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Portfolio

DATA ANALYTICS



PROFESSIONAL BACKGROUND

**CURRENTLY I AM IN THIED YEAR
YEAR PURSUING B.TECH-CSE. I HAVE
SECURED 7.5 CGPA (TILL 4TH SEM) AND
HAVE SEVERAL SKILLS
INCLUDING DATA ANALYSIS, PYTHON.**

**I AM GLAD TO TELL THAT I HAVE WORK
WITH TRAINITY COURSE AND LEARNA
ALL THE DATA ANALYSIS TOOL .I
REALLY APRICIATE TRAINITY TO
PROVIDE SUCH CONTENT.**

**I HAD WORK WITH DIFFRENT
PROJECTS WHICH ARE GIVEN IN THE
TRANIITIY COURSE .**

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1.DATA ANALYTICS PROCESS

THE PROBLEM

YOUR TASK IS TO GIVE THE EXAMPLE(S) OF SUCH A REAL-LIFE SITUATION WHERE WE USE DATA ANALYTICS AND LINK IT WITH THE DATA ANALYTICS PROCESS. YOU CAN PREPARE A PPT/PDF ON A REAL-LIFE SCENARIO EXPLAINING IT WITH THE ABOVE PROCESS (PLAN, PREPARE, PROCESS, ANALYZE, SHARE, ACT) AND SUBMIT IT AS PART OF THIS TASK.

SOLUTION

PLAN: I WANT TO BUY GROCERY FOR THE WHOLE MONTH.

PREPARE: I WILL BE LOOKING HOW MUCH IT IS GOING TO COST ME TO BUY THE GROCERY FOR THE ONE WHOLE MONTH. AND WHAT IS THE BUGGET FOR THE WHOLE MONTH.

PROCESS: I WILL GO KITCHEN AND LOOK FOR WHAT ARE THE NECESSARY THING THAT I WANT TO BUY AND WHAT ARE THE THINGS THAT I CAN SKIP TO BUY.

ANALYZE: OBVIOUSLY I WAN'T BUY THINGS THAT ARE NOT USED IN KITCHEN AND I WILL GOING TO LOOK FOR THE EXPIRE DATES OF THE PRODUCT SO IT CAN BE USED FOR THE LONG TIME .

SHARE: I WILL SHARE THE LIST OF THE ITEAMS THAT I NEED TO BUY TO THE SHOPEKIPPER OR IF I VISITE ANY GROCERY MALL/STORE I WILL WALK AROUND AND LOOK FOR THE ITEAMS ACCORDING TO THE LIST SO I DON'T SKIP ANY PRODUCT.

ACT: AFTER SHARING THE LIST OR WALKING IN THE MARKET AND COLLECTING THE ITEAM FINALLY I BUY IT WITHIN THE BUGET .

2.INSTAGRAM USER ANALYTICS

I have been provided the dataset which contains the usage of instagram with different users and their interaction with the instagram. I have to perform the analysis on mysql workbench.

THE PROBLEM

1.Loyal User Reward:

Your Task: Identify the five oldest users on Instagram from the provided database.

2.Inactive User Engagement:

Your Task: Identify users who have never posted a single photo on Instagram.

3.Contest Winner Declaration:

Your Task: Determine the winner of the contest and provide their details to the team.

4. Hashtag Research: Your

Task: Identify and suggest the top five most commonly used hashtags on the platform.

5. Ad Campaign Launch: Your

Task: Determine the day of the week when most users register on Instagram. Provide insights on when to schedule an ad campaign.

SOLUTION

FINDING 1

```
1  -- The oldest user of the instagram
2  • select created_at , username
3  from users
4  order by created_at asc
5  limit 5;
6
7
```

Result Grid



Filter Rows:

Export:



Wrap Cell C

	created_at	username
▶	2016-05-06 00:14:21	Darby_Herzog
	2016-05-06 13:04:30	Emilio_Bernier52
	2016-05-08 01:30:41	Elenor88
	2016-05-09 17:30:22	Nicole71
	2016-05-14 07:56:26	Jordyn.Jacobson2

WE CAN CLEARLY SEE THE 5 OLDEST USER OF THE INSTAGRAM

FINDING 2

```
select username  
from users  
where id not in (select user_id from photos);
```

username	
Aniya_Hackett	
Kasandra_Homenick	
Jaclyn81	
Rocio33	
Maxwell.Halvorson	
Tierra.Trantow	
Pearl7	
Ollie_Ledner37	
Mckenna17	
David.Osinski47	
Morgan.Kassulke	
Linnea59	
Duane60	
Julien_Schmidt	
Mike.Auer39	
Franco_Keebler64	
Nia_Haag	
Hulda.Macejkovic	
Leslie67	
Janelle.Nikolaus81	
Darby_Herzog	
Esther.Zulauf61	

Hulda.Macejkovic

Leslie67

Janelle.Nikolaus81

Darby_Herzog

Esther.Zulauf61

Bartholome.Bernhard

Jessyca_West

Esmeralda.Mraz57

Bethany20

WE CAN CLEARLY SEE THIS TH
PERSON WHO HAVE NEVER
POSTED ON INSTAGRAM

FINDING 3

```
SELECT username, photos.id AS photo_id, COUNT(photo_id)
FROM photos
JOIN likes ON photos.id = likes.photo_id
JOIN users ON photos.user_id = users.id
GROUP BY photos.id, users.username
ORDER BY like_count DESC
LIMIT 1;
```

	username	photo_id	like_count
▶	Zack_Kemmer93	145	48

**FROM THE ALL MEMBER OF
THE INSTAGRAM
ZACK_KEMMER93 HAS GOT
THE MOST LIKED ON SINGLE
PHOTO AND WON THE
CONTEST.**

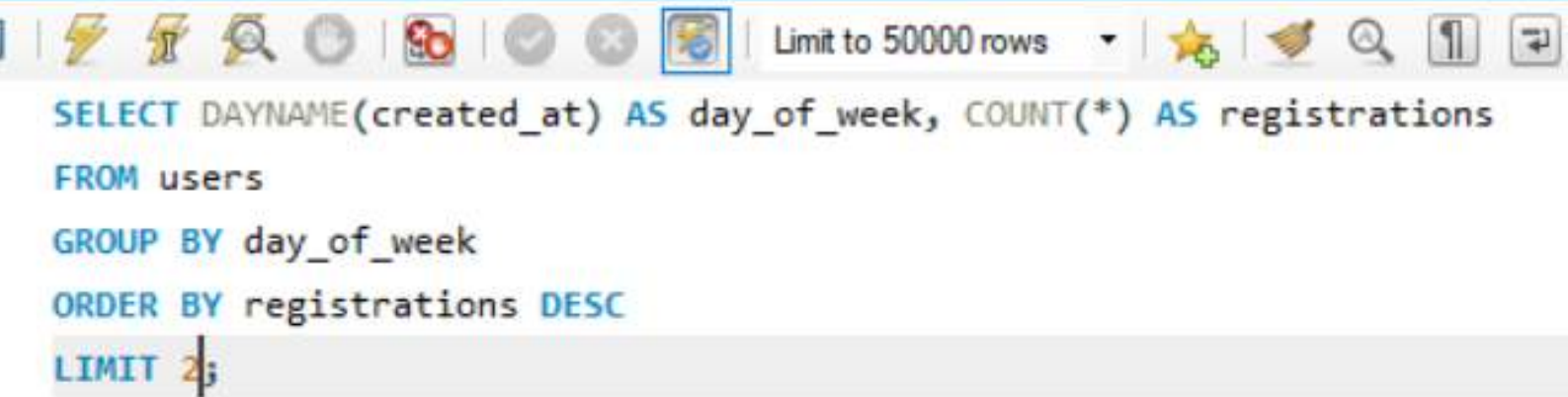
FINDING 4.

```
• SELECT tags.tag_name, COUNT(photo_tags.tag_id) AS usage_count  
FROM tags |  
JOIN photo_tags ON tags.id = photo_tags.tag_id  
GROUP BY tags.tag_name  
ORDER BY usage_count DESC  
LIMIT 5;
```

	tag_name	usage_count
▶	smile	59
	beach	42
	party	39
	fun	38
	concert	24

**THE ABOVE ARE THE MOST
USED HASHTAG ON THE
INSTAGRAM**

FINDING 5



The screenshot shows a SQL query editor with a toolbar at the top. The query is as follows:

```
SELECT DAYNAME(created_at) AS day_of_week, COUNT(*) AS registrations
FROM users
GROUP BY day_of_week
ORDER BY registrations DESC
LIMIT 2;
```

	day_of_week	registrations
▶	Thursday	16
	Sunday	16

**FROM THE WHOLE WEEK
THURSDAY AND SUNDAY
ARE THE DAY TO LAUNCH AD.**

CONCLUSION

I WOULD HAVE ANALYSE THAT EACH TASK HAS ITS OWN VALUE WITHIN. THE MOST INSIGHT THAT COME OUT IS THAT SUNDAY AND THURSDAY ARE THE DAY TO LAUNCH ANY AD. AND IF WE DO PARTNERSHIP WITH ZACK_KEMMER93 IT WILL BE BENEFICIAL AS IT GOT MOST LIKE ON THE SINGLE PHOTO. AND WE CAN TAKE HELP OF THE OLDEST USER TO KNOW MORE ABOUT THE INSTAGRAM AND FEATURES FROM EARLIER TILL NOW.

3. OPERATION ANALYTICS AND INVESTIGATING METRIC SPIKE

IN THIS I HAVE BEEN PROVIDED 2 DATASET CASE STUDY 1: JOB DATA ANALYSIS

WORKING WITH FOLLOWING COLUMNS:

- JOB_ID: UNIQUE IDENTIFIER OF JOBS
- ACTOR_ID: UNIQUE IDENTIFIER OF ACTOR
- EVENT: THE TYPE OF EVENT
(DECISION/SKIP/TRANSFER).
- LANGUAGE: THE LANGUAGE OF THE
CONTENT
- TIME_SPENT: TIME SPENT TO REVIEW THE
JOB IN SECONDS.
- ORG: THE ORGANIZATION OF THE ACTOR
- DS: THE DATE IN THE FORMAT
YYYY/MM/DD (STORED AS TEXT).

CASE STUDY 2: INVESTIGATING METRIC SPIKE

WORKING WITH FOLLOWING COLUMNS

- USERS: CONTAINS ONE ROW PER USER, WITH DESCRIPTIVE INFORMATION ABOUT THAT USER'S ACCOUNT.
- EVENTS: CONTAINS ONE ROW PER EVENT, WHERE AN EVENT IS AN ACTION THAT A USER HAS TAKEN (E.G., LOGIN, MESSAGING, SEARCH).
- EMAIL_EVENTS: CONTAINS EVENTS SPECIFIC TO THE SENDING OF EMAILS.

**ALL THE ANALYSE HAVE BEEN DONE ON
MYSQLWORK BENCH**

CASE STUDY 1: JOB DATA ANALYSIS

TASK TO PERFORM

- **JOBS REVIEWED OVER TIME:**

- OBJECTIVE: CALCULATE THE NUMBER OF JOBS REVIEWED PER HOUR FOR EACH DAY IN NOVEMBER 2020.
- YOUR TASK: WRITE AN SQL QUERY TO CALCULATE THE NUMBER OF JOBS REVIEWED PER HOUR FOR EACH DAY IN NOVEMBER 2020.

- **THROUGHPUT ANALYSIS:**

- OBJECTIVE: CALCULATE THE 7-DAY ROLLING AVERAGE OF THROUGHPUT (NUMBER OF EVENTS PER SECOND).
- YOUR TASK: WRITE AN SQL QUERY TO CALCULATE THE 7-DAY ROLLING AVERAGE OF THROUGHPUT. ADDITIONALLY, EXPLAIN WHETHER YOU PREFER USING THE DAILY METRIC OR THE 7-DAY ROLLING AVERAGE FOR THROUGHPUT, AND WHY.

- **LANGUAGE SHARE ANALYSIS:**

- OBJECTIVE: CALCULATE THE PERCENTAGE SHARE OF EACH LANGUAGE IN THE LAST 30 DAYS.
- YOUR TASK: WRITE AN SQL QUERY TO CALCULATE THE PERCENTAGE SHARE OF EACH LANGUAGE OVER THE LAST 30 DAYS.

- **DUPLICATE ROWS DETECTION:**

- OBJECTIVE: IDENTIFY DUPLICATE ROWS IN THE DATA.
- YOUR TASK: WRITE AN SQL QUERY TO DISPLAY DUPLICATE ROWS FROM THE JOB_DATA TABLE.

CASE STUDY 2: INVESTIGATING METRIC SPIKE

TASK TO PERFORM

- **WEEKLY USER ENGAGEMENT:**

- OBJECTIVE: MEASURE THE ACTIVENESS OF USERS ON A WEEKLY BASIS.
- YOUR TASK: WRITE AN SQL QUERY TO CALCULATE THE WEEKLY USER ENGAGEMENT.

- **USER GROWTH ANALYSIS:**

- OBJECTIVE: ANALYZE THE GROWTH OF USERS OVER TIME FOR A PRODUCT.
- YOUR TASK: WRITE AN SQL QUERY TO CALCULATE THE USER GROWTH FOR THE PRODUCT.

- **WEEKLY RETENTION ANALYSIS:**

- OBJECTIVE: ANALYZE THE RETENTION OF USERS ON A WEEKLY BASIS AFTER SIGNING UP FOR A PRODUCT.
- YOUR TASK: WRITE AN SQL QUERY TO CALCULATE THE WEEKLY RETENTION OF USERS BASED ON THEIR SIGN-UP COHORT.

- **WEEKLY ENGAGEMENT PER DEVICE:**

- OBJECTIVE: MEASURE THE ACTIVENESS OF USERS ON A WEEKLY BASIS PER DEVICE.
- YOUR TASK: WRITE AN SQL QUERY TO CALCULATE THE WEEKLY ENGAGEMENT PER DEVICE.

- **EMAIL ENGAGEMENT ANALYSIS:**

- OBJECTIVE: ANALYZE HOW USERS ARE ENGAGING WITH THE EMAIL SERVICE.
- YOUR TASK: WRITE AN SQL QUERY TO CALCULATE THE EMAIL ENGAGEMENT METRICS.

CASE STUDY 1: JOB DATA ANALYSIS

FINDING 1

```
SELECT
    ds AS review_date,
    COUNT(job_id) AS jobs_reviewed
FROM
    job_data
WHERE
    ds BETWEEN '11/01/2020' AND '11/30/2020'
GROUP BY
    review_date
ORDER BY
    review_date;
```

▶	11/25/2020	1
	11/26/2020	1
	11/27/2020	1
	11/28/2020	2
	11/29/2020	1
	11/30/2020	2

**WE CAN SEE THE
JOB REVIEWED
IN NOVEMBER**

FINDING 2

```
WITH daily_counts AS (  
    SELECT  
        ds AS review_date,  
        COUNT(job_id) AS event_count  
    FROM  
        job_data  
    GROUP BY  
        review_date  
) ,  
throughput AS (  
    SELECT  
        review_date,  
        event_count,  
        CAST(event_count AS FLOAT) / CAST(86400 AS FLOAT) AS throughput_per_sec  
    FROM  
        daily_counts  
)  
SELECT  
    AVG(throughput_per_sec) OVER (  
        ORDER BY review_date  
        ROWS BETWEEN 6 PRECEDING AND CURRENT ROW  
    ) AS rolling_avg_7_day_throughput  
FROM  
    throughput  
ORDER BY  
    review_date;
```

review_date	throughput_per_sec	rolling_avg_7_day_throughput
11/25/2020	0.000011574074074074073	0.000011574074074074073
11/26/2020	0.000011574074074074073	0.000011574074074074073
11/27/2020	0.000011574074074074073	0.000011574074074074073
11/28/2020	0.000023148148148148147	0.000014467592592592593
11/29/2020	0.000011574074074074073	0.000013888888888888889
11/30/2020	0.000023148148148148147	0.0000154320987654321

- WE CAN CLEARLY SEE THE 7 DAYS ROLLING, I WOULD CHOOSE A 7-DAY ROLLING AVERAGE BECAUSE IT REDUCES DAILY VARIATIONS BY CALCULATING THE AVERAGE OVER A FULL WEEK. THIS APPROACH HELPS HIGHLIGHT OVERALL TRENDS MORE CLEARLY, MAKING IT EASIER TO SPOT CONSISTENT SHIFTS IN PERFORMANCE WITHOUT BEING MISLED BY UNUSUAL SINGLE-DAY SPIKES OR DROPS.

FINDING 3

```
SELECT COUNT(*) as record_in_last_30days  
FROM job_data  
WHERE STR_TO_DATE(ds, '%m/%d/%Y') >= DATE_SUB(CURDATE(), INTERVAL 30 DAY);
```

	record_in_last_30days
▶	0

- **SINCE THERE ARE NO RECORDS FROM THE PAST 30 DAYS, WE DON'T HAVE ANY DATA TO ANALYZE THE DISTRIBUTION OF LANGUAGES USED DURING THIS PERIOD. THIS LACK OF RECENT DATA PREVENTS US FROM ASSESSING LANGUAGE TRENDS OVER THE LAST MONTH.**

FINDING 4.

```
SELECT ds, job_id, actor_id, `event`, `language`, time_spent, org, COUNT(*) as duplicate_count
FROM job_data
GROUP BY ds, job_id, actor_id, `event`, `language`, time_spent, org
HAVING COUNT(*) > 1;
```

ds	job_id	actor_id	event	language	time_spent	org	duplicate_count
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THERE ARE NO DUPLICATE ROWS IN THE GIVEN DATA SET

CASE STUDY 2: INVESTIGATING METRIC SPIKE

TASK TO PERFORM

FINDING 1

```
WITH user_activity AS (  
    SELECT  
        user_id,  
        YEARWEEK(occurred_at, 1) AS activity_week  
    FROM  
        events  
    GROUP BY  
        user_id,  
        YEARWEEK(occurred_at, 1)  
UNION ALL  
SELECT  
    user_id,  
    YEARWEEK(occurred_at, 1) AS activity_week  
FROM  
    email_events  
GROUP BY  
    user_id,
```

```
        user_id,  
        YEARWEEK(occurred_at, 1) AS activity_week  
FROM  
    email_events  
GROUP BY  
    user_id,  
    YEARWEEK(occurred_at, 1)  
,  
weekly_active_users AS (  
    SELECT  
        ua.activity_week,  
        COUNT(DISTINCT ua.user_id) AS active_users  
FROM  
    user_activity ua  
JOIN  
    users u ON ua.user_id = u.user_id  
WHERE  
    u.activated_at IS NOT NULL
```

```

        u.activated_at IS NOT NULL
    GROUP BY
        ua.activity_week
)
SELECT
    activity_week,
    active_users
FROM
    weekly_active_users
ORDER BY
    activity_week DESC;

```

	activity_week	active_users
▶	201435	4577
	201434	4458
	201433	4320
	201432	4178
	201431	4122
	201430	3987
	201429	3880
	201428	3767
	201427	3656
	201426	3556
	201425	3468
	201424	3355
	201423	3264
	201422	3153
	201421	3082
	201420	3004
	201419	2916
	201418	1408

FROM THE ANALYSIS ABOVE, WE CAN CONCLUDE THAT THE 35TH WEEK OF 2014 HAD THE HIGHEST NUMBER OF ACTIVE USERS, WITH OTHER WEEKS SHOWING LOWER LEVELS OF USER ACTIVITY IN COMPARISON.


FINDING 2

```
WITH MonthlyGrowth AS (  
    SELECT  
        DATE_FORMAT(created_at, '%Y-%m') AS month,  
        COUNT(user_id) AS new_users  
    FROM  
        users  
    GROUP BY  
        month  
)  
GrowthCalculation AS (  
    SELECT  
        month,  
        new_users,  
        LAG(new_users) OVER (ORDER BY month) AS previous_month_users  
    FROM  
        MonthlyGrowth  
)  
SELECT  
    month,  
    new_users,  
    LAG(new_users) OVER (ORDER BY month) AS previous_month_users  
FROM  
    MonthlyGrowth  
)  
SELECT  
    month,  
    new_users,  
    previous_month_users,  
    new_users - COALESCE(previous_month_users, 0) AS growth,  
    CASE  
        WHEN previous_month_users = 0 THEN NULL  
        ELSE (new_users - previous_month_users) * 100.0 / previous_month_users  
    END AS growth_percentage  
FROM  
    GrowthCalculation  
ORDER BY  
    month;
```

	month	new_users	previous_month_users	growth	growth_percentage
►	2013-01	160	NULL	160	NULL
	2013-02	160	160	0	0.00000
	2013-03	150	160	-10	-6.25000
	2013-04	181	150	31	20.66667
	2013-05	214	181	33	18.23204
	2013-06	213	214	-1	-0.46729
	2013-07	284	213	71	33.33333
	2013-08	316	284	32	11.26761
	2013-09	330	316	14	4.43038
	2013-10	390	330	60	18.18182
	2013-11	399	390	9	2.30769
	2013-12	486	399	87	21.80451
	2014-01	552	486	66	13.58025
	2014-02	525	552	-27	-4.89130
	2014-03	615	525	90	17.14286
	2014-04	726	615	111	18.04878
	2014-05	779	726	53	7.30028
	2014-06	873	779	94	12.06675
	2014-07	997	873	124	14.20389
	2014-08	1031	997	34	3.41023

THE ANALYSIS SHOWS THAT IN JANUARY 2013, WE GAINED 160 NEW USERS, WHICH WAS THE HIGHEST COUNT FOR THAT PERIOD. LATER, IN JULY 2014, WE SAW THE LARGEST GROWTH IN NEW USERS, WITH 124 ADDED, MARKING AN INCREASE OF 14.20%.

FINDING 3



```
COUNT(DISTINCT us.user_id) AS total_signups,
COUNT(DISTINCT wa.user_id) AS retained_users,
CASE
    WHEN COUNT(DISTINCT us.user_id) = 0 THEN 0
    ELSE COUNT(DISTINCT wa.user_id) * 100.0 / COUNT(DISTINCT us.user_id)
END AS retention_rate
FROM
    UserSignups us
LEFT JOIN
    WeeklyActivity wa ON us.user_id = wa.user_id
    AND wa.active_date >= us.signup_date
    AND wa.active_date < DATE_ADD(us.signup_date, INTERVAL 4 WEEK)
GROUP BY
    signup_week
ORDER BY
    signup_week;
```

```
    WITH UserSignups AS (  
    SELECT  
        user_id,  
        created_at AS signup_date  
    FROM  
        users
```

```
),  
WeeklyActivity AS (  
    SELECT  
        user_id,  
        created_at AS active_date  
    FROM  
        users  
    WHERE  
        created_at IS NOT NULL  
)
```

```
SELECT  
    DATE_FORMAT(us.signup_date, '%Y-%u') AS signup_week,  
    COUNT(DISTINCT us.user_id) AS total_signups,
```

2014-13	151	151	100.00000	2013-44	92	92	100.00000
2014-14	161	161	100.00000	2013-45	97	97	100.00000
2014-15	166	166	100.00000	2013-46	94	94	100.00000
2014-16	165	165	100.00000	2013-47	82	82	100.00000
2014-17	176	176	100.00000	2013-48	103	103	100.00000
2014-18	172	172	100.00000	2013-49	96	96	100.00000
2014-19	160	160	100.00000	2013-50	117	117	100.00000
2014-20	186	186	100.00000	2013-51	123	123	100.00000
2014-21	177	177	100.00000	2013-52	104	104	100.00000
2014-22	186	186	100.00000	2013-53	41	41	100.00000
2014-23	197	197	100.00000	2014-01	91	91	100.00000
2014-24	198	198	100.00000	2014-02	122	122	100.00000
2014-25	222	222	100.00000	2014-03	112	112	100.00000
2014-26	210	210	100.00000	2014-04	113	113	100.00000
2014-27	199	199	100.00000	2014-05	130	130	100.00000
2014-28	223	223	100.00000	2014-06	132	132	100.00000
2014-29	215	215	100.00000	2014-07	135	135	100.00000
2014-30	228	228	100.00000	2014-08	127	127	100.00000
2014-31	234	234	100.00000	2014-09	127	127	100.00000
2014-32	189	189	100.00000	2014-10	135	135	100.00000
2014-33	250	250	100.00000	2014-11	152	152	100.00000
2014-34	259	259	100.00000	2014-12	132	132	100.00000
2014-35	266	266	100.00000	2014-13	151	151	100.00000

	signup_week	total_signups	retained_users	retention_rate
▶	2013-01	26	26	100.00000
	2013-02	29	29	100.00000
	2013-03	47	47	100.00000
	2013-04	36	36	100.00000
	2013-05	30	30	100.00000
	2013-06	48	48	100.00000
	2013-07	41	41	100.00000
	2013-08	39	39	100.00000
	2013-09	33	33	100.00000
	2013-10	43	43	100.00000
	2013-11	33	33	100.00000
	2013-12	32	32	100.00000
	2013-13	33	33	100.00000
	2013-14	40	40	100.00000
	2013-15	35	35	100.00000
	2013-16	42	42	100.00000
	2013-17	48	48	100.00000
	2013-18	48	48	100.00000
	2013-19	45	45	100.00000
	2013-20	55	55	100.00000
	2013-21	41	41	100.00000
	2013-22	49	49	100.00000
	2013-23	51	51	100.00000
	2013-24	51	51	100.00000

	2013-24	51	51	100.00000
	2013-25	46	46	100.00000
	2013-26	57	57	100.00000
	2013-27	57	57	100.00000
	2013-28	52	52	100.00000
	2013-29	71	71	100.00000
	2013-30	66	66	100.00000
	2013-31	69	69	100.00000
	2013-32	66	66	100.00000
	2013-33	73	73	100.00000
	2013-34	71	71	100.00000
	2013-35	79	79	100.00000
	2013-36	65	65	100.00000
	2013-37	71	71	100.00000
	2013-38	84	84	100.00000
	2013-39	92	92	100.00000
	2013-40	81	81	100.00000
	2013-41	88	88	100.00000
	2013-42	74	74	100.00000
	2013-43	97	97	100.00000
	2013-44	92	92	100.00000
	2013-45	97	97	100.00000

THE ANALYSIS INDICATES THAT THE 35TH WEEK OF 2014 HAD THE HIGHEST NUMBER OF SIGNUPS, WHILE THE 1ST WEEK OF 2013 HAD THE FEWEST SIGNUPS.

FINDING 4.

```
SELECT
    YEARWEEK(occurred_at) AS activity_week,
    device,
    COUNT(DISTINCT user_id) AS active_users
FROM
    events
WHERE
    occurred_at IS NOT NULL
GROUP BY
    activity_week, device
ORDER BY
    active_users desc;
```

	2013-31	69	69	100.00000
	2013-32	66	66	100.00000
	2013-33	73	73	100.00000
	2013-34	71	71	100.00000
	2013-35	79	79	100.00000

DUE TO MORE OUTPUT I HVAVE MENTION FEW

BASED ON THE ANALYSIS, DURING THE 30TH WEEK OF 2014, THE DEVICE "MACBOOK PRO" HAD THE MOST ACTIVE USERS, WITH A TOTAL OF 322 USERS. ON THE OTHER HAND, IN THE 35TH WEEK OF THE SAME YEAR, THE "SAMSUNG GALAXY NOTE" HAD THE FEWEST ACTIVE USERS, WITH ONLY 1 USER RECORDED DURING THAT WEEK.

FINDING 5

```
WITH WeeklyActivity AS (  
  SELECT  
    YEARWEEK(occurred_at, 1) AS activity_week  
    user_id,  
    COUNT(*) AS activity_count  
  FROM  
    events  
  GROUP BY  
    activity_week, user_id  
)  
SELECT  
  activity_week,  
  COUNT(DISTINCT user_id) AS active_users,  
  SUM(activity_count) AS total_activities  
FROM  
  WeeklyActivity  
GROUP BY  
  activity_week  
ORDER BY  
  activity_week desc;
```


201435	1194	16432
201434	1203	16417
201433	1215	16862
201432	1266	18530
201431	1443	21706
201430	1363	20360
201429	1345	20938
201428	1355	20188
201427	1271	19158
201426	1249	19069
201425	1263	19198
201424	1219	18431
201423	1173	18805

DURING THE 35TH WEEK OF 2014, THERE WERE 1,194 ACTIVE USERS AND A TOTAL OF 16,432 ACTIVITIES, WHICH IS THE HIGHEST NUMBER OF ACTIVITIES RECORDED FOR THE ENTIRE PERIOD. DURING THE 35TH WEEK OF 2014, THERE WERE 1,194 ACTIVE USERS AND A TOTAL OF 16,432 ACTIVITIES, WHICH IS THE HIGHEST NUMBER OF ACTIVITIES RECORDED FOR THE ENTIRE PERIOD.

CONCLUSION

WITH ALL THE ANALYSIS I HAVE
LEARN MAY THINGS DURING THIS
DATASET SUCH AS ROLLING TOTAL ,
CTE ETC.

4,HIRING PROCESS ANALYTICS

I AM WORKING WITH HIRING
PROCRSS ANALYTICS DATASET . THIS
TIME I HAVE TO PERFORM ANALYSIS
USING EXCEL.

D. DEPARTMENTAL ANALYSIS: VISUALIZING DATA THROUGH CHARTS AND PLOTS IS A CRUCIAL PART OF DATA ANALYSIS.

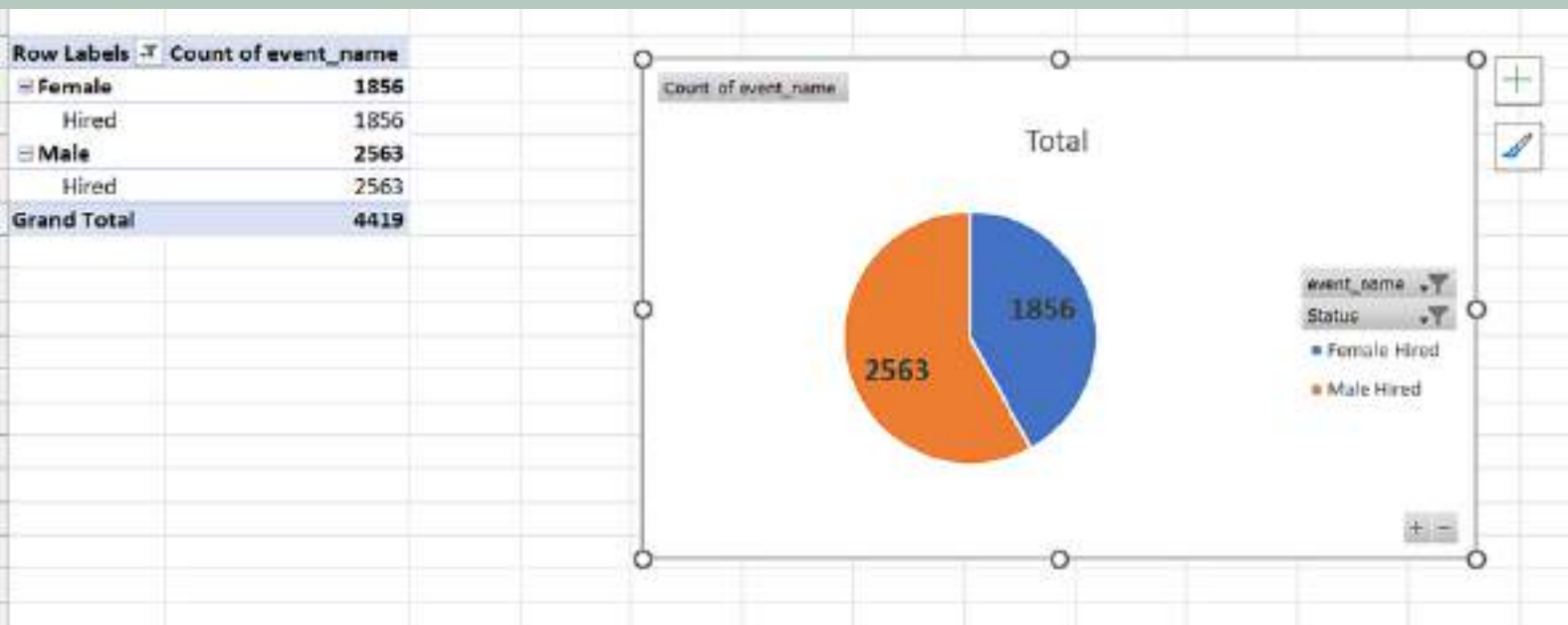
YOUR TASK: USE A PIE CHART, BAR GRAPH, OR ANY OTHER SUITABLE VISUALIZATION TO SHOW THE PROPORTION OF PEOPLE WORKING IN DIFFERENT DEPARTMENTS.

E. POSITION TIER ANALYSIS: DIFFERENT POSITIONS WITHIN A COMPANY OFTEN HAVE DIFFERENT TIERS OR LEVELS.

YOUR TASK: USE A CHART OR GRAPH TO REPRESENT THE DIFFERENT POSITION TIERS WITHIN THE COMPANY. THIS WILL HELP YOU UNDERSTAND THE DISTRIBUTION OF POSITIONS ACROSS DIFFERENT TIERS.

SOLUTION

FINDING 1



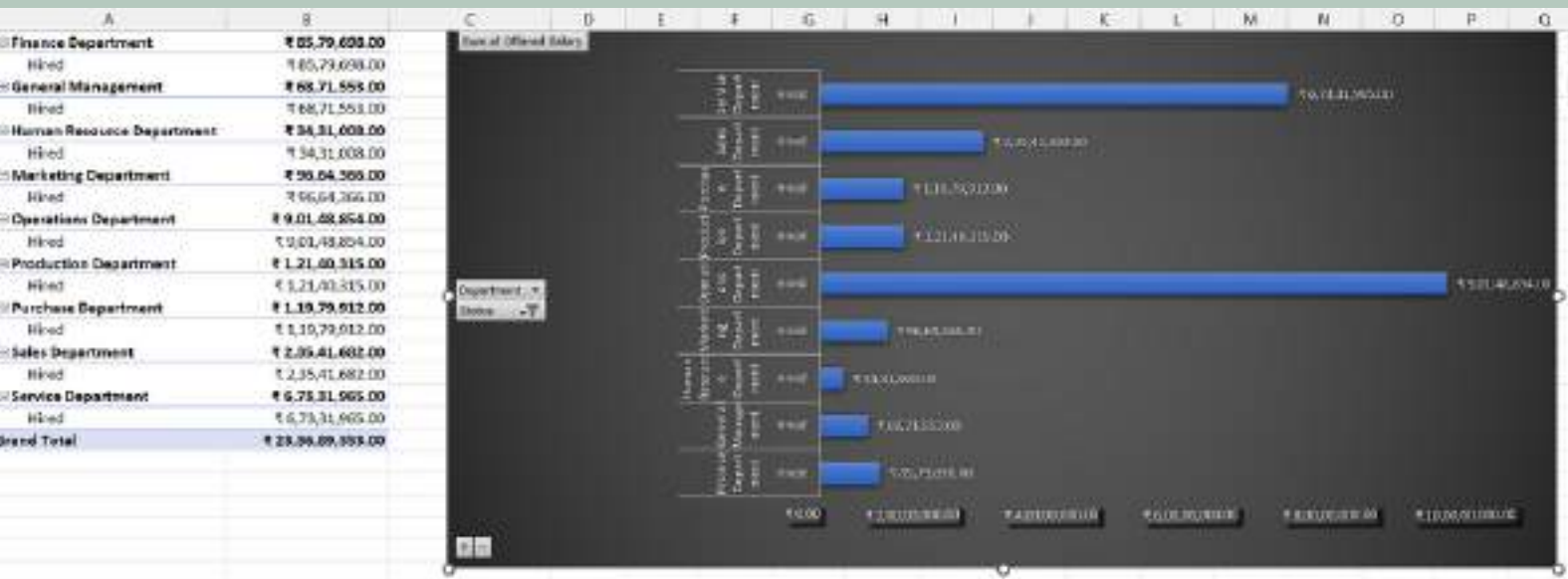
IN THE ABOVE FIGURE IT SHOWS THE MALE AND FEMALE EMPLOYEE HIRED.

FINDING 2

E	F	G	H	I
t	Post Name	Offered Salary		average_salary
artment	c8	₹ 56,553.00		₹ 49,983.03
artment	c5	₹ 22,075.00		

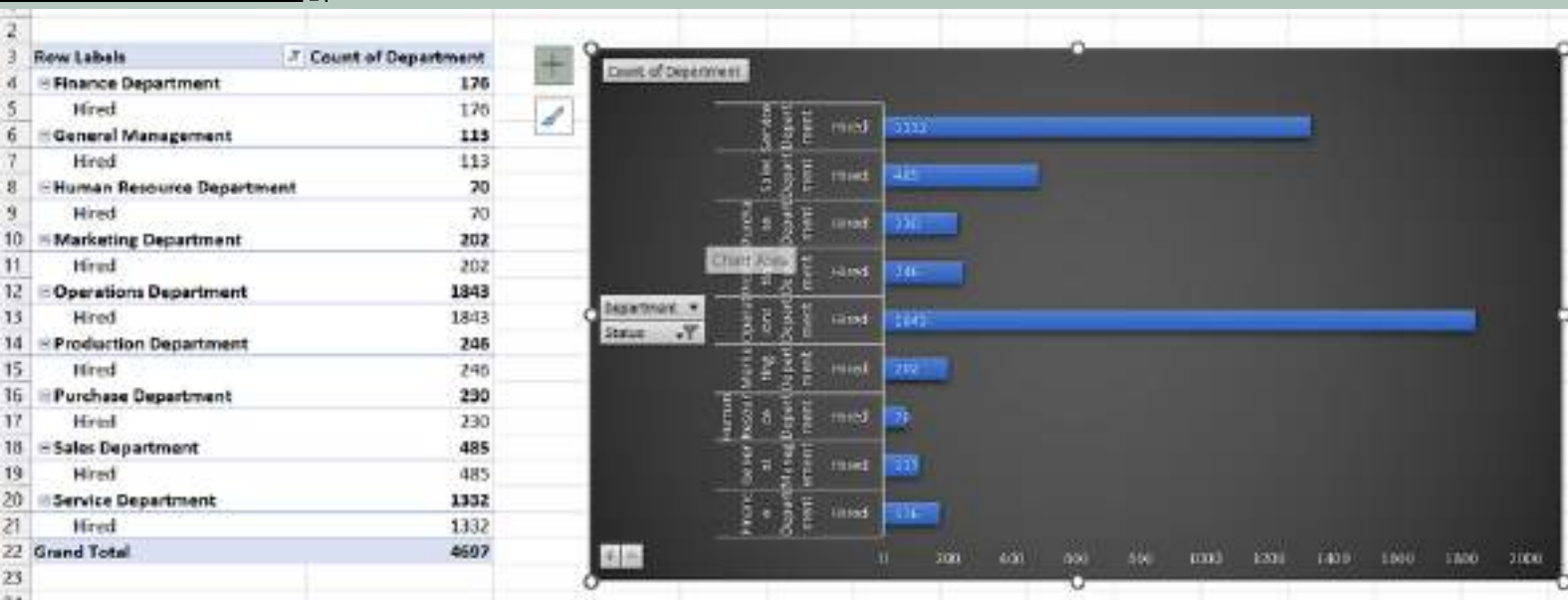
IN THE TASK WE HAVE TO FIND THE AVERAGE SALARY GIVEN BY THE COMPANY.

FINDING 3



SALARY DISTRIBUTION BAR CHART IS PLOTTED ACROSS DIFFRENT DEPARTMENT

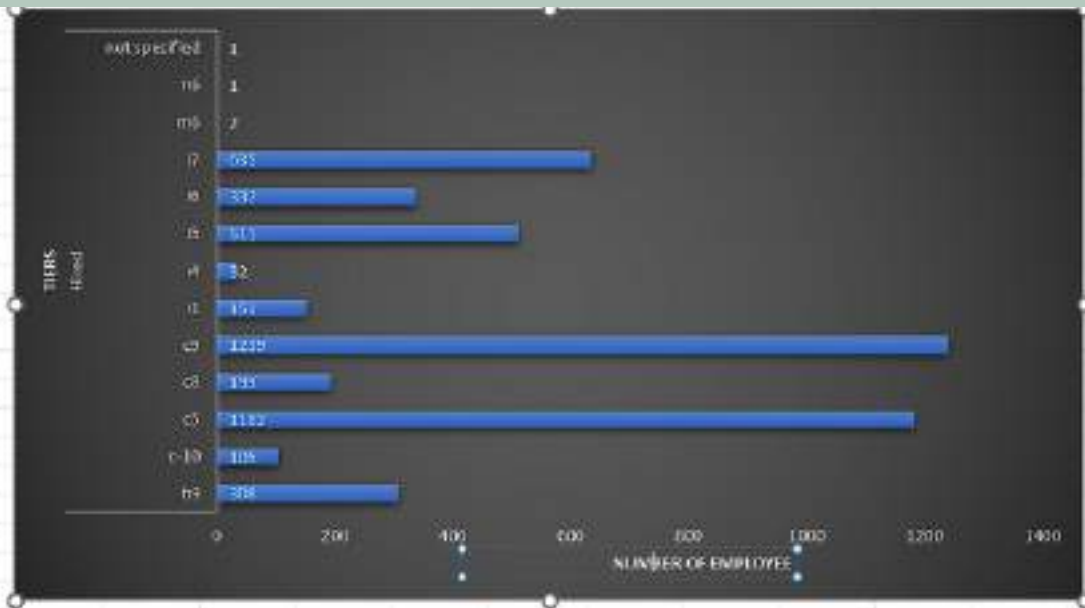
FINDING 4



TOTAL NUMBER OF MEMBERS HIRED AT DIFFRENT DEPARTMENT BAR CHART

FINDING 5

3	Row Labels	Count of Post Name
4	= Hired	4697
5	b9	308
6	c-10	105
7	c5	1182
8	c8	193
9	c9	1239
10	i1	151
11	j4	32
12	s5	511
13	s6	337
14	i7	635
15	m6	2
16	n0	1
17	not specified	1
18	Grand Total	4697
19		
20		
21		
22		
23		
24		
25		



TOTAL NUMBER OF MEMBERS HIRED AT
DIFFRENT DEPARTMENT BRANCH BAR CHART.

CONCLUSION

**AS I HAVE GOT HABDS IN EXPRENCIE
USING EXCEL AND PIVORT CHART
WITH DIFFRENT DIFFERENT TYPE OF
CHARTS.**

5.IMDB MOVIE ANALYSIS

**IN THIS I AM WORKING WITH MOVIES
DATASET TO ANALYSE DIFFERENT
TYPE OF TRENDS .**

STEPS INVOLVED TO MAINTAIN DATA

- **EXTRACT**
- **TRASFORM**
- **LOAD**

TASK TO PERFORM

1.MOVIE GENRE ANALYSIS: ANALYZE THE DISTRIBUTION OF MOVIE GENRES AND THEIR IMPACT ON THE IMDB SCORE.

TASK: DETERMINE THE MOST COMMON GENRES OF MOVIES IN THE DATASET. THEN, FOR EACH GENRE, CALCULATE DESCRIPTIVE STATISTICS (MEAN, MEDIAN, MODE, RANGE, VARIANCE, STANDARD DEVIATION) OF THE IMDB SCORES.

2.MOVIE DURATION ANALYSIS: ANALYZE THE DISTRIBUTION OF MOVIE DURATIONS AND ITS IMPACT ON THE IMDB SCORE.

TASK: ANALYZE THE DISTRIBUTION OF MOVIE DURATIONS AND IDENTIFY THE RELATIONSHIP BETWEEN MOVIE DURATION AND IMDB SCORE.

3.LANGUAGE ANALYSIS: SITUATION: EXAMINE THE DISTRIBUTION OF MOVIES BASED ON THEIR LANGUAGE.

TASK: DETERMINE THE MOST COMMON LANGUAGES USED IN MOVIES AND ANALYZE THEIR IMPACT ON THE IMDB SCORE USING DESCRIPTIVE STATISTICS.

4. DIRECTOR ANALYSIS: INFLUENCE OF DIRECTORS ON MOVIE RATINGS.

TASK: IDENTIFY THE TOP DIRECTORS BASED ON THEIR AVERAGE IMDB SCORE AND ANALYZE THEIR CONTRIBUTION TO THE SUCCESS OF MOVIES USING PERCENTILE CALCULATIONS.

5.BUDGET ANALYSIS: EXPLORE THE RELATIONSHIP BETWEEN MOVIE BUDGETS AND THEIR FINANCIAL SUCCESS.

TASK: ANALYZE THE CORRELATION BETWEEN MOVIE BUDGETS AND GROSS EARNINGS, AND IDENTIFY THE MOVIES WITH THE HIGHEST PROFIT MARGIN.

SOLUTION

FINDING 1

Drama	236
Comedy	208
Comedy Drama	189
Comedy Drama Romance	186
Comedy Romance	158
Drama Romance	152
Crime Drama Thriller	101
Horror	71
Action Crime Drama Thriller	68
Action Crime Thriller	65

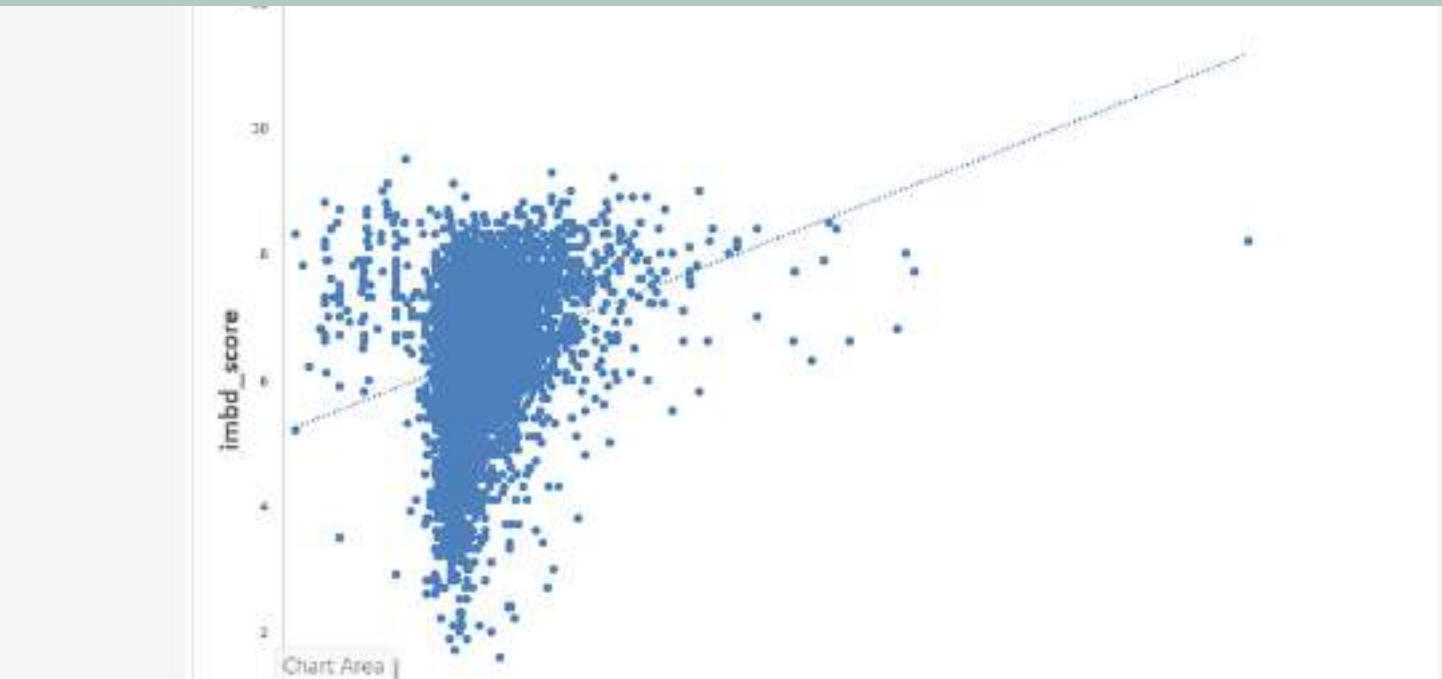
THIS ARE THE TOP 10 MOST LOVED MOVIE BY THE PEOPLE
IN WHICH DRAMA IS LEADING AMONG ALL

	mean	6.44159	
	median	6.6	
	mode	6.7	
	range	7.9	
	varience	1.2675	
	Standard Deviation	1.125833	

BEACAUSE OF LARGE DATASET I HAVE TAKEN
ALL GENRES IMBD MEAN, MEDIAN, MODE,
RANGE, VARIENCE, STANDARD DEVIATION

FINDING 2

	duration		
	mean	107.2268	
	median	103	
	mode	90	
	range	504	
	variance	632.1356	
	Standard Deviation	25.14231	



AS WE CAN SEE THAT THE LESS THE DURATION (BETWEEN 100-200MIN) IS THE MORE THE IMBD SCORE .

FINDING 3

4	English	4703
5	French	73
5	Spanish	40
7	Hindi	28
8	Mandarin	26

**TOP 5 MOST COMMON
LANGUAGES.**

1	language	mean	meadium	Standard Deviation
2	English	6.398873	6.5	1.12324006
3	French	7.038356	7.2	0.726985812
4	Spanish	6.9375	7.15	0.855056603
5	Hindi	6.632143	6.95	1.398955582
5	Mandarin	6.788462	7.05	1.042046802
7	German	7.342105	7.6	0.954123093
8	Japanese	7.394444	7.6	0.990823913
9	Russian	6.363636	6.5	1.383671007
10	Italian	7.227273	7.3	1.244259546
11	Cantonese	6.954545	7.2	0.704788814
12	Portuguese	7.4875	7.7	0.883883476
13	Korean	7.3875	7.5	0.825378701
14	Swedish	7.44	7.6	0.756967635
15	Hebrew	7.58	7.6	0.334664011
16	Danish	7.5	8.1	1.077032961
17	Arabic	7.38	7.4	0.884307639
18	Persian	7.575	7.95	1.203813385
19	Dutch	7.425	7.45	0.434932945
20	Norwegian	7.15	7.3	0.574456265
21	Polish	8.25	8.25	0.981495458
22	Chinese	5.666667	5.7	0.550757055
23	Thai	6.633333	6.6	0.450924975
24	Dari	7.5	7.5	0.141421356

**TOP 25 LANGUAGES WITH
MEAN , MEADIUM , SD.**

FINDING 4.

directors	average movie_imd	percent rank	Column1
John Blanchard	9.5	1	Top Director
Cary Bell	8.7	0.998	Other
Mitchell Altieri	8.7	0.998	Other
Sadyk Sher-Niyaz	8.7	0.998	Other
Charles Chaplin	8.6	0.997	Other
Mike Mayhall	8.6	0.997	Other
Damien Chazelle	8.5	0.996	Other
Majid Majidi	8.5	0.996	Other
Raja Menon	8.5	0.996	Other
Ron Fricke	8.5	0.996	Other
Christopher Nolan	8.425	0.995	Other
Sergio Leone	8.475	0.995	Other
Asghar Farhadi	8.4	0.991	Other
Bill Melendez	8.4	0.991	Other
Catherine Owens	8.4	0.991	Other
Jay Oliva	8.4	0.991	Other
Marius A. Markevicius	8.4	0.991	Other
Moustapha Akkad	8.4	0.991	Other
Rakeysh Omprakash Me	8.4	0.991	Other
Richard Marquand	8.4	0.991	Other
Robert Mulligan	8.4	0.991	Other
S.S. Rajamouli	8.4	0.991	Other
Fritz Lang	8.3	0.988	Other

THE RESULT IS OF TOP 24
BEST DIRECTOR AND
AMONG THEM JOHN
BLANCHARD IS AT TOP

FINDING 5

genres	budget	gross	movie_title	Profit Margin
Crime Drama	3250	48468408	The Mongol King	14012.3561
Horror	15000	107917283	Paranormal Activity	7193.48553
Horror	10000	48468408	Tin Can Man	4845.84075
Romance Short	13000	48468408	The Touch	3727.33904
Drama	15000	48468408	Stories of Our Lives	3230.22164
Drama Horror Thriller	17350	48468408	The Ridges	2792.56815
Biography Documentary	218	592014	Tamaton	2714.6606
Drama	20000	48468408	Hywheel	2422.42037
Fantasy Horror Sci-Fi	20000	48468408	Eraserhead	2422.42037
Family	20000	48468408	Dude, Where's My Dog?	2422.42037
Sci-Fi Thriller	20000	48468408	All Superheroes Must Die	2422.42037
Horror	60000	140530114	The Blair Witch Project	2341.16856
Comedy Romance	22000	48468408	Dry Spell	2202.10943
Drama Family	24000	48468408	Manito	2018.5169
Crime Drama Thriller	25000	48468408	Dutch Kills	1937.73630
Horror	40000	48468408	Raymond Did It	1210.71018
Drama	42000	48468408	This Is Martin Bonner	1153.00970
Comedy	45000	48468408	A True Story	1076.07572
Documentary	50000	48468408	Counting	968.368150
Documentary	50000	48468408	Bending Steel	968.368150
Biography Documentary	50000	48468408	The Image Revolution	968.368150
Horror Thriller	50000	48468408	Run, Hide, Die	968.368150
Fantasy Horror Mystery Thriller	50000	48468408	The Call of Cthulhu	968.368150
Crime Drama Horror	60000	48468408	Hayride	806.805792

Correlation

0.100967

THIS THE SOME MOST PROFITED GENRES.

CONCLUSION

THE PROJECT PROVIDED VALUABLE INSIGHTS INTO THE FACTORS THAT INFLUENCE MOVIE SUCCESS

ON IMDB. KEY TAKEAWAYS INCLUDE:

- BUDGETS AND EARNINGS
- GENRES
- DIRECTOR IMPACT
- LANGUAGE AND DURATION
- PROFITABILITY

6. BANK LOAN CASE STUDY

IN THIS DATASET WE ARE WORKING WITH BANK LOAN TO IDENTIFY WHY THEIR IS HUGE AMOUNT OF REJECTION IN THE BANK LOAN APPLICATION.

STEPS INVOLVED TO MAINTAIN DATA

- **EXTRACT**
- **TRASFORM**
- **LOAD**

TASK TO PERFORM

1.IDENTIFY MISSING DATA AND DEAL WITH IT APPROPRIATELY: AS A DATA ANALYST, YOU COME ACROSS MISSING DATA IN THE LOAN APPLICATION DATASET. IT IS ESSENTIAL TO HANDLE MISSING DATA EFFECTIVELY TO ENSURE THE ACCURACY OF THE ANALYSIS.

TASK: IDENTIFY THE MISSING DATA IN THE DATASET AND DECIDE ON AN APPROPRIATE METHOD TO DEAL WITH IT USING EXCEL BUILT-IN FUNCTIONS AND FEATURES.

2.IDENTIFY OUTLIERS IN THE DATASET: OUTLIERS CAN SIGNIFICANTLY IMPACT THE ANALYSIS AND DISTORT THE RESULTS. YOU NEED TO IDENTIFY OUTLIERS IN THE LOAN APPLICATION DATASET.

TASK: DETECT AND IDENTIFY OUTLIERS IN THE DATASET USING EXCEL STATISTICAL FUNCTIONS AND FEATURES, FOCUSING ON NUMERICAL VARIABLES.

3.ANALYZE DATA IMBALANCE: DATA IMBALANCE CAN AFFECT THE ACCURACY OF THE ANALYSIS, ESPECIALLY FOR BINARY CLASSIFICATION PROBLEMS. UNDERSTANDING THE DATA DISTRIBUTION IS CRUCIAL FOR BUILDING RELIABLE MODELS.

TASK: DETERMINE IF THERE IS DATA IMBALANCE IN THE LOAN APPLICATION DATASET AND CALCULATE THE RATIO OF DATA IMBALANCE USING EXCEL FUNCTIONS.

4.PERFORM UNIVARIATE, SEGMENTED UNIVARIATE, AND BIVARIATE ANALYSIS: TO GAIN INSIGHTS INTO THE DRIVING FACTORS OF LOAN DEFAULT, IT IS IMPORTANT TO CONDUCT VARIOUS ANALYSES ON CONSUMER AND LOAN ATTRIBUTES.

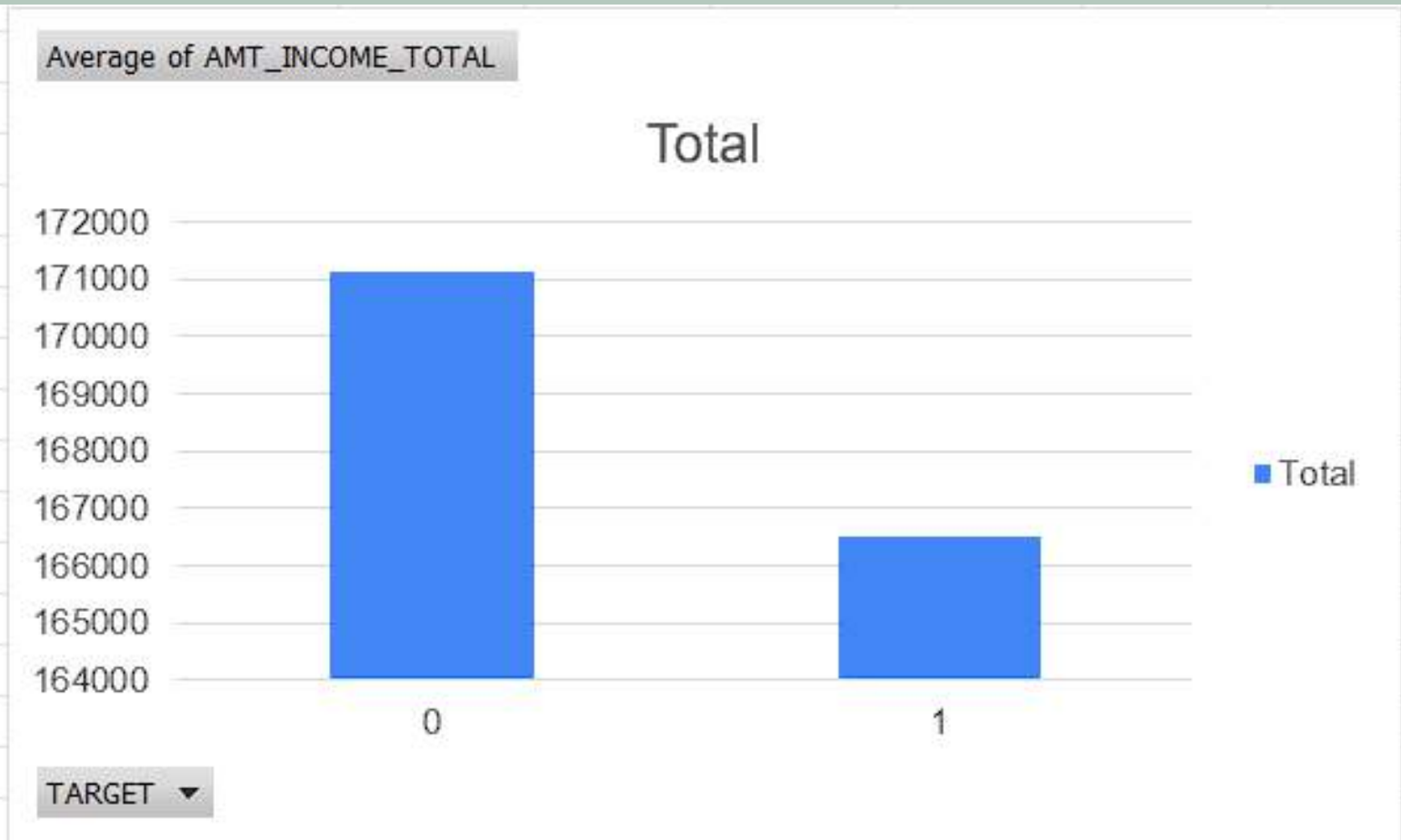
TASK: PERFORM UNIVARIATE ANALYSIS TO UNDERSTAND THE DISTRIBUTION OF INDIVIDUAL VARIABLES, SEGMENTED UNIVARIATE ANALYSIS TO COMPARE VARIABLE DISTRIBUTIONS FOR DIFFERENT SCENARIOS, AND BIVARIATE ANALYSIS TO EXPLORE RELATIONSHIPS BETWEEN VARIABLES AND THE TARGET VARIABLE USING EXCEL FUNCTIONS AND FEATURES.

5.IDENTIFY TOP CORRELATIONS FOR DIFFERENT SCENARIOS: UNDERSTANDING THE CORRELATION BETWEEN VARIABLES AND THE TARGET VARIABLE CAN PROVIDE INSIGHTS INTO STRONG INDICATORS OF LOAN DEFAULT.

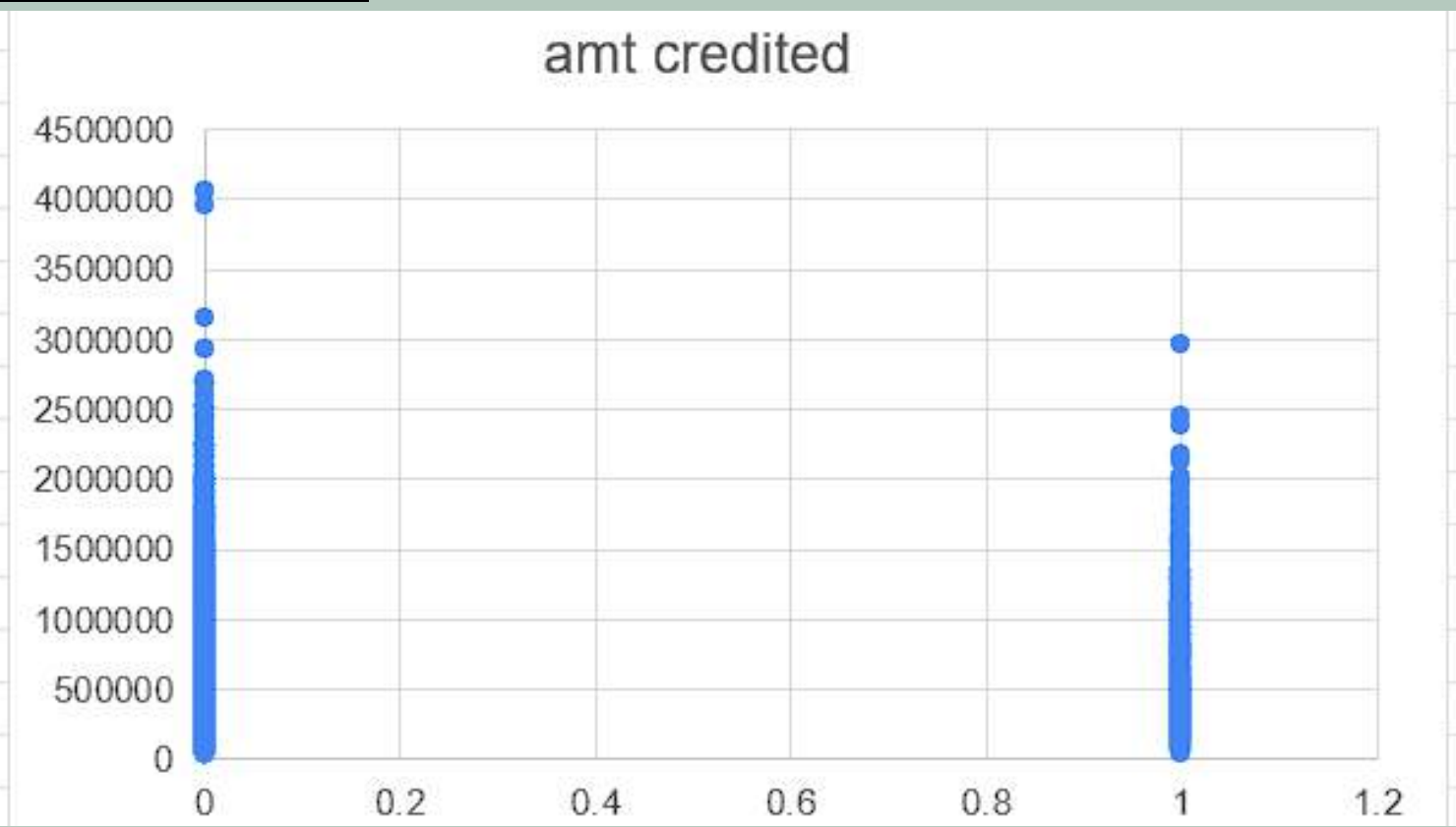
TASK: SEGMENT THE DATASET BASED ON DIFFERENT SCENARIOS (E.G., CLIENTS WITH PAYMENT DIFFICULTIES AND ALL OTHER CASES) AND IDENTIFY THE TOP CORRELATIONS FOR EACH SEGMENTED DATA USING EXCEL FUNCTIONS.

SOLUTION

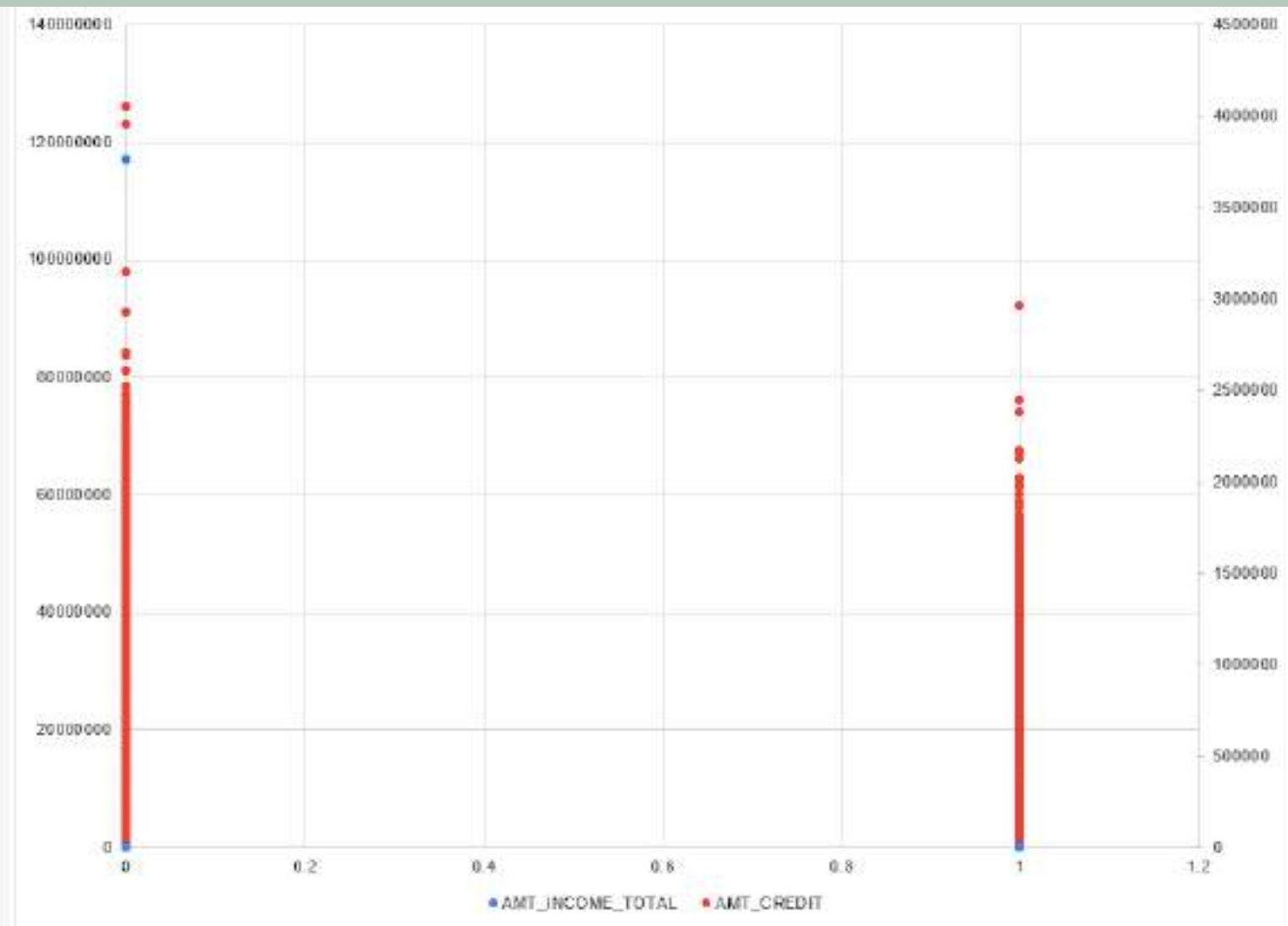
FINDING 1



FINDING 2

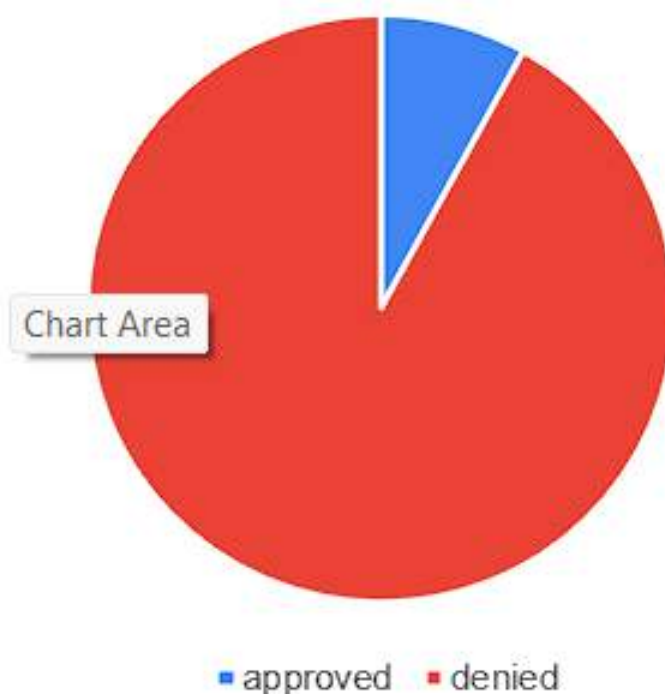


FINDING 3

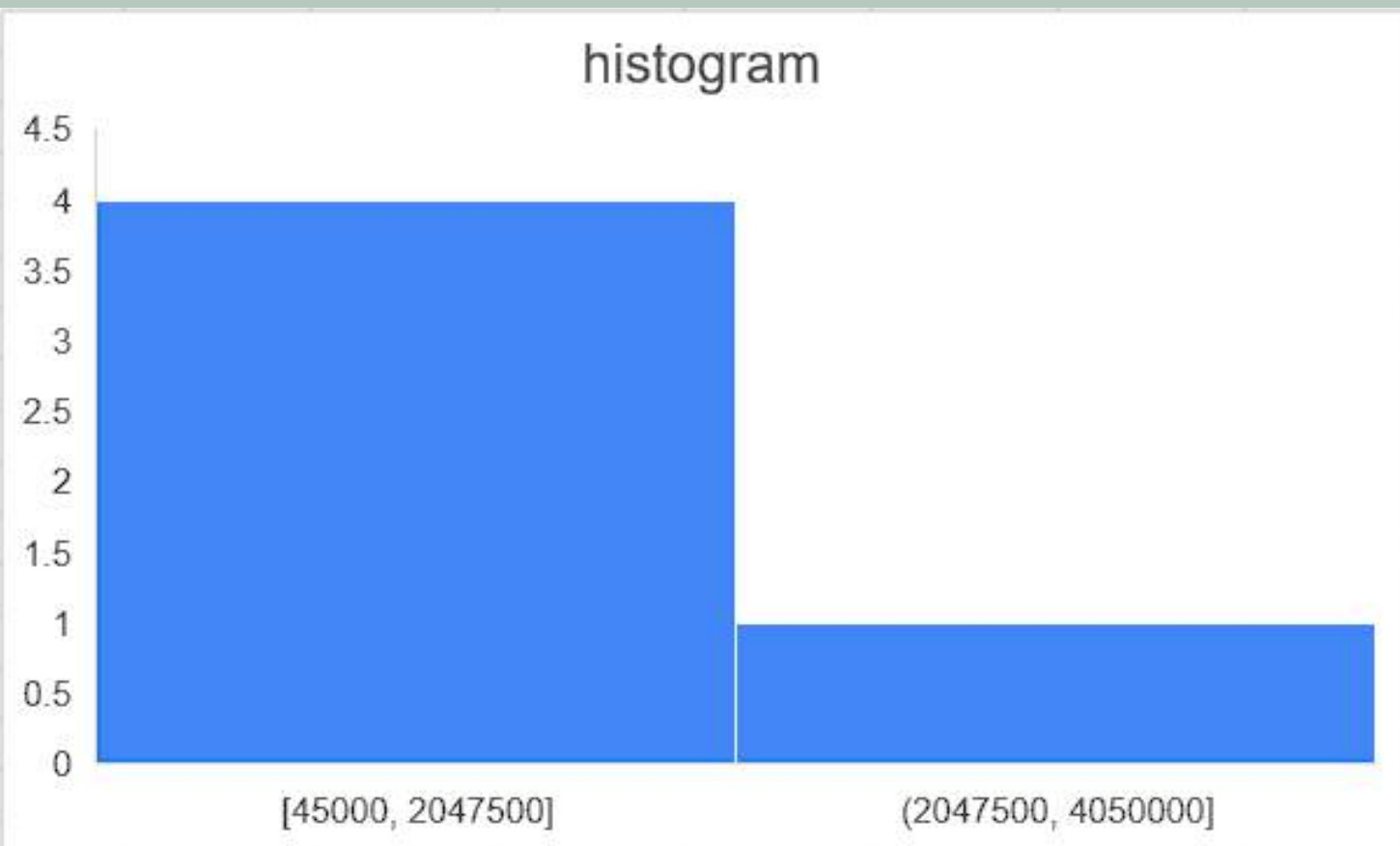


FINDING 4

data imbalance



FINDING 5



CONCLUSION

- AS I HAVE SEEN THAT THE MOST OF THE LOAN DATA HAVE BEEN REJECTED BY THE BANK .
- DATA HAVE BEEN IMBALANCE IN HUGE AMOUNT

7.ANALYZING THE IMPACT OF CAR FEATURES ON PRICE AND PROFITABILITY

STEPS INVOLVED TO MAINTAIN DATA

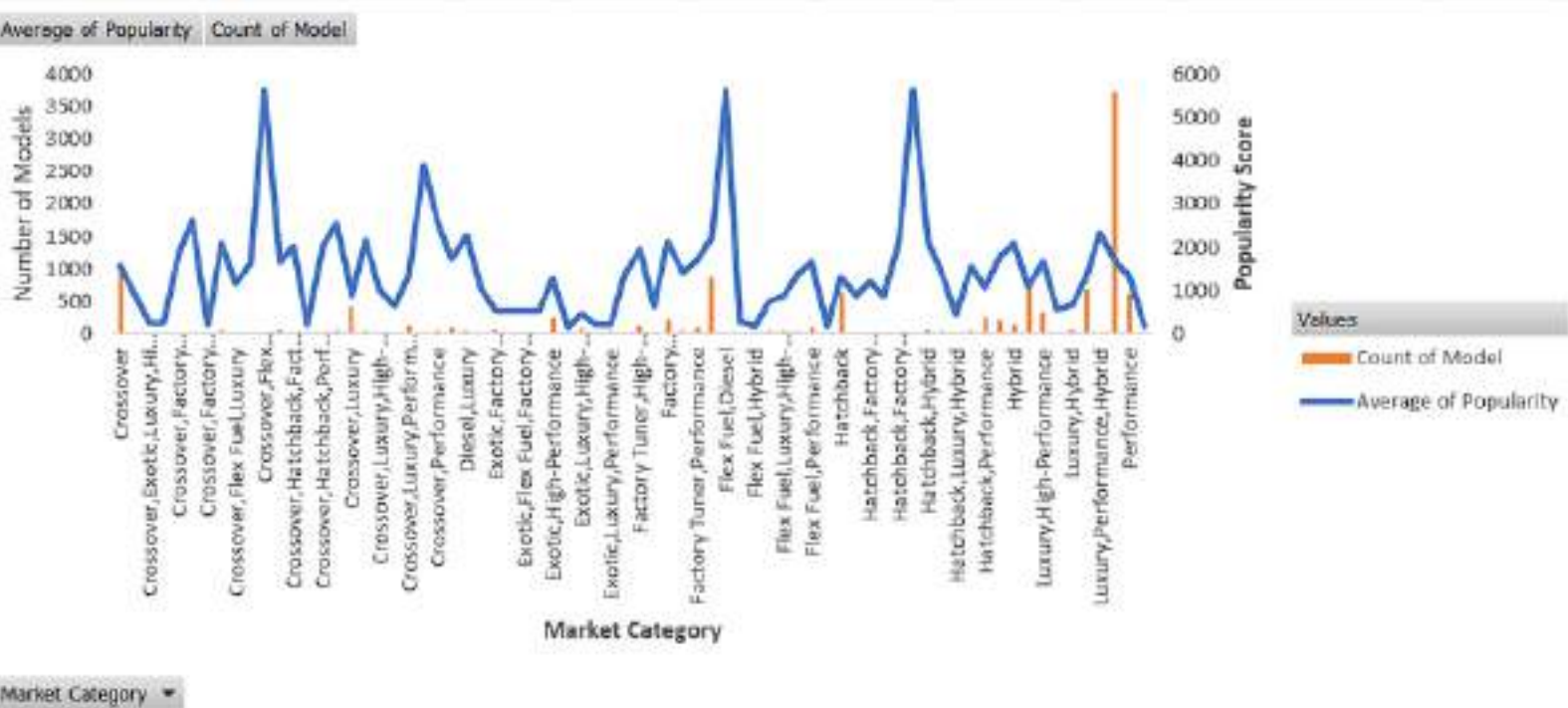
- EXTRACT
- TRANSFORM
- LOAD

TASK TO PERFORM

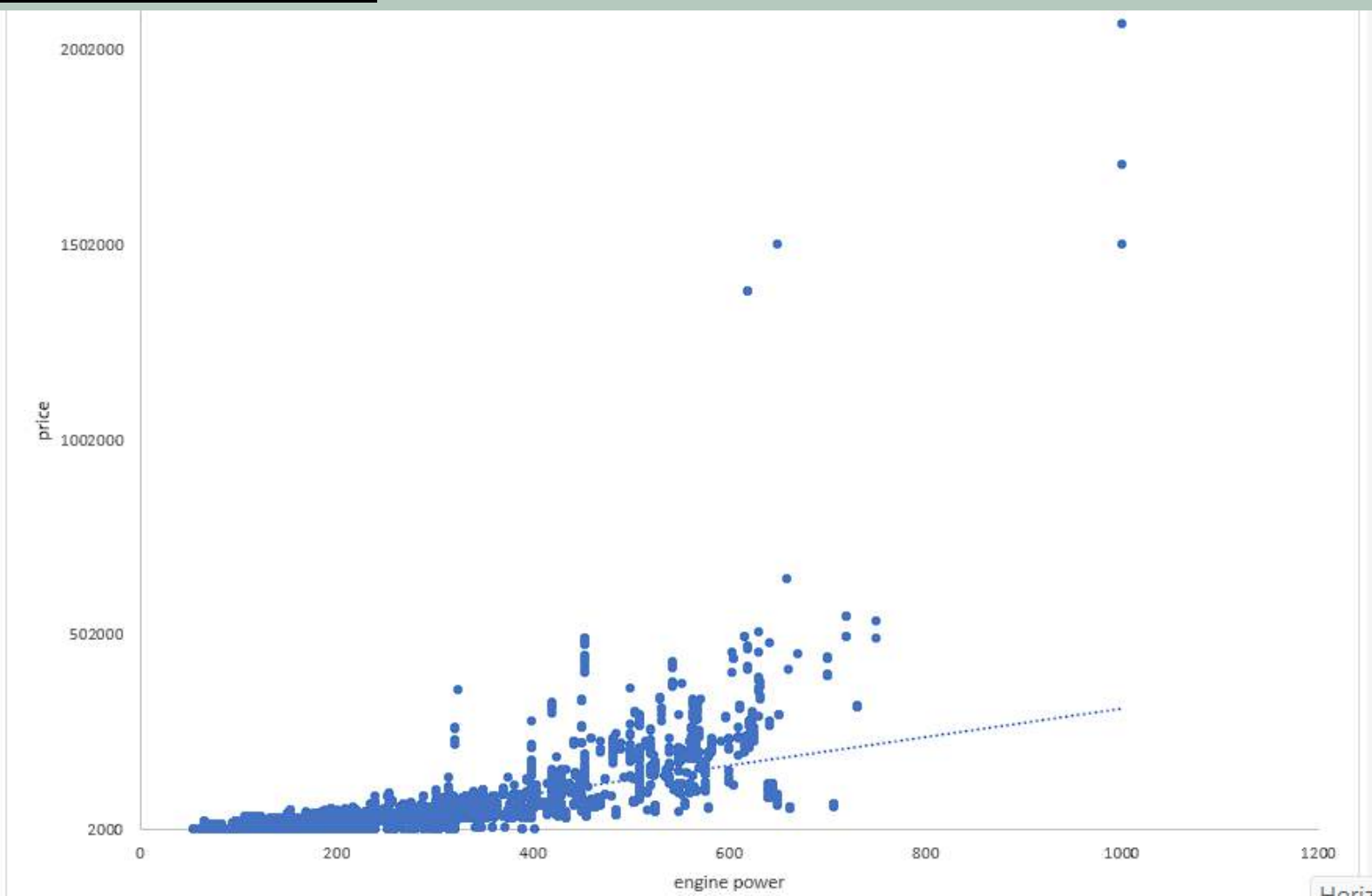
1. HOW DOES THE POPULARITY OF A CAR MODEL VARY ACROSS DIFFERENT MARKET CATEGORIES?
 - **TASK 1.A:** CREATE A PIVOT TABLE THAT SHOWS THE NUMBER OF CAR MODELS IN EACH MARKET CATEGORY AND THEIR CORRESPONDING POPULARITY SCORES.
 - **TASK 1.B:** CREATE A COMBO CHART THAT VISUALIZES THE RELATIONSHIP BETWEEN MARKET CATEGORY AND POPULARITY.
2. WHAT IS THE RELATIONSHIP BETWEEN A CAR'S ENGINE POWER AND ITS PRICE?
 - **TASK 2:** CREATE A SCATTER CHART THAT PLOTS ENGINE POWER ON THE X-AXIS AND PRICE ON THE Y-AXIS. ADD A TRENDLINE TO THE CHART TO VISUALIZE THE RELATIONSHIP BETWEEN THESE VARIABLES.
3. WHICH CAR FEATURES ARE MOST IMPORTANT IN DETERMINING A CAR'S PRICE?
 - **TASK 3:** USE REGRESSION ANALYSIS TO IDENTIFY THE VARIABLES THAT HAVE THE STRONGEST RELATIONSHIP WITH A CAR'S PRICE. THEN CREATE A BAR CHART THAT SHOWS THE COEFFICIENT VALUES FOR EACH VARIABLE TO VISUALIZE THEIR RELATIVE IMPORTANCE.
4. HOW DOES THE AVERAGE PRICE OF A CAR VARY ACROSS DIFFERENT MANUFACTURERS?
 - **TASK 4.A:** CREATE A PIVOT TABLE THAT SHOWS THE AVERAGE PRICE OF CARS FOR EACH MANUFACTURER.
 - **TASK 4.B:** CREATE A BAR CHART OR A HORIZONTAL STACKED BAR CHART THAT VISUALIZES THE RELATIONSHIP BETWEEN MANUFACTURER AND AVERAGE PRICE.
5. WHAT IS THE RELATIONSHIP BETWEEN FUEL EFFICIENCY AND THE NUMBER OF CYLINDERS IN A CAR'S ENGINE?
 - **TASK 5.A:** CREATE A SCATTER PLOT WITH THE NUMBER OF CYLINDERS ON THE X-AXIS AND HIGHWAY MPG ON THE Y-AXIS. THEN CREATE A TRENDLINE ON THE SCATTER PLOT TO VISUALLY ESTIMATE THE SLOPE OF THE RELATIONSHIP AND ASSESS ITS SIGNIFICANCE.
 - **TASK 5.B:** CALCULATE THE CORRELATION COEFFICIENT BETWEEN THE NUMBER OF CYLINDERS AND HIGHWAY MPG TO QUANTIFY THE STRENGTH AND DIRECTION OF THE RELATIONSHIP.

SOLUTION

FINDING 1



FINDING 2



FINDING 3

Prediction

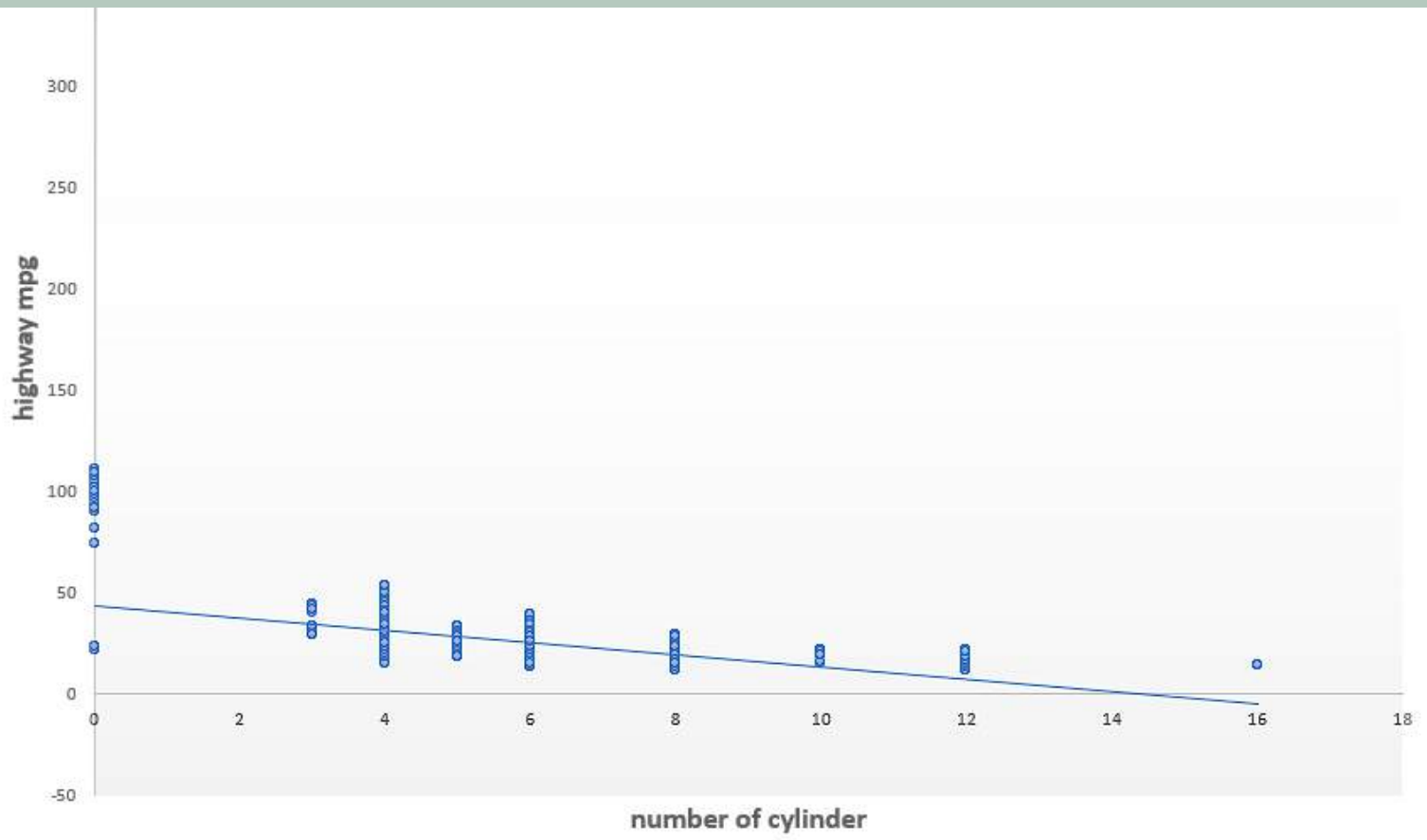


FINDING 4

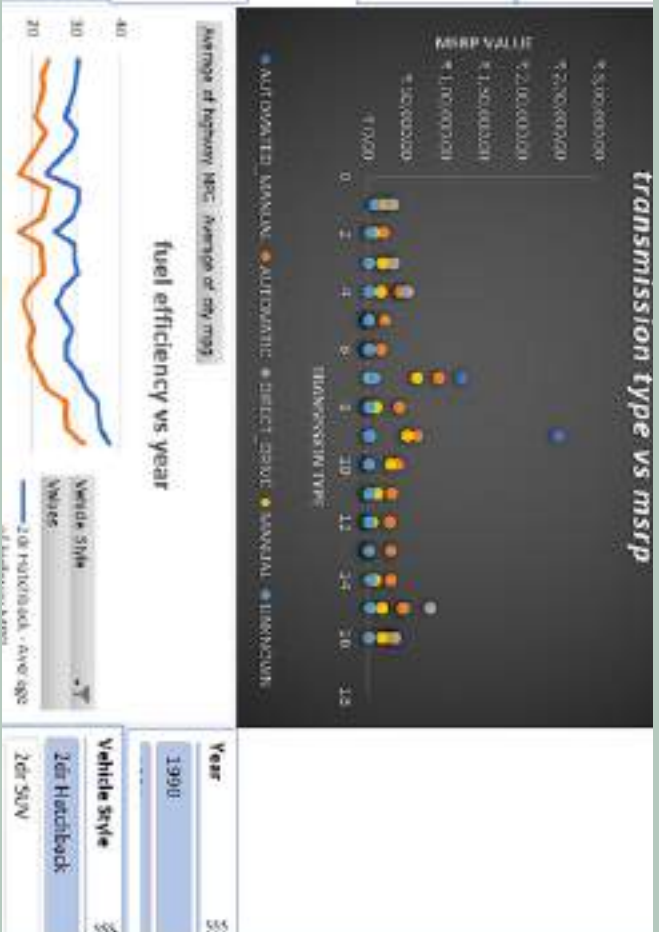
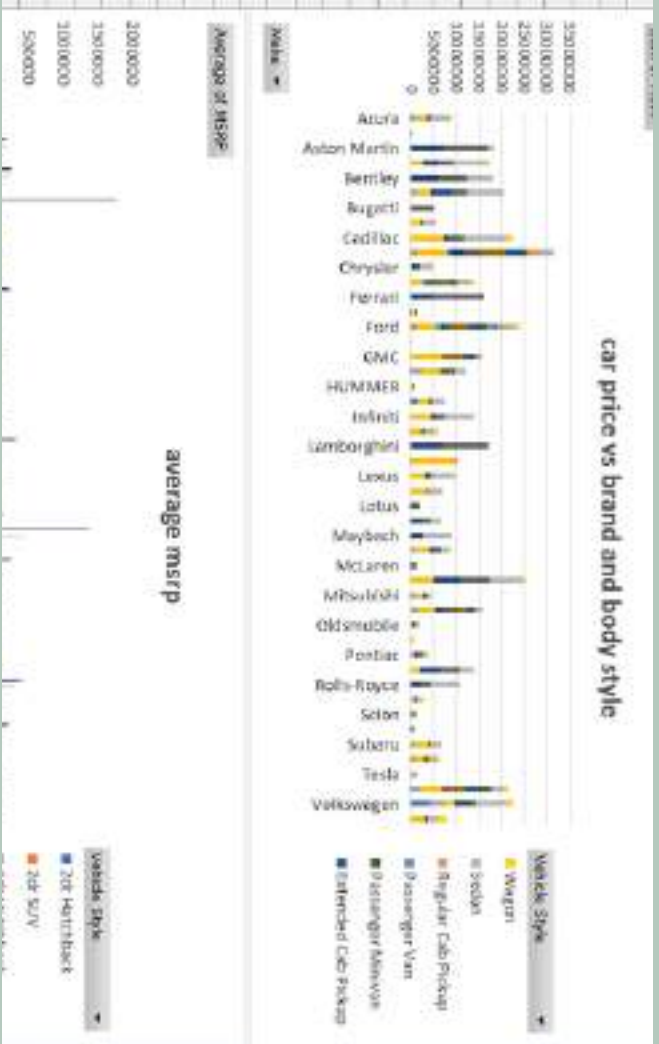
model vs avg price

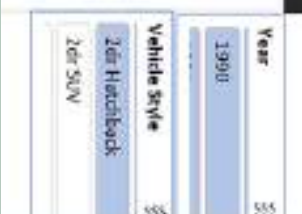
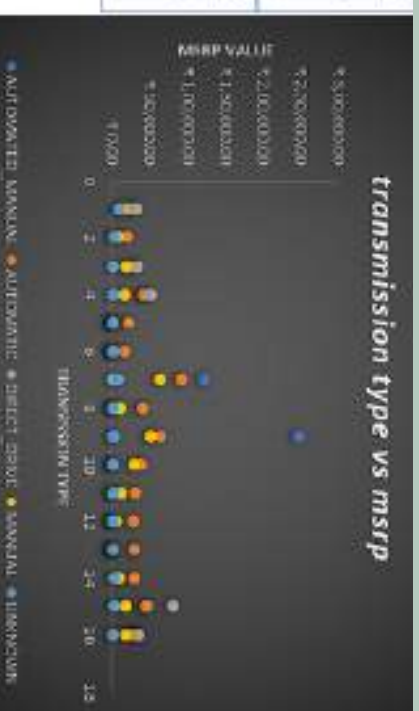
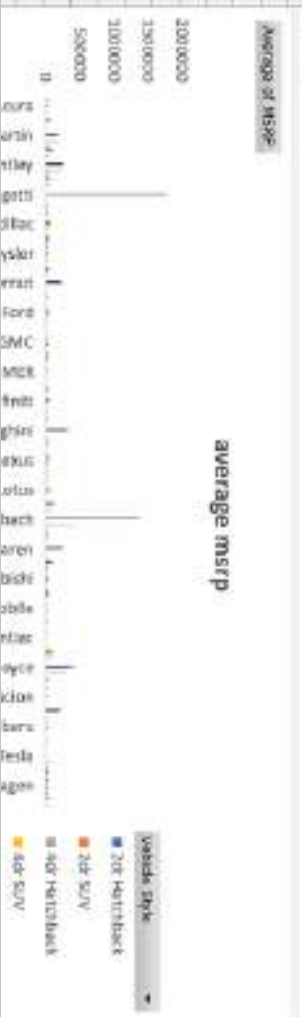
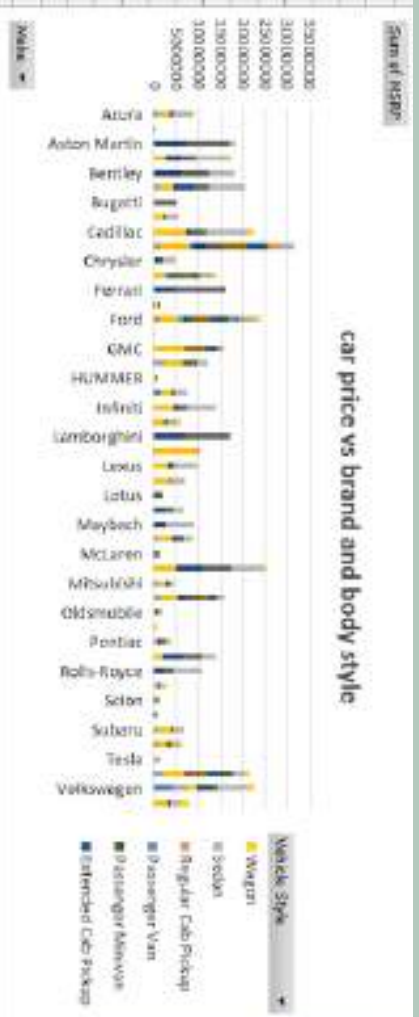


FINDING 5



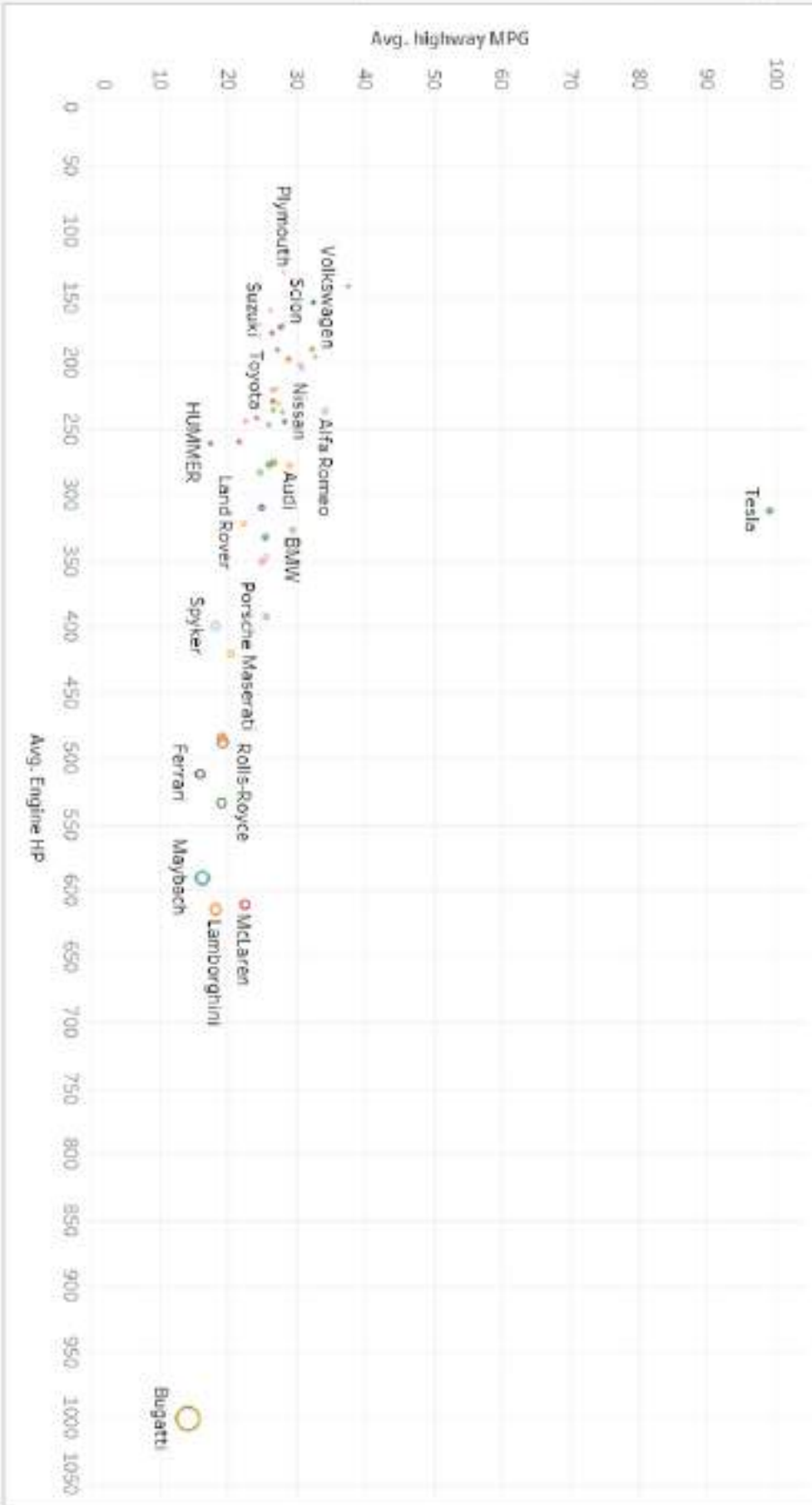
FINDING 5





Columns: Avg(Engine HP)
Avg(highway MPG)

Sheet 1



Avg(Merp)
3,122
500,000
1,000,000
1,500,000
1,757,225

Make
Lamborghini
Land Rover
Lexus
Lincoln
Lotus
Maserati
Maybach
Mazda
McLaren
Mercedes
Mitsubishi
Nissan
Oldsmobile
Plymouth
Pontiac
Porsche
Rolls-Royce
Saab
Scion
Spyker
Subaru
Suzuki
Tesla
Toyota
Volkswagen

Pages

Columns

AVGCity Mpg

Entire View

11)

Rows

AVGCity Mpg

Filters

Sheet1

Make

Circle

Color

Size

Label

Detail

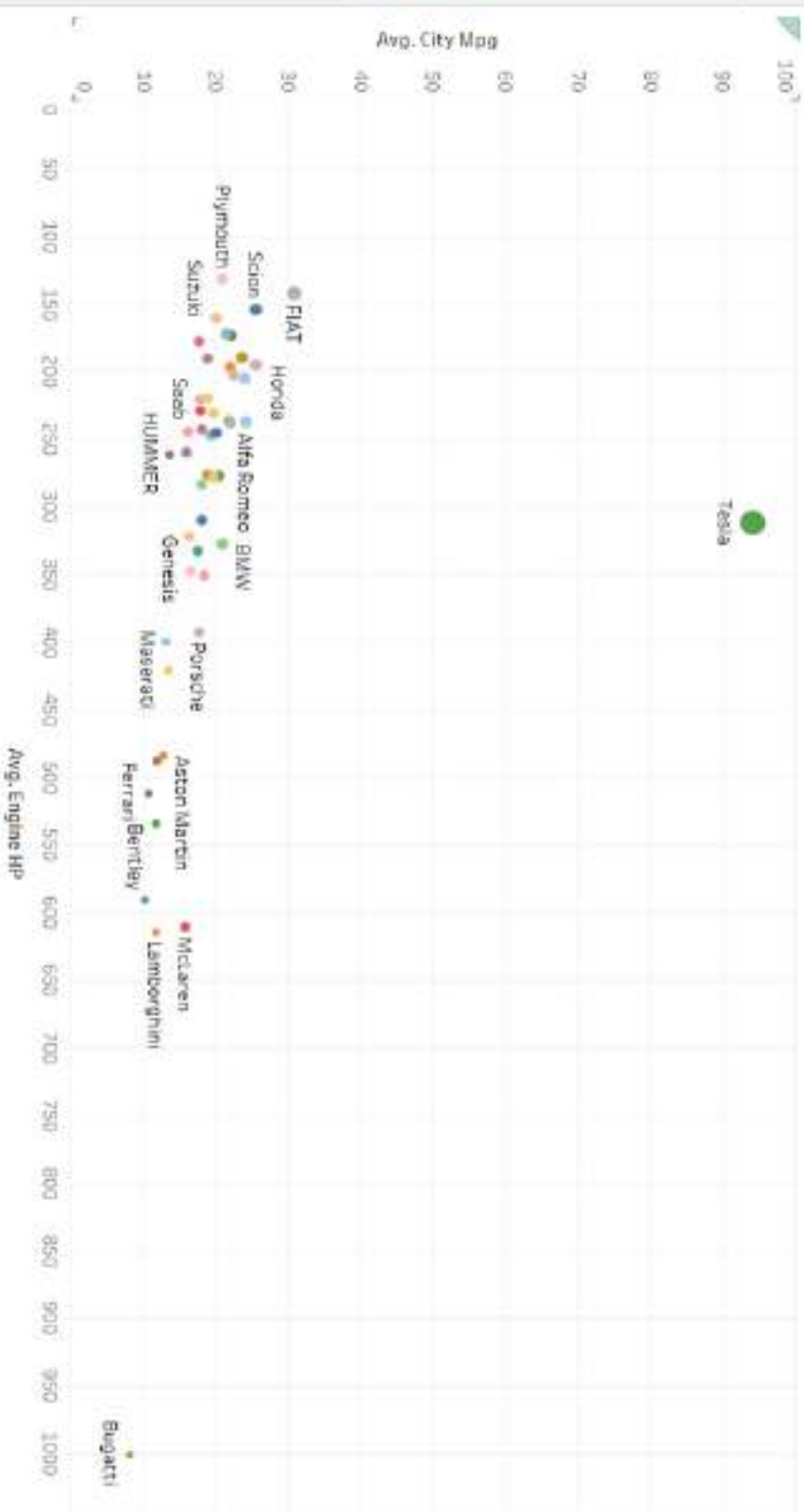
Tooltip

Make

AVGCity Mpg

ATTN(Make)

Avg. City Mpg



INSIGHTS

PRICING (MSRP) INSIGHTS: VEHICLES WITH HIGHER HORSEPOWER AND LARGER ENGINE SIZES GENERALLY HAVE A HIGHER MSRP.

TRANSMISSION TYPE AND BODY STYLE : MANUAL TRANSMISSION CARS TEND TO HAVE LOWER MSRPS, OFTEN ASSOCIATED WITH COMPACT OR SPORTS MODELS.

MARKET CATEGORY TRENDS: LUXURY AND PERFORMANCE CATEGORIES EXHIBIT THE HIGHEST MSRPS, WHILE ECONOMY AND COMPACT CARS TARGET AFFORDABILITY.

CONCLUSION

TABLES AND PIVOT ANALYSIS: PIVOT TABLE: DISPLAYED THE AVERAGE MSRP, POPULARITY SCORES, AND MPG GROUPED BY TRANSMISSION TYPE AND BODY STYLE.

GRAPHS AND CHARTS: SCATTER PLOT: SHOWED THE RELATIONSHIP BETWEEN HORSEPOWER AND MSRP, WITH DIFFERENT MARKERS FOR BODY STYLES AND TRANSMISSION TYPES.

BAR CHART: COMPARED THE NUMBER OF CAR MODELS IN EACH MARKET CATEGORY AND THEIR AVERAGE POPULARITY SCORES.

DASHBOARD: CREATED AN INTERACTIVE DASHBOARD IN EXCEL TO DYNAMICALLY FILTER AND VIEW INSIGHTS BASED ON BODY STYLE, MARKET CATEGORY, OR TRANSMISSION TYPE.

8.ABC CALL VOLUME TREND ANALYSIS

**IN THIS DATASET WE
ARE GOING TO WORK
WITH ABC CALL
VOLUME DATA TO
FIND OUT VALUABLE
INSIGHTS**

TASK TO PERFORM

1. **AVERAGE CALL DURATION:** DETERMINE THE AVERAGE DURATION OF ALL INCOMING CALLS RECEIVED BY AGENTS. THIS SHOULD BE CALCULATED FOR EACH TIME BUCKET.

YOUR TASK: WHAT IS THE AVERAGE DURATION OF CALLS FOR EACH TIME BUCKET?

2. **CALL VOLUME ANALYSIS:** VISUALIZE THE TOTAL NUMBER OF CALLS RECEIVED. THIS SHOULD BE REPRESENTED AS A GRAPH OR CHART SHOWING THE NUMBER OF CALLS AGAINST TIME. TIME SHOULD BE REPRESENTED IN BUCKETS (E.G., 1-2, 2-3, ETC.).

YOUR TASK: CAN YOU CREATE A CHART OR GRAPH THAT SHOWS THE NUMBER OF CALLS RECEIVED IN EACH TIME BUCKET?

3. **MANPOWER PLANNING:** THE CURRENT RATE OF ABANDONED CALLS IS APPROXIMATELY 30%. PROPOSE A PLAN FOR MANPOWER ALLOCATION DURING EACH TIME BUCKET (FROM 9 AM TO 9 PM) TO REDUCE THE ABANDON RATE TO 10%. IN OTHER WORDS, YOU NEED TO CALCULATE THE MINIMUM NUMBER OF AGENTS REQUIRED IN EACH TIME BUCKET TO ENSURE THAT AT LEAST 90 OUT OF 100 CALLS ARE ANSWERED.

YOUR TASK: WHAT IS THE MINIMUM NUMBER OF AGENTS REQUIRED IN EACH TIME BUCKET TO REDUCE THE ABANDON RATE TO 10%?

4. **NIGHT SHIFT MANPOWER PLANNING:** CUSTOMERS ALSO CALL ABC INSURANCE COMPANY AT NIGHT BUT DON'T GET AN ANSWER BECAUSE THERE ARE NO AGENTS AVAILABLE. THIS CREATES A POOR CUSTOMER EXPERIENCE. ASSUME THAT FOR EVERY 100 CALLS THAT CUSTOMERS MAKE BETWEEN 9 AM AND 9 PM, THEY ALSO MAKE 30 CALLS AT NIGHT BETWEEN 9 PM AND 9 AM. THE DISTRIBUTION OF THESE 30 CALLS IS AS FOLLOWS:

YOUR TASK: PROPOSE A MANPOWER PLAN FOR EACH TIME BUCKET THROUGHOUT THE DAY, KEEPING THE MAXIMUM ABANDON RATE AT 10%.

STEPS INVOLVED TO MAINTAIN DATA

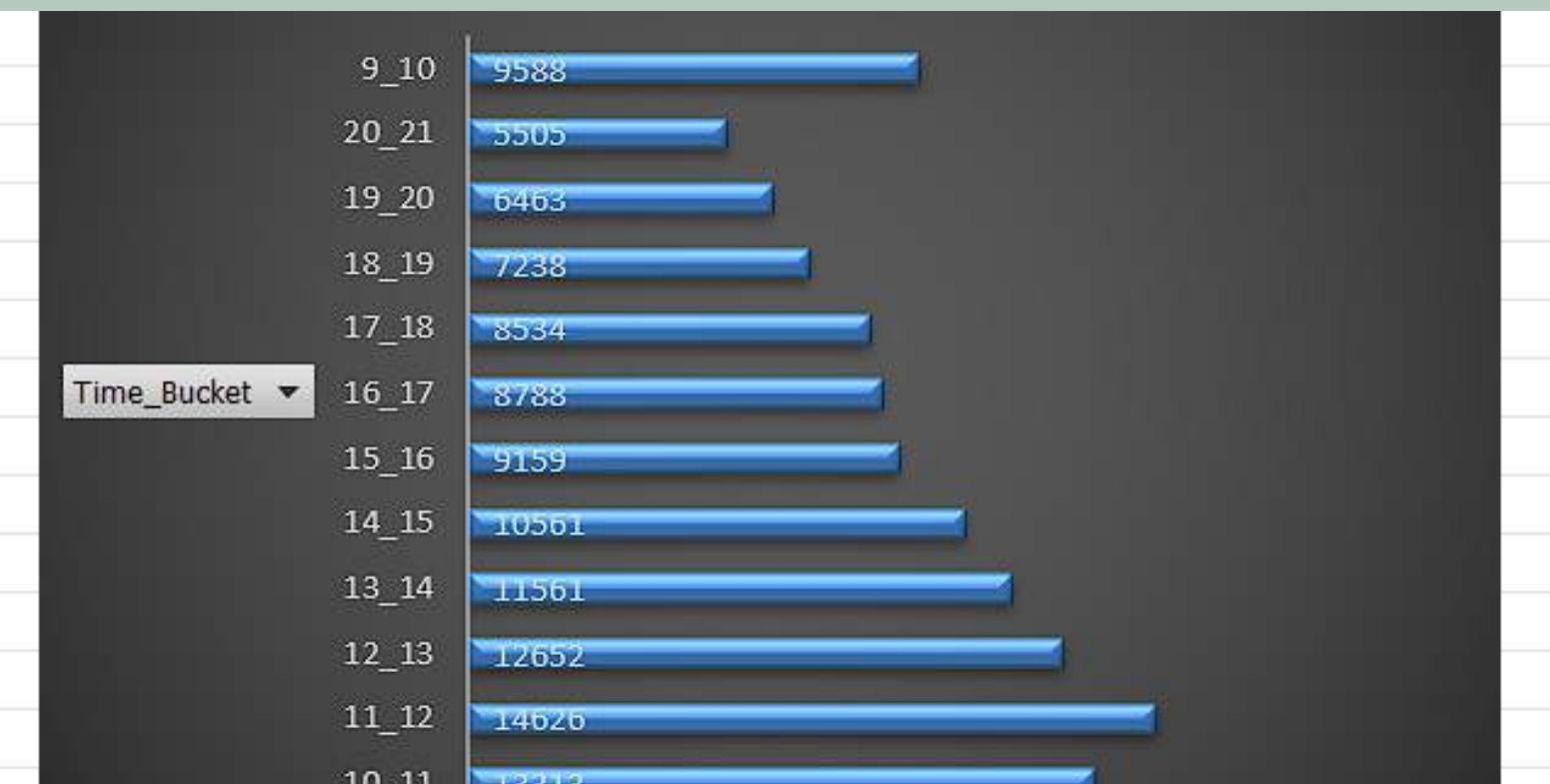
- **EXTRACT**
- **TRANSFORM**
- **LOAD**

SOLUTION

FINDING 1

Row Labels	Average of Call_Seconds (s)
10_11	97.42402163
11_12	116.7837413
12_13	144.7250237
13_14	149.5409567
14_15	146.9693211
15_16	169.8968228
16_17	181.4393491
17_18	179.7245137
18_19	174.3246753
19_20	144.5825468
20_21	105.9491371
9_10	92.01032541
Grand Total	139.5321473

FINDING 2



FINDING 3

12_13	12652	637
13_14	11561	582
14_15	10561	532
15_16	9159	461
16_17	8788	443
17_18	8534	430
18_19	7238	365
19_20	6463	326
20_21	5505	277
9_10	9588	483

FINDING 4.

Time Slot	Calls	Agents Required
9 pm - 10 pm	3	1
10 pm - 11 pm	3	1
11 pm - 12 am	2	1
12 am - 1 am	2	1
1 am - 2 am	1	1
2 am - 3 am	1	1
3 am - 4 am	1	1
4 am - 5 am	1	1
5 am - 6 am	3	1
6 am - 7 am	4	1
7 am - 8 am	4	1
8 am - 9 am	5	1

Working Hours: 9 hours/day						
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Breaks: 1.5 hours/day						
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Effective Working Hours: $9 - 1.5 = 7.5$ hours/day

Time on Calls: 60% of 7.5 hours = $0.60 * 7.5 = 4.5$ hours/day

Working Days per Month: 6 days/week * 4 weeks - 4 days (unplanned leaves) = 20 days/month

Total Hours per Month:	$4.5 \text{ hours/day} \times 20 \text{ days} = 90 \text{ hours/month}$	
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each call takes 3 minutes on average								

Calls handled per hour by one agent = 60 minutes / 3 minutes = 20 calls/hour

Calls handled per month by one agent = 20 calls/hour * 90 hours = 1800 calls/month

Total night calls per month = 30 calls/night * 30 days = 900 calls/month

Number of agents required = Total night calls / Calls handled per agent								

CONCLUSION

- 1.**AVERAGE CALL DURATION** : IDENTIFYING TIME BUCKETS WITH LONGER AVERAGE CALL DURATIONS HELPS IN OPTIMIZING AGENT ALLOCATION AND ENSURING ADEQUATE COVERAGE DURING PEAK HOURS.
- 2.**CALL VOLUME DISTRIBUTION**:VISUALIZING CALL VOLUMES HELPS IN UNDERSTANDING CUSTOMER CALLING BEHAVIOR, ALLOWING FOR BETTER PLANNING AND ALLOCATION OF AGENTS TO MEET DEMAND DURING PEAK PERIODS.
- 3.**DAY SHIFT MANPOWER PLANNING**:BY ENSURING THAT THE NUMBER OF AGENTS IS SUFFICIENT TO HANDLE 90% OF THE CALLS DURING EACH TIME BUCKET, CUSTOMER SATISFACTION CAN BE SIGNIFICANTLY IMPROVED.
- 4.**NIGHT SHIFT MANPOWER PLANNING**:BY ALLOCATING AGENTS BASED ON THE DISTRIBUTION OF NIGHT CALLS AND ENSURING COVERAGE THROUGHOUT THE NIGHT, WE CAN PROVIDE BETTER CUSTOMER SERVICE AND REDUCE THE ABANDON RATE.
- 5.**AGENT PRODUCTIVITY** : UNDERSTANDING AGENT PRODUCTIVITY HELPS IN ACCURATELY CALCULATING THE NUMBER OF AGENTS REQUIRED TO HANDLE THE EXPECTED CALL VOLUME, ENSURING OPTIMAL MANPOWER ALLOCATION

APPENDIX

PROJECT2:

HTTPS://1DRV.MS/P/C/E76F6F243AB164ED/EAWJSAUXX-
NNOY_XCEK4SHUBM7FA-LYGCKH8KJLSRJ0TIW?E=N5YOD4

PROJECT3:

HTTPS://1DRV.MS/P/C/E76F6F243AB164ED/EBKYKF4AE95DP9IT7RMREWABRS
XQCT6D2SYPGG7SL4OM-A?E=JAJKL7

PROJECT 4:

HTTPS://1DRV.MS/P/C/E76F6F243AB164ED/ESGLH2PKJH5JQWVLLJ8IY1CBMQ
BRM-C_WBSFUX3BCADS0A?E=QFAH2I

PROJECT5:

HTTPS://1DRV.MS/P/C/E76F6F243AB164ED/ERODBPXIADHLQXEEUV9EB8
GBUAIPTQKL_LFHUUYFIWDCIA?E=QHEINB

PROJECT 6:

HTTPS://1DRV.MS/P/C/E76F6F243AB164ED/ECW3SU_TW-
VJPUXN4_3OESEBIA4OCSLUZ6GXFAHYBA6_XG?E=HW6DJI

PROJECT 7:

HTTPS://1DRV.MS/P/C/E76F6F243AB164ED/EXR28AHLCLASNKS15DGR
RoBGKZBE2LE5QQMC-RV56ZCGG?E=SZQMPA

PROJECT 8:

HTTPS://1DRV.MS/P/C/E76F6F243AB164ED/EA65IASBN_XMHVCJP2L
3HGQBVOD7_R7H0BWLYH5_OIFWVA?E=X0PVAT