2023	91304009983	13/09/2023	22/09/2023	317	1527007025	317003250								
196	16/09/2023	NR913244 G SUREH 3-244/2 ,M MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5190	14/09/2023	91304009983	13/09/2023	22/09/2023	317	1527007011	317003256								
197	25/09/2023	NR913244 FOULIA BL 4-6/6 ,MO MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5195	14/09/2023	91304009983	13/09/2023	22/09/2023	317	1527007025	317003257								
198	19/09/2023	NR913244 CHINTHAK 2-220 ,MC MOHAM MOHAM DOMESTIC	1	1000	HP23053339783HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	109585	7/10/2023	91304001666	16/09/2023	22/09/2023	317	1527007025	317003258								
199	10/10/2023	NR913244 VENKAT R 2-276 ,MC MOHAM MOHAM DOMESTIC	1	1000	HP23053339598HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	109586	7/10/2023	91304001666	16/09/2023	22/09/2023	317	1527007011	317003254								
200	10/10/2023	NR913244 SRI LATHA 2-158/1 ,KAN KANCHAN KANCHAN DOMESTIC	1	1000	HP23053339922HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5444	7/10/2023	91304001695	16/09/2023	22/09/2023	317	1527007025	317003256								
201	11/10/2023	NR913244 BESTHA LI 8-194 ,KAN KANCHAN KANCHAN DOMESTIC	1	1000	HP23053339922HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5313	25/09/2023	91304001298	25/09/2023	317	1527007027	317003261									
202	11/10/2023	NR913244 BESHTHA LI 8-194 ,KAN KANCHAN KANCHAN DOMESTIC	1	1000	HP23053339922HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5249	19/09/2023	91304001112	19/09/2023	29/09/2023	317	1527007011	317003262								
203	11/10/2023	NR913244 A JEMINI 11-168 ,MC MOHAM MOHAM NON-DOM	1	1000	HP23053339922HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	154291	30/10/2023	91304002255	4/11/2023	29/10/2023	317	1527007025	317003263								
204	4/11/2023	NR913244 SARITHA 2-115/2 ,MC MOHAM MOHAM NON-DOM	2	1000	HP23082135473HIMACHAL ENERGY	1	5-30A	1	800	1200	59	154290	30/10/2023	91304002255	4/11/2023	29/10/2023	317	1527016011	317003277								
205	4/11/2023	NR913244 SARITHA 2-115/2 ,MC MOHAM MOHAM DOMESTIC	1	1000	HP23082135473HIMACHAL ENERGY	1	5-30A	1	800	1200	59	154291	30/10/2023	91304002255	4/11/2023	29/10/2023	317	1527016011	317003278								
206	4/11/2023	NR913244 BOKKA ER 2-254 ,MC MOHAM MOHAM NON-DOM	1	1000	HP23082135473HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	281667	17/10/2023	91304002257	4/11/2023	29/10/2023	317	1527007025	317003274								
207	4/11/2023	NR913244 A MAHAL 7-110/3 ,M MOHAM MOHAM NON-DOM	1	1000	HP23082135473HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	281668	17/10/2023	91304002259	4/11/2023	29/10/2023	317	1527007025	317003275								
208	4/11/2023	NR913244 BOKKA ER 7-110/3 ,M MOHAM MOHAM NON-DOM	2	1000	HP23082135473HIMACHAL ENERGY	1	5-30A	1	800	1200	29.5	154294	30/10/2023	91304002259	4/11/2023	29/10/2023	317	1527016011	317003276								
209	14/11/2023	NR913244 KOMIRE 1-47 ,M MOHAM MOHAM NON-DOM	1	1000	HP23082135473HIMACHAL ENERGY	1	5-30A	1	800	1200	29.5	154295	30/10/2023	91304002259	4/11/2023	29/10/2023	317	1527016011	317003277								
210	22/11/2023	NR913244 BOKKA ER 2-254 ,MC MOHAM MOHAM NON-DOM	2	1000	HP23082135473HIMACHAL ENERGY	1	5-30A	1	800	1200	29.5	154296	30/10/2023	91304002259	4/11/2023	29/10/2023	317	1527016011	317003278								
211	7/2/2024	NR913244 BEIJAN BI 7-104 ,MC MOHAM MOHAM NON-DOM	2	1000	HP23122158214HIMACHAL ENERGY	1	5-30A	1	800	1200	29.5	5911															

225	7/3/2024	NIR913243GUTTI CHI-3-203	MC MOHAMM. MOHAMM. NON-DOM	1	1000 HE23122157870/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 14/02/2024 913040005064	/13/2024 16/03/2024	31V	12/20/16014 31/03/225
226	7/3/2024	NIR913243G KESHAV 2-57	MOI MOHAMM. MOHAMM. DOMESTIC	2	1000 HE23122158051/HIMACHAL ENERGY	1 5-30A	1	800	1200	29.5	96110 14/02/2024 913040005065	/7/2024 14/03/2024	31V	15/20/16006 31/03/229
227	18/03/2024	NIR913243J BHUMLI-2-114/2	MOI MOHAMM. MOHAMM. NON-DOM	1	1000 HE23122158051/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 14/02/2024 913040005194	18/03/2024 23/03/2024	31V	15/20/16636 31/03/229
228	18/03/2024	NIR913243J BHUMLI-2-114/2	MOI MOHAMM. MOHAMM. DOMESTIC	2	1000 HE23122158051/HIMACHAL ENERGY	1 5-30A	1	800	1200	29.5	96110 14/02/2024 913040005195	18/03/2024 23/03/2024	31V	15/20/16636 31/03/229
229	18/03/2024	NIR913243J PASUPULI-2-164	MC MOHAMM. MOHAMM. NON-DOM	1	1000 HE23122158051/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 14/02/2024 913040005193	18/03/2024 23/03/2024	31V	15/20/16641 31/03/300
230	2/4/2024	NIR913243S SEETHA 5-166	MC MOHAMM. MOHAMM. DOMESTIC	2	1000 HE24013567900/HPL - ELECTRIC	1 5-30A	1	800	1200	29.5	96110 14/02/2024 913040005351	2/4/2024 16/04/2024	31V	15/20/16641 31/03/302
231	8/4/2024	NIR913243 KAVALIN N-12-124	MC MOHAMM. MOHAMM. DOMESTIC	1	1000 HE24013567903/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 14/02/2024 913040005408	8/4/2024 15/04/2024	31V	15/20/16606 31/03/303
232	16/04/2024	NIR913243KOUDEI BE-7-137	MC MOHAMM. MOHAMM. AGRICULT	1	1000 HE24013567228/HPL - ELECTRIC	1 5-30A	1	600	1200	29.5	96110 16/04/2024 913040005408	16/04/2024 25/04/2024	31V	15/20/16615 31/03/304
233	9/5/2024	NIR913244B LAXMI 8-194	KAI KANCHAN MOHAMM. NON-DOM	1	1000 HE24013567453/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 9/5/2024 913040005566	9/5/2024 17/05/2024	31V	15/27/16662 31/03/308
234	9/5/2024	NIR913244B LAXMI 8-194	KAI KANCHAN MOHAMM. DOMESTIC	2	1000 HE24013567451/HPL - ELECTRIC	1 5-30A	1	800	1200	29.5	96110 9/5/2024 913040005566	9/5/2024 17/05/2024	31V	15/27/16662 31/03/309
235	9/5/2024	NIR913244S SERI RAM 1-26	MOI MOHAMM. MOHAMM. DOMESTIC	1	1000 HE24013567452/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 9/5/2024 913040005566	9/5/2024 17/05/2024	31V	15/27/16641 31/03/310
236	10/5/2024	NIR913244KOUDEI BE-7-137	MC MOHAMM. MOHAMM. AGRICULT	1	1000 HE24013567607/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 10/5/2024 913040005576	17/05/2024 25/04/2024	31V	15/27/16621 31/03/311
237	10/5/2024	NIR913244KOUDEI BE-7-137	MC MOHAMM. MOHAMM. DOMESTIC	1	1000 HE24013567608/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 10/5/2024 913040005576	17/05/2024 25/04/2024	31V	15/27/16621 31/03/311
238	10/5/2024	NIR913244KOUDEI BE-7-137	MC MOHAMM. MOHAMM. AGRICULT	1	1000 HE24013567609/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 10/5/2024 913040005576	17/05/2024 25/04/2024	31V	15/27/16621 31/03/311
239	21/06/2024	NIR913244M HARISH-1-23	MOI MOHAMM. MOHAMM. DOMESTIC	1	1000 HE24013567610/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 21/06/2024 91303006315	21/06/2024 26/06/2024	31V	15/27/16606 31/03/321
240	25/06/2024	NIR913244KOUDEI BE-7-137	MC MOHAMM. MOHAMM. NON-DOM	1	1000 HE2403693665/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 25/06/2024 913030060853	25/06/2024 28/06/2024	31V	15/27/16662 31/03/323
241	4/7/2024	NIR913244PITLA CHES-9971	M MOHAMM. MOHAMM. DOMESTIC	2	1000 HE24036937440/HPL - ELECTRIC	1 5-30A	1	800	1200	29.5	96110 4/7/2024 913030060905	4/7/2024 12/07/2024	31V	15/27/16606 31/03/324
242	16/07/2024	NIR913244P PADMAM 3-100	MC MOHAMM. MOHAMM. AGRICULT	1	1000 HE24036937515/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 16/07/2024 913030060926	16/07/2024 24/07/2024	31V	15/27/16607 31/03/325
243	16/07/2024	NIR913244P VINOD KUDHAR 6-53/1	MC MOHAMM. MOHAMM. DOMESTIC	1	1000 HE24036937612/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 16/07/2024 913030061164	14/08/2024 23/08/2024	31V	15/27/16601 31/03/311
244	14/08/2024	NIR913244POKKA V-3-60	MOI MOHAMM. MOHAMM. DOMESTIC	1	1000 HE24036938336/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 14/08/2024 913030061164	14/08/2024 23/08/2024	31V	15/27/16601 31/03/311
245	24/08/2024	NIR913244KONREDI 1-47	MOI MOHAMM. MOHAMM. AGRICULT	1	1000 HE24036937758/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 24/08/2024 913030061222	24/08/2024 29/08/2024	31V	15/27/16662 31/03/339
246	4/9/2024	NIR913244B BHARAT-83	MOI MOHAMM. MOHAMM. DOMESTIC	2	1000 HE24036937955/HPL - ELECTRIC	1 5-30A	1	800	1200	29.5	96110 4/9/2024 913030061273	4/9/2024 12/07/2024	31V	15/27/16606 31/03/345
247	4/9/2024	NIR913244B RADHIK 6-53/1	M MOHAMM. MOHAMM. NON-DOM	1	1000 HE24036937955/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 4/9/2024 913030061276	4/9/2024 12/07/2024	31V	15/27/16607 31/03/346
248	4/9/2024	NIR913244B VINOD KUDHAR 6-53/1	M MOHAMM. MOHAMM. DOMESTIC	2	1000 HE24036937955/HPL - ELECTRIC	1 5-30A	1	800	1200	29.5	96110 4/9/2024 913030061277	4/9/2024 12/07/2024	31V	15/27/16602 31/03/347
249	4/9/2024	NIR913244B VINOD KUDHAR 6-53/1	M MOHAMM. MOHAMM. AGRICULT	1	1000 HE24036937955/HPL - ELECTRIC	1 5-30A	1	800	1200	29.5	96110 4/9/2024 913030061278	4/9/2024 12/07/2024	31V	15/27/16604 31/03/338
250	16/07/2024	NIR913244B VINOD KUDHAR 6-53/1	M MOHAMM. MOHAMM. DOMESTIC	1	1000 HE24036937955/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 16/07/2024 913030061279	16/07/2024 24/07/2024	31V	15/27/16601 31/03/332
251	16/07/2024	NIR913244B VINOD KUDHAR 6-53/1	M MOHAMM. MOHAMM. AGRICULT	1	1000 HE24036937955/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 16/07/2024 913030061280	16/07/2024 24/07/2024	31V	15/27/16601 31/03/332
252	17/07/2024	NIR913244B VINOD KUDHAR 6-53/1	M MOHAMM. MOHAMM. DOMESTIC	1	1000 HE24036937955/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 17/07/2024 913030061281	17/07/2024 24/07/2024	31V	15/27/16601 31/03/332
253	18/07/2024	NIR913244B VINOD KUDHAR 6-53/1	M MOHAMM. MOHAMM. NON-DOM	1	1000 HE24036937955/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 18/07/2024 913030061282	18/07/2024 24/07/2024	31V	15/27/16601 31/03/332
254	18/11/2023	NIR913243POTHNEF 1-3-4/5/1	MOIKARIA MOKARIA AGRICULT	2	4000 HE23091213905/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 18/11/2023 91304002559	18/11/2023 26/11/2023	31V	15/27/16650 31/03/339
255	18/11/2023	NIR913243KAVALIN 1-2/74	M MUKARIA MOKARIA DOMESTIC	1	1000 HE23091213905/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 18/11/2023 91304002561	18/11/2023 26/11/2023	31V	15/27/16650 31/03/345
256	18/11/2023	NIR913243KAVALIN 1-2/74	M MUKARIA MOKARIA AGRICULT	2	4000 HE23091213905/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 18/11/2023 91304002561	18/11/2023 26/11/2023	31V	15/27/16650 31/03/345
257	2/4/2024	NIR913243G KAVELINT 1-5/42	M MOKARIA MOKARIA DOMESTIC	1	1000 HE23122158073/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 2/4/2024 91304002573	2/4/2024 14/02/2024	31V	15/27/16650 31/03/346
258	2/4/2024	NIR913243G KAVELINT 1-5/42	M MOKARIA MOKARIA AGRICULT	1	1000 HE23122158073/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 2/4/2024 91304002573	2/4/2024 14/02/2024	31V	15/27/16650 31/03/346
259	2/4/2024	NIR913243G KAVELINT 1-5/42	M MOKARIA MOKARIA DOMESTIC	2	4000 HE23091213905/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 2/4/2024 91304002574	2/4/2024 14/02/2024	31V	15/27/16650 31/03/347
260	2/4/2024	NIR913243G KAVELINT 1-5/42	M MOKARIA MOKARIA AGRICULT	1	1000 HE23091213905/HIMACHAL ENERGY	1 5-30A	1	200	1200	29.5	96110 2/4/2024 91304002574	2/4/2024 14/02/2024	31V	15/27/16650 31/03/347
261	14/08/2024	NIR913243KATIKA Y-3-1/1	MUMUKALA MOKARIA NON-NON	1	1000 HE23082133452/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 14/08/2024 913030061164	14/08/2024 23/08/2024	31V	15/27/16650 31/03/348
262	14/08/2024	NIR913243KATIKA Y-3-1/1	MUMUKALA MOKARIA AGRICULT	2	4000 HE23082133452/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 14/08/2024 913030061164	14/08/2024 23/08/2024	31V	15/27/16650 31/03/348
263	4/9/2024	NIR913243KATIKA Y-3-1/1	MUMUKALA MOKARIA DOMESTIC	1	1000 HE23082133452/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 4/9/2024 913030061164	4/9/2024 12/07/2024	31V	15/27/16650 31/03/349
264	4/9/2024	NIR913243KATIKA Y-3-1/1	MUMUKALA MOKARIA AGRICULT	2	4000 HE23082133452/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 4/9/2024 913030061164	4/9/2024 12/07/2024	31V	15/27/16650 31/03/349
265	18/11/2023	NIR913223 KURVA 3-3-42	NAN NANCCHAR NANCCHAR INDUSTRI	1	1000 HE230535339374/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 18/11/2023 91303061052	27/07/2024 27/07/2024	31V	15/27/16650 31/03/625
266	18/11/2023	NIR913223 KURVA 3-3-42	NAN NANCCHAR NANCCHAR DOMESTIC	2	1000 HE230535339374/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 18/11/2023 91303061052	27/07/2024 27/07/2024	31V	15/27/16650 31/03/625
267	23/09/2023	NIR913223 KURVA 3-3-42	NAN NANCCHAR NANCCHAR DOMESTIC	1	1000 HE230535339374/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 23/09/2023 91304001261	23/09/2023 30/09/2023	31V	15/27/16650 31/03/625
268	23/09/2023	NIR913223 KURVA 3-3-42	NAN NANCCHAR NANCCHAR DOMESTIC	2	1000 HE230535339374/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 23/09/2023 91304001261	23/09/2023 30/09/2023	31V	15/27/16650 31/03/625
269	23/09/2023	NIR913223 KURVA 3-3-42	NAN NANCCHAR NANCCHAR DOMESTIC	1	1000 HE230535339374/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 23/09/2023 91304001261	23/09/2023 30/09/2023	31V	15/27/16650 31/03/625
270	23/09/2023	NIR913223 KURVA 3-3-42	NAN NANCCHAR NANCCHAR DOMESTIC	2	1000 HE230535339374/HPL - ELECTRIC	1 5-30A	1	200	1200	29.5	96110 23/09/2023 91304001261	23/09/2023		

423	17/05/2024 NR913244G KONDAI 3-1/98, LITUNGAI PAULINGAI PAIDOMESTIK	1	1000 HP2401356743A HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2379750693	12/7/2024	913036069177	12/7/2024 18/07/2024	331	1577018051	31/10/2020
424	12/7/2024 NR913244A BUJII BA 4-2/61 LITUNGAI PAULINGAI PAAGRICULT	1	500 HP2403693735H HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5245	19/09/2023	913040612019	11/10/2023 30/09/2023	332	1577023015	31/10/2015
427	22/09/2023 NR913235S KISHAN 2-51 MAN MANGAM MANGAM DOMESTIK	1	1000 HP2305353387H HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	782	10/03/2024	913040672109	11/10/2023 11/10/2023	332	1577023015	31/10/2015
428	1/1/2023 NR913235S KISHAN 2-51 MAN MANGAM MANGAM DOMESTIK	1	1000 HE2310215093H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	581	8/1/2024	91304067625	9/1/2024 16/01/2024	332	1577023015	31/10/2015
429	9/1/2024 NR913242S SABAATI 2-16 MAN MANGAM MANGAM DOMESTIK	1	500 HE2312215152H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	674273	6/2/2024	91304064419	11/10/2024 12/10/2024	332	1577023015	31/10/2015
430	7/2/2024 NR913243S NARRY 2-20 MAN MANGAM MANGAM DOMESTIK	1	500 HE2312215152H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	674274	6/2/2024	91304064418	11/10/2024 12/10/2024	332	1577023015	31/10/2015
431	7/2/2024 NR913243K LINQYA 1-16 MAN MANGAM MANGAM DOMESTIK	1	500 HE2312215152H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	674275	6/2/2024	91304067623	11/10/2024 12/10/2024	332	1577023015	31/10/2015
432	17/02/2024 NR913243K LINQYA 2-78 1.M MANGAM MANGAM DOMESTIK	1	1000 HE2312215095H HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	674276	16/02/2024	91304067623	11/10/2024 12/10/2024	332	1577023015	31/10/2015
433	1/1/2023 NR913243K PASUJULI 8-20 DHA DHARMA DHARMA DOMESTIK	1	500 HE2308213559H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	519375	26/10/2023	91304062177	11/10/2023 11/10/2023	332	1577023005	31/10/2019
434	1/1/2023 NR913242S NALLAMA 7-30 DHA DHARMA DHARMA DOMESTIK	1	500 HE2312215152H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	96114	7/3/2024	91304065066	7/3/2024 15/03/2024	332	1577030008	31/09/2019
442	7/3/2024 NR913242S NALLAMA 7-30 DHA DHARMA DHARMA DOMESTIK	1	500 HE2305539873H HPL - ELECTRIC	1	15-30A	1	200	708	29.5	5288	22/09/2023	913040601221	22/09/2023 09/09/2023	332	1577024025	31/09/2009
447	22/09/2023 NR913242S CHENGHU 3-53/1 HPL KOULI VENKATRI DOMESTIK	1	1000 HP2305539873H HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5288	22/09/2023	913040601223	22/09/2023 09/09/2023	332	1577024025	31/09/2009
448	22/09/2023 NR913242S CHENGHU 3-53/1 HPL KOULI VENKATRI DOMESTIK	1	500 HP2305539873H HPL - ELECTRIC	1	15-30A	1	200	708	29.5	519374	26/10/2023	91304062176	11/10/2023 9/11/2023	332	1577023005	31/10/2019
449	11/10/2023 NR913242S CHENGHU 3-53/1 HPL KOULI VENKATRI DOMESTIK	1	500 HP2305539873H HPL - ELECTRIC	1	15-30A	1	200	708	29.5	5434	*****	91304061690	*****	332	1577023005	31/10/2019
451	11/10/2023 NR913242S CHENGHU 3-53/1 HPL KOULI VENKATRI DOMESTIK	1	1000 HE2312215152H HIMACHAL ENERGY	1	15-30A	1	200	29.5	5433	*****	91304061691	*****	332	1577023005	31/10/2019	
452	10/2/2024 NR913242S CHENGHU 3-53/1 HPL KOULI VENKATRI DOMESTIK	1	1000 HE2312215152H HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	22064833	10/2/2024	91304064489	10/2/2024 12/2/2024	332	1577023005	31/10/2019
453	22/09/2023 NR913242S VADDE SA 4-26, VEN VENKATRI VENKATRI DOMESTIK	1	500 HP2305539873H HPL - ELECTRIC	1	15-30A	1	200	708	29.5	5294	22/09/2023	9130406227	22/09/2023 09/09/2023	332	1577024025	31/09/2009
454	19/09/2023 NR913242S KOURID GC 2-51, VEN VENKATRI VENKATRI DOMESTIK	1	1000 HP2305539873H HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	109572	12/9/2023	91304061109	19/09/2023 29/09/2023	342	1577034025	31/09/2005
455	6/10/2023 NR913242S PAKARAK 3-23/1, VEN VENKATRI VENKATRI DOMESTIK	2	1000 HP2305539873H HPL - ELECTRIC	1	15-30A	1	200	800	29.5	7705	7/10/2023	91304061569	6/10/2023 12/10/2023	342	1577024025	31/09/2005
456	17/10/2023 NR913242S PAKARAK 3-23/1, VEN VENKATRI VENKATRI DOMESTIK	1	1000 HE2308213556H HIMACHAL ENERGY	1	15-30A	1	200	1416	29.5	281664	*****	913040601813	*****	332	1577023005	31/10/2019
457	7/11/2023 NR913242S PAKARAK 3-23/1, VEN VENKATRI VENKATRI DOMESTIK	1	1000 HE2308213556H HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	5438	4/1/2023	913030630375	7/11/2023 14/11/2023	342	1577024025	31/09/2009
458	11/7/2024 NR913244K PADHAM 3-37, VEN VENKATRI VENKATRI AGRICULT	1	1000 HP2405693690H HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	2378759940	11/7/2024	91303060953	11/7/2024 18/07/2024	342	1577024025	31/09/2005
459	8/1/2024 NR913244K RAU 3-30/1, VEN VENKATRI VENKATRI AGRICULT	1	500 HE221021510976H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	5469	8/1/2024	913040603567	8/1/2024 16/01/2024	346	1577019007	31/09/2007
460	11/7/2024 NR913244R RAU 3-6, YELKEVLI CHE YELKE CHE DOMESTIK	1	500 HP2405693690H HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2378759940	11/7/2024	91303060949	11/7/2024 18/07/2024	346	1577024025	31/09/2005
473	11/7/2024 NR913244R SHANKA 8- YELKEVLI CHE YELKE CHE GENERAL	1	500 HP2405693690H HPL - ELECTRIC	1	15-30A	1	200	600	29.5	281664	*****	913040601813	*****	332	1577023005	31/10/2019
477	7/6/2024 NR913244S SARA BAL 6-9, MOHOHMANA MATAMP DOMESTIK	1	1000 HP2309355503H HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	2347566555	7/6/2024	91303060754	7/6/2024 14/06/2024	342	1577024025	31/09/2009
478	12/2/2024 NR913244S D SRINU 3-7/12, M MUNDALI MUNDALI DOMESTIK	1	500 HE22122151803H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	2224441774	12/2/2024	913030604505	12/2/2024 20/02/2024	339	1577013041	31/09/2009
479	12/2/2024 NR913244S SMTI GAN 3-7/13, M MUNDALI MUNDALI DOMESTIK	1	500 HE22122151803H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	2224441774	12/2/2024	913030604505	12/2/2024 20/02/2024	339	1577013041	31/09/2009
480	4/7/2024 NR913244S D SRINU 3-7/13, M MUNDALI MUNDALI DOMESTIK	1	500 HE22122151803H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	2347566555	4/7/2024	91303060933	4/7/2024 14/06/2024	346	1577024025	31/09/2005
481	4/7/2024 NR913244S D SRINU 3-7/13, M MUNDALI MUNDALI DOMESTIK	1	500 HE22122151803H HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	281666	11/7/2024	91303060949	11/7/2024 18/07/2024	346	1577024025	31/09/2005
482	26/10/2023 NR913243S MIDAVANA 7-21 BOR BORING T GUVVANI DOMESTIK	1	1000 HE2308213561B HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	30514	27/10/2023	91303061983	26/10/2023 28/10/2023	360	1577023005	31/10/2019
483	9/7/2024 NR913244S MIDAVANA 7-21 BOR BORING T GUVVANI DOMESTIK	1	1000 HP2405693702L HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	30514	27/10/2024	91303062030	9/7/2024 14/06/2024	360	1577023005	31/10/2019
484	19/09/2024 NR913244S K LAXMAN 7-72, MAM MADRA M MADRA DOMESTIK	1	1000 HE23122151803H HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	30514	25/03/2024	913030605112	19/09/2024 23/03/2024	361	1577023015	31/10/2010
485	19/04/2024 NR913242S CHOWDAN 7-72, MAM MADRA M MADRA M NON-DOM	1	1000 HP240133567632H HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	305159	15/03/2024	913030605429	19/04/2024 25/04/2024	361	1577023015	31/10/2010
486	4/7/2024 NR913244S D SHANAK 2-1/24 M MUNDALI MUNDALI DOMESTIK	1	500 HP24036937440 HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2370582146	4/7/2024	91303060903	4/7/2024 12/7/2024	359	1577112025	31/09/2002
487	26/10/2023 NR913244S D SHANAK 2-1/24 M MUNDALI MUNDALI DOMESTIK	1	1000 HE2308213561B HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	281666	11/7/2023	91303060983	26/10/2023 28/10/2023	360	1577023005	31/10/2019
488	26/07/2024 NR913244S D SHANAK 2-1/24 M MUNDALI MUNDALI DOMESTIK	1	1000 HP2403693702L HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	30514	27/10/2024	91303060924	26/07/2024 28/10/2024	360	1577023005	31/10/2019
489	3/8/2024 NR913244S D SHANAK 2-1/24 M MUNDALI MUNDALI DOMESTIK	1	1000 HP2403693702L HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	30534	2/8/2024	913030601104	3/8/2024 7/8/2024	361	1577023015	31/10/2010
490	3/8/2024 NR913244S D SHANAK 2-1/24 M MUNDALI M MADRA M MADRA M NON-DOM	1	1000 HP2403693702L HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	30533	2/8/2024	913030601105	3/8/2024 7/8/2024	361	1577023015	31/10/2010
491	12/1/2024 NR913244S D SHANAK 2-1/24 M MUNDALI M MADRA M MADRA M NON-DOM	1	1000 HE23102150886H HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	519674	12/1/2024	9130304003777	12/1/2024 17/01/2024	362	1577023015	31/10/2010
492	23/09/2024 NR913244S D SHANAK 2-1/24 M MUNDALI M MADRA M MADRA M NON-DOM	1	1000 HP2403693702L HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	30534	10/9/2024	91303061375	23/09/2024 14/10/2024	362	1577023015	31/10/2010
493	3/1/2024 NR913244S D SHANAK 2-1/24 M MUNDALI M MADRA M MADRA M NON-DOM	1	500 HE23122151803H HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	5621	2/1/2024	91303061376	3/1/2024 17/01/2023	363	1577023015	31/10/2009
494	10/1/2024 NR913242S B NAGEH+8-00, BOY BOY GUD BOY GUD DOMESTIK	1	1000 HE23102150886H HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	305128	6/1/2024	9130306005182	10/1/2024 16/01/2024	363	1577023015	31/10/2009
495	2/4/2024 NR913242S B NAGEH+8-00, BOY BOY GUD BOY GUD AGRICULT	1	1000 HP2401356735H HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	228707876	2/4/2024	913040605183	2/4/2024 15/04/2024	363	1577023015	31/10/2009

500	8/10/2024 NR913244 KATRAVAT 4-151 BABBARANI KIBAPANI KIN-NON-DON	1	500	HP24036938231-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2454056585	8/10/2024 91303061428	8/10/2024 15/10/2024	365	152703401	365000010	
501	11/10/2023 NR913233 K GOPAL 1-12 ,PEDIPEDA TH-CHOWDA TH-PEDA TH-DOMESTIC	2	1000	HP23053539923-HPL - ELECTRIC	1	15-30A	1	800	1416	59	5435 #####	91304001694 #####	11/10/2023 20/10/2023	366	1527028015	365000041	
502	13/12/2023 NR913233 K RAJ KUM-3-04 ,PEDIPEDA TH-CHOWDA TH-PEDA TH-DOMESTIC	1	1000	HE23092138220-HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	22928 #####	913040022822	13/12/2023 19/12/2023	366	1527007003	365000042	
503	13/12/2023 NR913233 KATHRAVAT 4-016 ,PEDIPEDA TH-CHOWDA TH-PEDA TH-DOMESTIC	1	1000	HE23092138220-HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	229301 #####	913040022821	13/12/2023 19/12/2023	366	1527028005	365000043	
504	6/1/2024 NR913243 NANKU B-2-58/3 ,PE PEDDA TH-PEDA TH-DOMESTIC	1	500	HE2310150818-HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	5844 5/1/2024	91304003512	6/1/2024 10/1/2024	366	1527028006	365000044	
505	22/02/2024 NR913243 K THULSIR 3-35 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	1000	HE23122157781-HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	305155 21/02/2024	91304004748	22/02/2024 26/02/2024	366	1527019072	365000045	
506	6/3/2024 NR913243 K SHANKA 3-11 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	1000	HE23122157905-HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	305156 22/02/2024	91304005050	6/3/2024 14/03/2024	366	1527028006	365000046	
507	6/3/2024 NR913243 KATHRAVAT 2-79 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	500	HE23122158262-HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	2262169745	6/3/2024	91304005054	6/3/2024 14/03/2024	366	1527028006	365000047
508	6/3/2024 NR913243 KATHRAVAT 1-42 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	500	HE23122158261-HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	2262208983	6/3/2024	91304005056	6/3/2024 14/03/2024	366	1527028005	365000048
509	6/5/2024 NR913244 K BUJJI BA 3-050 ,PE PEDDA TH-CHOWDA TH-DOMESTIC	1	1000	HP24013567690-HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	281768	3/5/2024	91304005546	6/5/2024 13/05/2024	367	1527007036	365000049
510	23/05/2024 NR913244 KATHRAVAT 3-50/1 ,PE PEDDA TH-CHOWDA TH-DOMESTIC	1	1000	HP24013567117-HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	2355797002	23/05/2024	91303606650	23/05/2024 27/05/2024	366	1527007036	365000050
511	23/05/2024 NR913244 KATHRAVAT 3-50 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	1000	HP24013567118-HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	2355787202	23/05/2024	91303606650	23/05/2024 27/05/2024	366	1527007036	365000051
512	11/6/2024 NR913244 V RAGYA 2-59/1 ,PE PEDDA TH-CHOWDA TH-DOMESTIC	1	1000	HP23093550336-HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	2350991241	11/6/2024	91303606763	11/6/2024 17/06/2024	366	1527007036	365000052
513	24/07/2024 NR913244 KATHRAVAT 3-47 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	500	HP24036937977-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2393892352	24/07/2024	913036061024	24/07/2024 31/07/2024	356	1527007036	365000053
514	14/08/2024 NR913244 K CHEENA 3-09/23 ,P PEDDA TH-CHOWDA TH-DOMESTIC	1	500	HP24036938336-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2416360138	14/08/2024	913036061166	14/08/2024 22/08/2024	356	1527007036	365000054
515	24/08/2024 NR913244 K HAREESI 5-1 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	1000	HP24036938687-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2421882781	24/08/2024	913036061232	24/08/2024 28/08/2024	356	1527007036	365000055
516	24/08/2024 NR913244 KATHRAVAT 3-53 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	500	HP24036938526-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2421882781	24/08/2024	913036061232	24/08/2024 28/08/2024	356	1527007036	365000056
517	24/08/2024 NR913244 KATHRAVAT 2-53 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	500	HP24036938773-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2421892072	24/08/2024	913036061232	24/08/2024 28/08/2024	356	1527007036	365000057
518	2/9/2024 NR913244 K CHATTRI 3-47 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	500	HP24036936919-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2427865608	2/9/2024	913036061263	2/9/2024 10/09/2024	356	1527007036	365000058
519	13/09/2024 NR913244 K CHAMPI 2-64 ,PEDIPEDA TH-CHOWDA TH-DOMESTIC	1	1000	HP24036937514-HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	305238	6/8/2024	913036061125	13/09/2024 23/09/2024	356	1527007036	365000059
520	16/10/2024 NR913244 K LALLI BA 3-09/24 ,P PEDDA TH-CHOWDA TH-DOMESTIC	1	500	HP24036938526-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2454666040	16/10/2024	913036061457	16/10/2024 27/10/2024	356	1527007036	365000060
521	6/3/2024 NR913243 LAXMII 3-100 ,GU GUNGYA 9-GUNGYA DOMESTIC	1	1000	HP23122158261-HIMACHAL ENERGY	1	15-30A	1	200	600	29.5	242188266	6/3/2024	913036061232	24/08/2024 28/08/2024	357	1527005030	365000003
522	24/07/2024 NR913244 VISLAVATI 4-87 ,KUCHUKHAMI CHOWDA NON-DON	1	500	HP24036937977-HPL - ELECTRIC	1	15-30A	1	200	600	29.5	2393871278	24/07/2024	913036061024	24/07/2024 31/07/2024	358	1527101056	365000008
523	8/9/2023 NR913233 HIRHA MAR 8-36/6 ,NAN NANCHAR DESAI PALI DOMESTIC	2	5000	7796292-HPL - ELECTRIC	3	3-50A	1	4000	7080	59	5156	8/9/2023	91304008680	8/9/2023 20/09/2023	359	1527270719	365001688
524	8/9/2023 NR913233 NAUKKA GH 9-27 ,NAN NANCHAR DESAI PALI DOMESTIC	1	1000	HP23053539759-HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5149	8/9/2023	91304008681	8/9/2023 20/09/2023	359	1527007036	365000059
525	8/9/2023 NR913233 NAUKKA GH 9-27 ,NAN NANCHAR DESAI PALI DOMESTIC	1	1000	HP23053539759-HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5148	8/9/2023	91304008682	8/9/2023 20/09/2023	359	1527007036	365000056
526	8/9/2023 NR913233 NAUKKA GH 9-27 ,NAN NANCHAR DESAI PALI DOMESTIC	1	1000	HP23053539759-HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5147	8/9/2023	91304008683	8/9/2023 20/09/2023	359	1527007036	3650001691
527	23/09/2023 NR913233 DONGALI SYNO 452 DESAI PALI DESAI PALI NON-DON	1	2000	HP23053539334-HPL - ELECTRIC	1	15-30A	1	400	2832	29.5	5279	22/09/2023	91304001252	23/09/2023 30/09/2023	359	1527101056	365000008
528	9/10/2023 NR913233 BALU SHA SYNO 453 NANCHAR DESAI PALI DOMESTIC	2	1000	HP23053539490-HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5149	8/9/2023	91304002860	8/9/2023 20/09/2023	359	1527270719	365001688
529	11/10/2023 NR913233 D LAXMI 9-84 ,NAN NANCHAR DESAI PALI DOMESTIC	1	1000	HP23053539922-HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5148	8/9/2023	91304002861	8/9/2023 20/09/2023	359	1527015060	365001689
530	11/10/2023 NR913233 SRI CHEENI 6-89/31 ,NAN NANCHAR DESAI PALI DOMESTIC	1	1000	HP23053539922-HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5147	8/9/2023	91304002862	8/9/2023 20/09/2023	359	1527015060	365001690
531	11/10/2023 NR913233 SWT KRISHI 5-85/10 ,NAN NANCHAR DESAI PALI NON-DON	1	1000	HP230821354522-HIMACHAL ENERGY	1	15-30A	1	200	1416	29.5	5483	1/10/2023	91304001837	1/10/2023 26/10/2023	359	1527015060	365001691
532	18/10/2023 NR913233 SMT KRISHI 5-85/10 ,NAN NANCHAR DESAI PALI NON-DON	2	1000	HP230821354522-HIMACHAL ENERGY	1	15-30A	1	800	1416	59	5478	1/10/2023	91304001838	18/10/2023 26/10/2023	359	1527015060	365001692
533	18/10/2023 NR913233 T ANANT 1-17 ,NAN NANCHAR DESAI PALI DOMESTIC	2	1000	HP23082135591-HIMACHAL ENERGY	1	15-30A	1	800	1416	59	281650	3/10/2023	91304001617	9/10/2023 13/10/2023	359	1527270719	365001695
534	1/11/2023 NR913233 GANGARA 8-16 ,NAN NANCHAR DESAI PALI DOMESTIC	1	1000	HP23082135592-HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5447	9/10/2023	91304001692	9/10/2023 20/10/2023	359	1527015060	365001693
535	1/11/2023 NR913233 GANGARA 8-16 ,NAN NANCHAR DESAI PALI NON-DON	1	1000	HP23082135592-HPL - ELECTRIC	1	15-30A	1	200	1416	29.5	5448	9/10/2023	91304001693	9/10/2023 20/10/2023	359	1527015060	365001699
536	2/11/2023 NR913233 DEPOT M/T SRIC BUI NANCHAR NANCHAR NON-DON	2	8000	7796485-HPL - ELECTRIC	3	3-50A	1	6400	9600	59	458223	19/10/2023	91304002106	2/11/2023 9/11/2023	359	1527015060	3650001702
537	18/11/2023 NR913233 MOHAMAI 8-43 ,DESIDESAI PALI DESAI PALI DOMESTIK	2	1000	HP23091338061-HIMACHAL ENERGY	1	15-30A	1	800	1416	59	5479	17/10/2023	913040021842	18/10/2023 26/10/2023	360	1527015060	3650001704
538	18/11/2023 NR913233 DONGALA 9-85 ,DESIDESAI PALI DESAI PALI DOMESTIK	1	1000	HP230921330865-HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	5661	17/11/2023	913040022556	18/11/2023 26/11/2023	360	1527015060	3650001709
539	18/11/2023 NR913233 MYADAM C9-92 ,DESIDESAI PALI DESAI PALI DOMESTIK	1	1000	HP230921330855-HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	5657	17/11/2023	91304002280	18/11/2023 26/11/2023	360	1527015060	3650001720
540	1/12/2023 NR913233 K CHAND-SYNO 380 NANCHAR DESAI PALI NON-DON	1	1000	HP23092133087-HIMACHAL ENERGY	1	15-30A	1	200	1200	29.5	5659	1/12/2023	91304002280	1/12/2023 26/12/2023	360	1527015060	3650001720
541	6/3/2024 NR913243 SHASHI KL 3-14/27 ,NANCHAR DESAI PALI NON-DON	2	1000	HP23122158261-HIMACHAL ENERGY	1	15-30A	1	800	1200	29.5	2262140570	6/3/2024	91304005053	6/3/2024 14/03/2024	369	1527015060	3650001745
542	13/03/2024 NR913243 DR APARI SYNO 413,414/1A1 NANCHAR DOMESTIK	2	1000	HP23122157875-HIMACHAL ENERGY	1	15-30A	1	2400	1200	29.5	2265945329	13/03/2024	913040051135	13/03/2024 18/03/2024	369	1527015060	3650001747
543	13/03/2024 NR913243 YERUKALI 8-50 ,NAN NANCHAR DESAI PALI AGRICULT	1	1000	HP24036938336-HPL - ELECTRIC	1	15-30A	1	200	1200	29.5	19892	29/02/202					

509	6/5/2024	NR913244 K BUJI BA 3-050 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	1000 HP24013567650HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	2335797002	23/05/2024	91303060650	23/05/2024	27/05/2024	366	1527007036	366000050	
510	23/05/2024	NR913244 KATRAVA 3-60/1 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	1000 HP24013567117HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	2335787827	23/05/2024	91303060650	23/05/2024	27/05/2024	366	1527007036	366000051	
511	23/05/2024	NR913244 KATRAVA 3-60 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	1000 HP24013567118HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	2335787827	23/05/2024	91303060650	23/05/2024	27/05/2024	366	1527007036	366000052	
512	11/6/2024	NR913244 VRAGYA 2-69/1 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	1000 HP240393553375HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	2350991241	11/6/2024	91303060763	11/6/2024	17/06/2024	366	1527007036	366000052	
513	24/07/2024	NR913244 KATRAVA 3-47 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	500 HP24036937977HPL - ELECTRIC	1	5-30A	1	200	600	29.5	2393897352	24/07/2024	91303061166	14/08/2024	23/08/2024	366	1527007036	366000053	
514	14/08/2024	NR913244 K CHEENA 3-09/3 ,PEDIA TH-CHOWDAIDOMESTIC	1	500 HP24036938335HPL - ELECTRIC	1	5-30A	1	200	600	29.5	2413630138	14/08/2024	91303061166	14/08/2024	23/08/2024	366	1527007036	366000054	
515	24/08/2024	NR913244 K HARRESI 1-51 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	1000 HP24036936871HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	2421887781	24/08/2024	91303061222	24/08/2024	28/08/2024	366	1527007036	366000055	
516	16/10/2024	NR913244 K LAXMAN 1-53 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	500 HP24036938526HPL - ELECTRIC	1	5-30A	1	200	600	29.5	2421888770	24/08/2024	91303061222	24/08/2024	28/08/2024	366	1527007036	366000056	
517	24/08/2024	NR913244 KATRAVA 2-53 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	500 HP24036937758HPL - ELECTRIC	1	5-30A	1	200	600	29.5	2421893072	24/08/2024	91303061222	24/08/2024	28/08/2024	366	1527007036	366000057	
518	27/09/2024	NR913244 K CHAITRI 3-47 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	500 HP24036936919HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	2427865408	27/09/2024	91303061263	27/09/2024	10/09/2024	366	1527007036	366000058	
519	13/09/2024	NR913244 K CHAMP 2-64 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	1000 HP24036938514HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	305238	6/8/2024	91303061325	13/09/2024	19/09/2024	366	1527007036	366000059	
520	16/10/2024	NR913244 K LALLI BA 3-09/21 ,PEPEDIA TH-CHOWDAIDOMESTIC	1	500 HP24036938526HPL - ELECTRIC	1	5-30A	1	200	600	29.5	2459466640	16/10/2024	91303061457	16/10/2024	24/10/2024	366	1527007036	366000060	
521	6/3/2024	NR913244 K LAXMAN 1-100 ,GUENGUAI GUNGUNA DOMESTIC	1	500 HE232122161586HIMACHAL ENERGY	1	5-30A	1	200	600	29.5	2421888770	24/08/2024	91303061222	24/08/2024	28/08/2024	367	1527007036	366000061	
522	24/07/2024	NR913244 VISLAVITA 4-87 ,KUCIKUCHAMI CHOWDAIDOMESTIC	1	500 HP24036937971HPL - ELECTRIC	1	5-30A	1	200	600	29.5	2393871278	24/07/2024	91303061024	24/07/2024	31/07/2024	368	1527310166	368000008	
523	8/9/2023	NR913244 HIRA NAR 8-36/6 ,NANANCHAR DESPALIDOMESTIC	2	5000 -776291HPL - ELECTRIC	3	5-30A	1	4000	7080	59	5156	8/9/2023	91304000860	8/9/2023	20/09/2023	369	1527327019	366000168	
524	8/9/2023	NR913244 NANAKA GH 9-27 ,NANANCHAR DESPALIDOMESTIC	1	1000 HP23053539759HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5149	8/9/2023	91304000861	8/9/2023	20/09/2023	369	1527019066	366000169	
525	8/9/2023	NR913244 NANAKA GH 9-27 ,NANANCHAR DESPALIDOMESTIC	1	1000 HP23053539759HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5148	8/9/2023	91304000862	8/9/2023	20/09/2023	369	1527019028	366000169	
526	8/9/2023	NR913244 NANAKA GH 9-27 ,NANANCHAR DESPALIDOMESTIC	1	1000 HP23053539759HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5147	8/9/2023	91304000863	8/9/2023	20/09/2023	369	1527019055	3660001691	
527	23/09/2023	NR913244 DONGALI SYNO 452 ,DESPALIDOMAINON-DOM	1	2000 HP23053539374HPL - ELECTRIC	1	5-30A	1	200	600	29.5	5279	22/09/2023	91304001252	23/09/2023	30/09/2023	369	1527019005	3660001692	
528	9/10/2023	NR913244 BALU SHIA SYNO 453 ,NANANCHAR DESPALIDOMESTIC	2	1000 HP23053539490HPL - ELECTRIC	1	5-30A	1	200	800	1416	59	281650	3/10/2023	91304001617	3/10/2023	13/10/2023	369	1527019055	3660001695
529	11/10/2023	NR913244 D LAXMI 9-84 ,NANANCHAR DESPALIDOMESTIC	1	1000 HP23053539294HPL - ELECTRIC	1	5-30A	1	200	1416	29.5	5447	11/10/2023	91304001692	11/10/2023	20/10/2023	369	1527019038	3660001698	
530	11/10/2023	NR913244 SMT KRISH 5-85/10 ,NANANCHAR DESPALIDOMESTIC	1	1000 HE23082153452HIMACHAL ENERGY	1	5-30A	1	200	1416	29.5	5483	17/10/2023	91304001693	17/10/2023	28/10/2023	369	1527019028	3660001702	
531	18/10/2023	NR913244 SMT KRISH 5-85/10 ,NANANCHAR DESPALIDOMESTIC	2	1000 HE23082153452HIMACHAL ENERGY	1	5-30A	1	200	1416	29.5	5478	17/10/2023	91304001838	18/10/2023	23/10/2023	369	1527019028	3660001703	
532	18/10/2023	NR913244 T ANANT 1-17 ,NANANCHAR DESPALIDOMESTIC	2	1000 HE23082153591HIMACHAL ENERGY	1	5-30A	1	200	1416	29.5	5479	17/10/2023	91304001842	18/10/2023	23/10/2023	369	1527019028	3660001704	
533	18/10/2023	NR913244 TANANT 1-17 ,NANANCHAR DESPALIDOMESTIC	1	1000 HE23082153590HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	5495	10/10/2023	91304002107	11/10/2023	21/10/2023	369	1527019028	3660001705	
534	1/1/2023	NR913244 GANGADA 8-16 ,NANANCHAR DESPALIDOMESTIC	1	1000 HE23082153590HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	109600	30/10/2023	91304002106	1/11/2023	3/11/2023	369	1527019028	3660001706	
535	1/1/2023	NR913244 GANGADA 8-16 ,NANANCHAR DESPALIDOMESTIC	2	8000 -776485HPL - ELECTRIC	3	5-30A	1	6400	9600	59	458923	19/10/2023	91304002180	2/11/2023	9/11/2023	369	1527019029	3660001713	
536	2/1/2023	NR913244 DEPOT M TH-TRIC BU/NANANCHAR NANANCHAR NON-DOM	2	1000 HE23091318061HIMACHAL ENERGY	1	5-30A	1	800	1200	59	5668	17/11/2023	91304002552	18/11/2023	23/11/2023	369	1527019032	3660001718	
537	18/11/2023	NR913244 DONGALA 9-85 ,DES PALIDOMESTIC	1	1000 HE23091318061HIMACHAL ENERGY	1	5-30A	1	800	1200	29.5	5661	17/11/2023	91304002552	18/11/2023	26/11/2023	369	1527019032	3660001719	
538	18/11/2023	NR913244 MYADAN C/9-92 ,DESPALIDOMESTIC	1	1000 HE23091318085HIMACHAL ENERGY	1	5-30A	1	800	1200	29.5	5657	17/11/2023	91304002556	18/11/2023	26/11/2023	369	1527019032	3660001720	
539	6/12/2023	NR913244 K CHAND SYNO 380 ,NANANCHAR DESPALINON-DOM	1	1000 HE23092138107HIMACHAL ENERGY	1	5-30A	1	600	1200	29.5	5809	5/12/2023	91304002809	6/12/2023	13/12/2023	369	1527019029	3660001721	
540	6/12/2023	NR913244 K CHAND SYNO 380 ,NANANCHAR DESPALINON-DOM	2	1000 HE231221281626HIMACHAL ENERGY	1	5-30A	1	800	1200	59	22621405761	6/3/2024	91304005533	6/3/2024	14/03/2024	369	1527019028	3660001724	
541	6/3/2024	NR913244 SHASHI KL 3-14/21 ,NANANCHAR DESPALINON-DOM	2	1000 HE231221281587HIMACHAL ENERGY	1	5-30A	1	800	1200	59	22669495329	13/03/2024	91304005125	13/03/2024	26/03/2024	369	1527019028	3660001725	
542	13/03/2024	NR913244 APARIN SYNO 413,414/MA1 ,NANANCHAR DOMESTIC	1	1000 HE231221281587HIMACHAL ENERGY	1	5-30A	1	800	1200	59	22669495329	13/03/2024	91304005125	13/03/2024	26/03/2024	369	1527019028	3660001726	
543	18/03/2024	NR913244 YERUKO 4-55/1 ,AN AMUDAL DOMESTIC	1	1000 HP24036938336HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	24132062	14/08/2024	91303061164	14/08/2024	22/08/2024	369	1527019028	3660001728	
544	14/08/2024	NR913244 VADEEL VE 1-16 ,NANANCHAR DESPALIDOMESTIC	1	1000 HP24036938277HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	2432765943	9/9/2024	91303061228	9/9/2024	19/09/2024	369	1527019029	3660001729	
545	9/9/2024	NR913244 KARUPAKI 7-81/2 ,NANANCHAR DESPALIDOMESTIC	1	1000 HP240369387912HPL - ELECTRIC	1	5-30A	1	200	1200	29.5	24327569328	9/9/2024	91303061228	9/9/2024	19/09/2024	369	1527019029	3660001730	
546	9/9/2024	NR913244 KARUPAKI 7-81/2 ,NANANCHAR DESPALIDOMESTIC	1	1000 HE231221281587HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	24327569328	9/9/2024	91303061228	9/9/2024	19/09/2024	369	1527019029	3660001731	
547	9/9/2024	NR913244 KARUPAKI 7-81/2 ,NANANCHAR DESPALIDOMESTIC	1	1000 HE231221281587HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	24327569328	9/9/2024	91303061228	9/9/2024	19/09/2024	369	1527019029	3660001732	
548	9/9/2024	NR913244 KARUPAKI 7-81/2 ,NANANCHAR DESPALIDOMESTIC	1	1000 HE231221281587HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	24327569328	9/9/2024	91303061228	9/9/2024	19/09/2024	369	1527019029	3660001733	
549	9/9/2024	NR913244 KARUPAKI 7-81/2 ,NANANCHAR DESPALIDOMESTIC	1	1000 HE231221281587HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	24327569328	9/9/2024	91303061228	9/9/2024	19/09/2024	369	1527019029	3660001734	
550	9/9/2024	NR913244 KARUPAKI 7-81/2 ,NANANCHAR DESPALIDOMESTIC	1	1000 HE231221281587HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	24327569328	9/9/2024	91303061228	9/9/2024	19/09/2024	369	1527019029	3660001735	
551	20/02/2024	NR913244 LAUDIYA 4-65/1 ,AN AMUDAL DOMESTIC	1	1000 HE231221281587HIMACHAL ENERGY	1	5-30A	1	200	1200	29.5	24327569328	9/9/2024	91303061228	9/9/2024	19/09/2024	369	1527019029	36600	

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