## **Objective Questions**

## Q1. What is the total no. of tables present in the data?

**Ans:** There is only one table in the data.

## Q2. What is the total no. of attributes present in the data?

**Ans:** There are total 35 attributes present in the data.

## Q3. The data consists of some inconsistent and missing values so ensure that the data used for further analysis is cleaned.

Ans: Yes, the data contained some missing and inconsistent values.

- > To understand the data structure:
  - I thoroughly **analyzed each column** to understand its significance.
  - Applied FILTER functions on key columns such as Consultation Type, CallStatus, Website, FreeCall, and FreeChat.
- > Null values were contextually valid:
  - If Consultation Type = "Chat", then CallStatus is null (which is expected).
  - Similarly, if it's a **FreeCall** or a chat session, then Amount may be blank, which is logical.
- > Inconsistent values handled:
  - Found inconsistencies between Amount and NetAmount.
  - Updated **NetAmount** using the formula:
  - NetAmount = Total Amount OperationCost
- > Missing values in the **Region** column:
  - Filled using **IF** and **ISBLANK** functions to assign default (Indian) value.
- > Date components extracted:
  - Extracted Year, Month, and Day using TEXT(), YEAR(), MONTH(), and DAY() functions for better time-based analysis.
- > Final step data pruning:
  - Selected only the relevant columns required for analyzing Objective and Subjective questions, removing unnecessary columns like \_\_v, callvrType, time duration etc.

## Q4. What is the change in daily call volume day by day and also find the average daily call volume.

## Ans: To answer this question I have followed step by step by process

## Selected unique dates

> Extracted unique values from the Date column using the **UNIQUE()** function.

## Calculated daily call volume

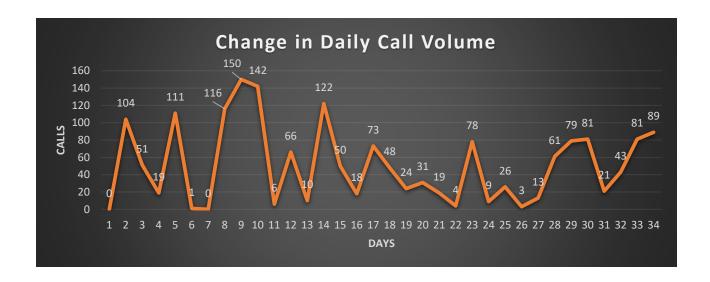
> Used the **CallSid** column to count the number of calls on each day (non-blank **CallSid** values indicate valid calls).

## Computed average daily call volume

> Applied the **AVERAGE()** function to the daily call counts to get the mean volume of calls per day.

## Calculated daily change in call volume

➤ Used a formula like = CurrentDayVolume - PreviousDayVolume to compute the day-over-day change in call volume.



Average Daily Call Volume 246.0294

#### **Observations:**

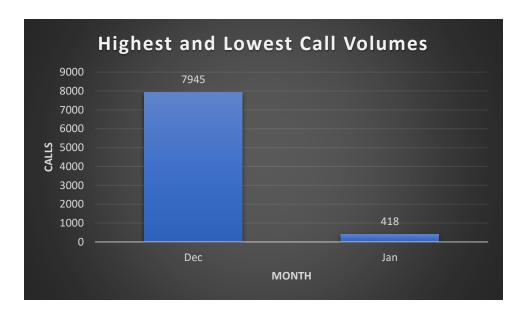
- > The daily call volume **fluctuates significantly** from one day to the next.
- > There are several **sharp spikes and drops**, indicating irregular call traffic (e.g., big spike on Day 9, sudden dip on Day 11).
- This pattern shows high volatility, possibly due to changes in user behaviour, campaign activity, or external events.
- > The average change in daily call volume is approximately **246 calls**, showing large day-to-day variability.
- No clear trend of increase or decrease the changes are erratic, not progressive

Reference: The Analysis is provided on sheet name "CallVolume"

## Q5. Which months experienced the highest and lowest call volumes?

## Ans Approach I have taken:

- > Created a **Pivot Table** using the dataset.
- > Placed "Month" in the Rows section to group data month-wise.
- Placed "CallSID" in the Values section and set it to Count, representing the total number of calls per month.



#### Observation:

- > The month with the highest call volume is: December
- > The month with the lowest call volume is: January

Reference: The Analysis is provided on sheet name "pivot"

## Q6. What is the total operational cost for that month?

## Ans: Approach I have Taken:

- Created a Pivot Table using the dataset.
- Placed "Month" in the Rows section to organize data by month.
- Placed "Astrologer's Earnings" in the Values section to calculate the total operational cost, since it is the only cost-related field provided in the data.



#### Observation:

- ➤ In **December** month the total operational cost was **73,925.84**
- In January month the total operational cost was 3873.59

**Note:** For the Q5 & Q6 Here we don't have much data for January month there are only 3 dates available in the data so we can't exactly say that in December month the call was the highest or operational cost is high

Reference: The Analysis is provided on sheet name "pivot"

## Q7. What is the average number of calls handled per agent per day?

#### Ans: Approach I have taken:

#### 1. Total Number of Calls:

Created a Pivot Table by filtering **consultation type = "Call"** and using the **callSid** column in the *Values* section to count total calls.

#### 2. Total Number of Agents:

Applied the **UNIQUE filter** on the Agent Name column to extract the distinct number of agents.

## 3. Total Number of Days:

Used the **UNIQUE filter** on the Date column to get the number of active days in the dataset.

## 4. Applied the Formula:

Average calls per agent per day 
$$=\frac{8363}{149\times34}\approx1.65$$

TotalNo.ofCalls	8363
TotalNo.ofAgents/gurus	149
TotalNo.ofDays	34
AverageNo.ofCallsPerAgentPerDay	1.650809

Reference: This Analyses is Provided in the sheet name

## Q8. How many repeat callers are there, and what percentage of total calls do they represent?

## Ans: Approach:

- 1. Step 1 Total Calls:
  - Created a Pivot Table using the userId column in Rows.
  - > Filtered the consultation type to "Call".
  - > In Values, used the callSid column to get total calls in grand total.
  - > Then we use **COUNT** function to get the **total no.of callers**
- 2. Step 2 Identify Repeat Callers:
  - > For each user\_id I will reduce -1 call from its total calls
  - > This gave the count of **repeat calls**.
  - > Then again I will use **COUNT** function to count **totalno.of repeat** callers
- 3. Step 3 Calculate Percentage of Repeat Calls:

**Used this formula**: Percentage of Repeat Calls = 
$$\left(\frac{\text{Total Calls by Repeat Callers}}{\text{Total Calls}}\right) \times 100$$

<sup>&</sup>quot;AverageCallsPerAgent"



3629
704

#### Observation

Total Callers: 3,629Repeat Callers: 704

> Repeat Calls percentage: ≈ 56.61%

## **Key Insights**

- 704 out of 3,629 callers (≈19.4%) are repeat callers, which means nearly 1 in 5 users called more than once.
- ➤ These repeat callers contributed to 56.61% of total calls, indicating a high level of re-engagement or unresolved concerns.
- Repeat callers are making more calls per person compared to first-time users.

## This suggests:

- A strong dependency on the service (engagement).
- Or a potential gap in issue resolution during the first call.

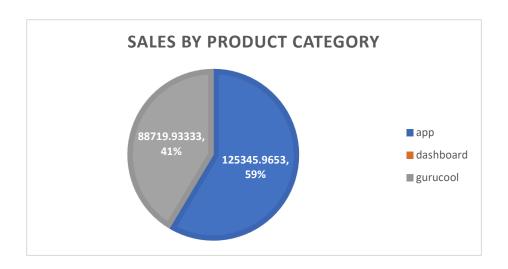
Reference: This Analyses is Provided in the sheet name "RepeatCallers"

## Q9. What are the total sales generated by the call centre for each product category?

**Ans:** Approach I have taken:

To calculate total sales per product category:

- Created a Pivot Table.
- Rows: Selected Website column (each product is tied to a domain or website).
- > Values: Selected Amount column to calculate total sales.
- > This aggregated total sales for each product offering.



#### Observation:

> App: ₹125,435.95 Highest sales contributor.

**Gurucool**: ₹88,719.93 Performing well, close to the App.

> **Dashboard**: ₹0 No sales

## **Key Insight:**

> Sales are largely driven by the **App** and **Gurucool**.

> **Dashboard** requires further investigation either underperforming or inactive.

Reference: The Analysis is provided on sheet name "pivot"

## Q10 How many calls were made for each user ID and guru ID?

Ans: Approach I have taken to find the **total number of calls made per User ID** and Guru ID, follow these steps:

## 1. Filter the data:

> Apply a filter where **Consultation Type = 'Call'** to ensure only call-based interactions are counted.

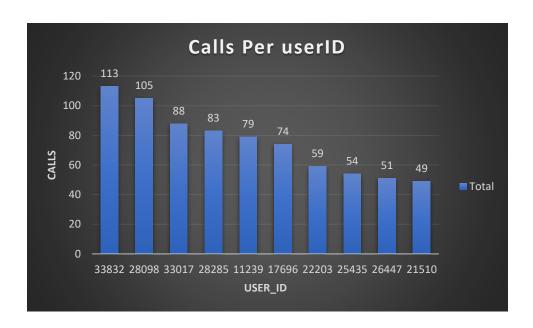
### 2. For User ID Call Count:

- > Create a **Pivot Table**.
- > Set Rows as userId.
- Set Values as callSid (use Count function).
- > This gives the total number of calls made by each user.

#### 3. For Guru ID Call Count:

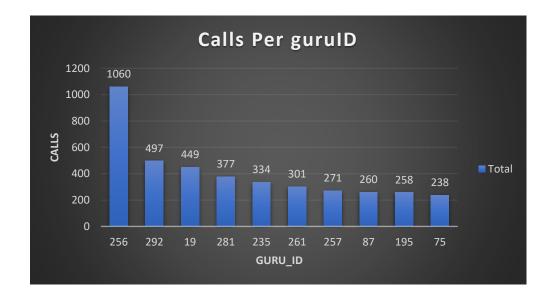
- Similarly, create another Pivot Table.
- > Set Rows as guruld.
- Set Values as callSid (Count).

> This gives the total number of calls received or handled by each Guru.



#### Observation of userID:

- > User 33832 made the highest number of calls: 113 calls
- > Followed closely by User 28098 with 105 calls
- > Remaining top users (e.g., 33017, 28285, 11239) made between **70–90 calls**
- > The top 10 users range between 49 to 113 calls
- > Steep drop in call frequency beyond the top few users



### Observation of guruID

- Guru 256 handled a very high volume of calls: 1060 calls drastically higher than others
- > Next highest: Guru 292 (497 calls) and Guru 19 (449 calls)
- > Others fall in the 200-400 range
- > Clear imbalance in call distribution among gurus

## **Key Insights from User & Guru Call Activity**

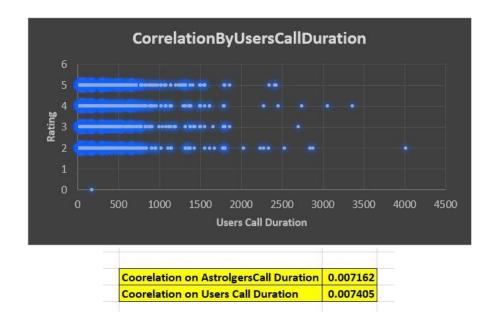
- 1. **User 33832** made the highest number of calls (**113**), indicating high engagement.
- 2. **Guru 256** handled the most calls, showing high involvement or demand.
- 3. Overall, the analysis helps identify **top active users and gurus**, useful for planning **resource allocation** and **personalized follow-ups**.

Reference: The Analysis is provided on the sheet name "pivot\_2"

## Q11. What is the correlation between call duration and customer satisfaction?

**Ans:** Approach to Analyze Correlation Between Call Duration and Customer Satisfaction

- 1. Filtered the data in general selecting the key columns:
  - Consultation Type = "Call"
  - > Call Status = "Completed"
- 2. For User Duration:
  - Column1: User Call Duration
  - Column2: Rating
  - > Applied CORREL function between User Call Duration and Rating
- 3. Repeated the same process for **Astrologer Duration**:
  - Applied CORREL between Astrologer Call Duration and Rating
- 4. Compared correlation values to determine relationship strength.



#### Observation:

Both User Call Duration and Astrologer Call Duration show very weak positive correlation with Ratings.

- 1. Correlation values:
  - > Users: **0.0074**
  - > Astrologers: **0.0071**
- 2. The **scatter plot** shows no clear trend or pattern, indicating that **longer calls** do not necessarily lead to higher satisfaction.
- 3. This suggests that **call duration alone is not a strong predictor** of customer satisfaction.

## **Key Insights:**

- ➤ **Very low correlation** (~0.007) between call duration and ratings for both users and astrologers.
- > This means that longer calls doesn't affect satisfaction other factors like communication quality matter more maybe.
- Pattern is consistent across both users and astrologers.

Reference: This Analysis is provided on Sheet name "Correlation"

## Q12 Which guru has the highest and lowest customer satisfaction scores?

**Ans:** The approach I have taken

Created a **pivot table**:

> Rows: Guru Names

> Values: Average of Rating

Applied sort (Largest to Smallest) on average ratings

Identified the guru with the highest and lowest average customer satisfaction

Created a **chart** to visually represent the comparison



### **Observation:**

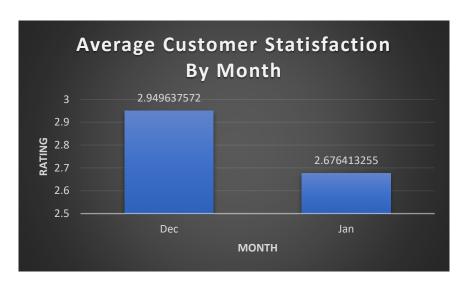
- > Top Rated Gurus:
  - Astro Pujaa Rai and Tarot Mystical received the highest average ratings
    (~7.5), indicating strong customer satisfaction and consistent positive
    feedback.
- > Lowest Rated Guru:
  - Astro Chandan received the lowest average rating (~1), reflecting significant dissatisfaction among customers.
- > Other Mentions:
  - **Tarot Rittika** also has a notably low rating, though slightly higher than Astro Chandan.

Reference: The Analysis is provided on sheet name "pivot"

## Q13 What is the average customer satisfaction score by month?

## Ans: Approach I have taken

- 1. I created a pivot table:
  - > Rows: Month
  - > Values: Average of the *Rating* column
- 2. Then used a **bar chart** to visualize the comparison between months.



### **Observation:**

The average customer satisfaction score for **December** month is **2.95** and for **January** the score is **2.67** 

**Note:** Although they look like they are almost close but it is not exact because in December month there are only 3 dates but in January we are considering whole month 31 dates so by this we cant exactly say that this the average customer satisfaction score

Reference: The Analysis is provided on sheet name "pivot"

## 14 How many categorical columns are there in the data? categorical and continuous data

Ans: There are total 35 columns are present in the data in which

27 columns are categorical and 8 columns are continuous

Categorical columns: [\_id, user, chatStatus, guru, guruname, consulation\_type, website, refund\_status, isWhilteListUser queue ,free\_call , free\_chat , createdAt, updatedAt , \_v , statmentEntry\_id , chartStartTime ,chartEndTime , callChannel , callvrType , callStatus ,callSid , astrologersCallStatus , uid , gid ,region , usercallStatus ]

**Continous columns:** [ chatSeconds, amount , astrologersCallDuration, usersCallDuration , astrologersEarnings , netAmount ,rating , timeDuration ]

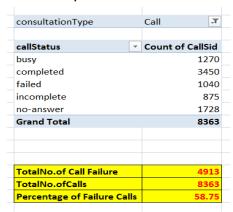
## **Subjective Questions**

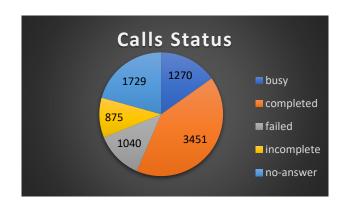
## Q1. Should the investment be used to hire more agents, improve training programs, or upgrade call centre technology?

Ans: Before making an investment my approach was to analyze Call Status data to understand the extent of unsuccessful call experiences from a customer perspective and then evaluate the overall performance of Gurus (agents) using their average ratings.

## 1. Call Status Analysis

- Created a pivot table in rows I selected callStatus
- In values I have selected callSid to know how many calls are been failed or success
- Considered the following as **unsuccessful calls**: no-answer, busy, failed, and incomplete.





### 2. Guru Performance Analysis

- Created pivot table in rows selected guru names in values selected Average rating
- > Calculated average rating per Guru.
- Filtered Gurus with average ratings > 4. Setting a bench to identify the guru's performance

•			
Row Labels	▼ Average of rating		
Acharya Arti S	4	Total No. of Gurus	149
Aham T	4.785714286	TotalNo.of Gurus having Greater than 4 rating	35
Astro Ashok	4.172859451	Percentage of guru performing well	23.49
Astro Lakshmi	4.134920635		
Astro Manish SM	4.352941176		
Astro Niddhi Guptaa	4.5		
Astro Pujaa Rai	7.5		
Astro Ruchi	4.442728443		
Astro Saraswat	5.61111111		
Astro Trisha	5.424324324		
Astro Dr Balkrisna	4.20657277		
Actro Himanshu	4 210525215		

## Insights:

- > A significant **58.75% of calls failed**, which is a critical concern likely due to technical inefficiencies or unavailability of agents.
- > Only 23.49% of Gurus are rated well by users, indicating a quality gap in service delivery.
- > The **AstrologersCallStatus** column had many missing values and was excluded for reliability reasons; **CallStatus** was used instead.

#### Recommendations:

> Invest in Guru Training Programs

*High Priority*: Only **23.49%** of Gurus are high performers. Focused training can enhance service quality and improve customer satisfaction.

- Upgrade Call Center Technology Medium Priority: With a 58.74% call failure rate, improving technical infrastructure like routing systems and server reliability is essential.
- Avoid Hiring More Agents (for now) Low Priority: Adding more agents without fixing skill gaps or tech issues may not address the root problems.

**Reference:** This analysis is provided on sheet name "pivot 2"

Q2. What are the potential risks of each investment option (hiring, training, technology upgrades), and how can they be mitigated? Name the chart/spreadsheet function you will use for solving the problem.

Ans:

## 1. Hiring More Agents

- > **Risk:** Increased headcount may lead to underutilization if call volume doesn't grow proportionately. It may also lead to higher operational costs without improving performance.
- Mitigation: First assess current agent utilization and peak-time load before hiring. Use forecasting models to predict future call demand.
- Function/Chart Used:
  - > COUNTIFS() to analyze call volume per agent
  - > Pivot Table to check agent-wise load
  - ➤ Line Chart for trend analysis of calls per day/week

### 2. Investing in Training Programs

- Risk: If training is not well-targeted, it may not improve performance and result in wasted effort and cost.
- > **Mitigation:** Focus training efforts only on low-rated Gurus. Track performance improvement before and after training.
- Function/Chart Used:
  - > AVERAGEIFS() to track average rating per Guru
  - > **Pivot Table** for before-after rating comparison
  - > Bar Chart to visualize performance improvement

## 3. Upgrading Call Center Technology

- Risk: Technology upgrades may not yield expected results if the root cause of failures is unrelated (like Guru unavailability). It also requires upfront investment.
- > **Mitigation:** Conduct root cause analysis of call failures first. Test upgrades in phases to validate improvement.
- Function/Chart Used:
  - > COUNTIF() for call status breakdown
  - > Pie Chart to visualize failed vs successful calls
  - > Trend Chart (Line) to measure improvement post-tech changes

Q3. How does AstroSage's call center performance compare to AstroGuru's average call volume, customer satisfaction, and agent performance? Will you use any aggregation function or a visualization here to solve the problem?

Ans: I currently don't have any relevant data for AstroGuru's performance.

#### I. If such data were available:

- ➤ I would use **Pivot Tables** to perform aggregation.
- > I would use **charts** for clear and effective visualization.

## II. This would help:

- Compare average call volume, customer satisfaction, and agent performance between AstroSage and AstroGuru.
- Generate useful insights for data-driven decision-making.

Q4. How can the call center improve its handling of peak call periods to ensure high customer satisfaction?

Mention the functionality you will use for giving the suggestions, will it be any aggregated function or a visualization?

**Ans**: To answer this question I have done two key analysis which helps in understanding how to improve its handling peak call periods and also a high customer satisfaction

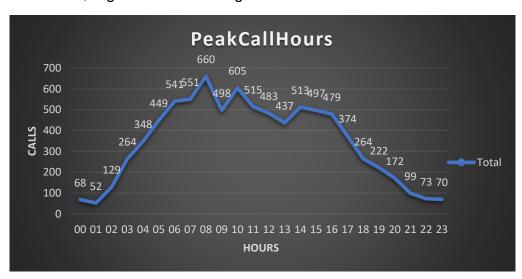
## 1. Identifying Peak Call Hours

i. Created a **pivot table** with:

a. **Rows**: Hour (0–23)

b. Values: Count of (representing number of calls)

This helped **identify time windows with the highest volume of call consultations**, regardless of which guru handled them.



## 2. Hourly Call Distribution Per Agent

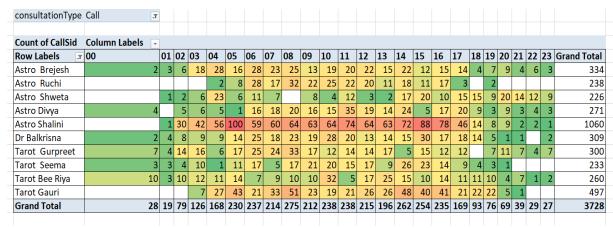
Created another **pivot table** with:

Rows: Guru NamesColumns: Hour

> Values: Count of CallSid

Applied a **colour scale heatmap using Conditional Formatting** for visual clarity. This allowed me to see **which agents are handling the most calls** and during **which hours**, helping identify:

- Overloaded gurus
- Underutilized time slots
- Potential staffing mismatches



## **Key Insights**

#### **Peak Hours for Consultations**

- i. Calls start increasing from 03:00 AM.
- ii. Peak consultation hours are:
  - > **08:00 AM to 12:00 PM** (highest spike at 08:00 with ~660+ calls).
  - > Another busy slot is 12:00 PM to 04:00 PM.
- iii. Traffic significantly drops after 18:00.

### **Top Performing Gurus (by Call Volume)**

- i. Astro Shalini handled the highest number of calls: 1,060.
- ii. Followed by:

Tarot Gauri: 497 calls
 Astro Brejesh: 334 calls
 Dr Balkrisna: 309 calls

iii. These gurus are likely preferred by users during peak hours.

## **Key Recommendations**

#### 1. Align Guru Availability with Peak Hours

Schedule top performers (e.g., Astro Shalini, Tarot Gauri, Astro Brejesh) between **08:00–16:00**. Reassign underutilized gurus (e.g., Astro Ruchi, Astro Shweta) to assist during peak or cover off-peak slots.

## 2. Strengthen Morning Coverage

Boost agent support or enable auto-queue handling between **03:00–08:00** to reduce wait times.

#### 3. Incentivize High Performers

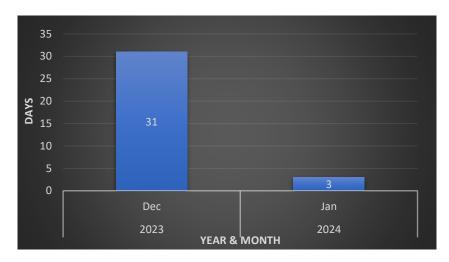
Introduce rewards to sustain top guru performance during high-demand periods.

## 4. Drive Off-Peak Engagement

Use late-hour promos post 18:00 to improve call volume during quiet period

## Q5. Based on historical data, what strategic initiatives should be prioritized to improve efficiency and customer satisfaction?

**Ans:** To answer this question, I first examined the historical data available in the dataset.



the date distribution in the graph as we can observed that:

- > For 2023, data is limited to only one month (31 days)
- > For **2024**, the data spans just **three days**

Given the limited time coverage, the dataset does not support long-term trend analysis. However, we can still extract valuable insights and propose **short-term strategic initiatives** to improve operational efficiency and enhance customer satisfaction.

To achieve this, I performed two key analyses:

#### 1. Guru-wise Peak Hour Call Handling

- i. Used a pivot table with:
  - > Rows: quruName
  - > Columns: hour
  - > Values: Count of CallSid
- ii. Filtered to show only gurus handling more than 100 calls
- iii. Helped identify top-performing gurus and their efficiency during peak hours

#### 2. Call Status vs Hour Analysis

- i. Created a pivot with:
  - > Rows: **call status** (e.g., completed, missed, no-answer, failed)
  - > Columns: hour
  - > Values: Count of CallSid

- ii. Highlighted hours with high volumes of missed or failed calls
- iii. Used **conditional formatting** to easily identify problem hours

Total.No.ofGurus	128
Total.No.ofActive Gurus	26
Percentage of well performing gurus in peak hours	20.31

		_	_	_		_			_		_			_		٠,		_		_					_
consultationType	Call .T																								
Count of CallSid	Column Labe 🔻																								
Row Labels	00	01	02	03	04	05	06	07	80	09	10	11	12	13	14	15	16	17	18	19	20	21	22	23	<b>Grand Total</b>
busy	12	2	14	33	52	70	108	75	95	70	92	81	89	72	71	67	72	62	46	41	17	17	8	4	1270
completed	10	16	58	101	138	208	239	264	302	225	249	219	177	189	232	225	227	169	84	59	26	8	12	13	3450
failed	10	4	7	33	41	48	62	52	60	64	81	57	63	39	65	64	46	47	35	44	60	27	17	14	1040
incomplete	5	8	15	38	36	50	60	64	72	55	60	46	49	47	54	56	49	34	31	17	14	7	3	5	875
no-answer	31	22	35	59	81	73	72	96	131	84	123	112	105	90	91	85	85	62	68	61	55	40	33	34	1728
Grand Total	68	52	129	264	348	449	541	551	660	498	605	515	483	437	513	497	479	374	264	222	172	99	73	70	8363

## **Insights**

## 1. Over-Reliance on a Few Agents:

A small set of gurus are handling a majority of calls, while others are underutilized.

#### 2. Service Gaps During Peak Hours:

Certain peak hours show high missed/failed call rates, indicating potential overload or availability issues.

## 3. Underperformance at Specific Hours:

Several time slots have low completion rates and high dropout (missed/no-answer), affecting customer experience.

#### Recommendations

#### 1. Reallocate Staff Based on Peak Demand

- Schedule top-performing gurus during high-traffic hours
- Rotate or assign underutilized gurus to share peak hour load

#### 2. Strengthen Guru Training Programs

- Provide performance-based feedback to low-performing gurus
- Focus on improving responsiveness and reducing call dropouts

## 3. Introduce Smart Call Handling Features

- Enable callback scheduling for unanswered calls
- Implement intelligent call routing to connect users with available and experienced gurus

## 4. Enhance System Reliability

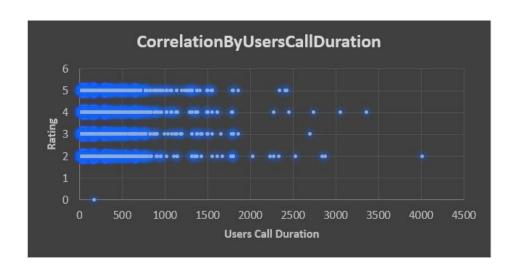
Investigate root causes of failed calls and optimize infrastructure

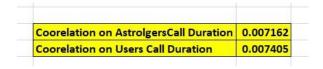
> Improve platform stability to minimize missed and dropped calls

Reference: This Analysis is provided on sheet name "pivot 4"

Q6. What can be the key factors contributing to high customer satisfaction scores, and how can these be leveraged to improve overall performance? What is the basis for the suggestions? And mention how you decided if the satisfaction score affects the ratings.

Ans: To answer this question, I first examined the relationship between call duration and customer satisfaction ratings using correlation analysis.





## **Findings from Correlation Analysis:**

- I calculated the **Pearson correlation coefficients** between satisfaction ratings Using CORREL Function on rating and call duration:
  - Astrologers' call duration → 0.0071
  - o Users' call duration → 0.0074
- These values are **very close to zero**, indicating **no significant linear relationship** between call duration and customer satisfaction.

The scatterplot (CorrelationByUsersCallDuration) further confirms this — satisfaction scores are spread across different call durations without a clear upward or downward trend.

## **Basis for the Suggestions:**

Since call duration does **not** strongly influence satisfaction, I infer that **qualitative factors** are likely driving higher customer satisfaction. These could include:

- Quality of astrological advice
- Empathy and communication style of astrologers
- Prompt connection times
- Ease of using the platform

## **Key Factors Contributing to High Satisfaction Scores:**

- 1. **Astrologer Expertise & Communication**: Customers likely value clarity, empathy, and actionable advice more than the call duration itself.
- 2. **First-call Resolution**: Faster and more relevant solutions may lead to higher satisfaction, regardless of how long the call lasted.
- 3. **Platform UX/UI and Reliability**: Smooth app experience, no lag or drop-offs, and quick astrologer connection can enhance user experience.
- 4. **Personalization & Trust**: Repeat customers may prefer astrologers who remember their context and provide tailored suggestions.

#### **Recommendations to Improve Overall Performance:**

- **Train astrologers** on soft skills, empathy, and communication strategies to boost satisfaction.
- Introduce **post-call feedback analysis** to identify top-performing astrologers and replicate best practices.
- Implement **matchmaking algorithms** that connect users with astrologers based on past satisfaction or preferences.
- Regularly monitor and enhance technical performance (e.g., app load time, call stability).

Reference: This Analysis is provided on sheet name "Correlation"

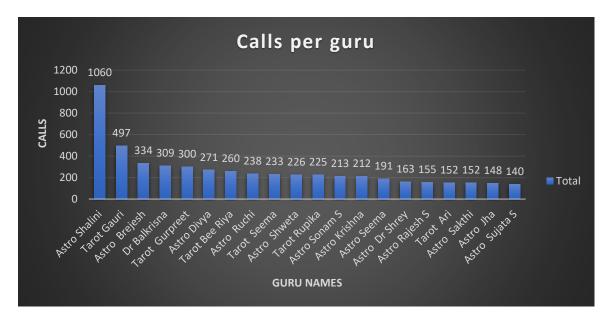
7Q How should the call centre balance the workload among agents to ensure optimal performance and avoid burnout?

Mention your approach and spreadsheet function for the answer.

**Ans:** The Approach I have taken:

- 1. I created a **Pivot Table** from the dataset with the following configuration:
  - > Filter: Consultation Type = "Call"
  - > Rows: Guru Names
  - Values: Count of Call SID (to get the total number of calls handled by each agent)

This provided a summary of **calls per agent**, which is visualized in the bar chart titled **"Calls per Guru."** 



#### Insights:

- There is a significant imbalance in the number of calls handled by agents.
- > Astro Shalini handled 1060 calls, which is double or triple the number of calls most other agents handled.
- > The next highest, **Tarot Gauri**, handled only **497 calls**, less than half of what Astro Shalini did.
- Many agents are handling fewer than 250 calls, while others handle above 300–1000.

This skewed distribution can lead to **agent burnout**, **customer dissatisfaction**, and **inconsistent service quality** 

The recommendations to have a call centre balance the workload among agents to ensure optimal performance and avoid burnout are:

#### 1. Rebalance Call Distribution:

- Implement round-robin routing or load-balancing algorithms in the call routing system.
- > Limit the maximum number of calls per agent per day/week to prevent burnout.

### 2. Use Agent Utilization Metrics:

Track average call duration and quality metrics to allocate complex queries to experienced agents and simpler ones to less loaded agents.

## 3. Hiring or Training Support:

If top-performing agents are overburdened, consider recruiting new agents or training existing low-performing agents to share the load.

## 4. Monitor Performance Periodically:

> Use this **Pivot Table method weekly** to **track agent workloads** and update routing rules accordingly.

#### 5. Introduce Incentives:

Use the data to recognize top contributors but also set performance caps to avoid long-term fatigue.

Reference: This Analysis is provided on sheet name "pivot\_2"

## Q8. What new technologies or tools could be implemented to enhance call centre operations and customer service?

**Ans:** Based on the current technologies we can upgrade the website by integrating some additional functualties like

- ➤ Al Chatbots Instantly resolve common queries and reduce agent load.
- > IVR(Interactive Voice Response) & Smart Call Routing Direct calls to the right agent faster.
- CRM(Customer Relationship Management) Integration Gives agents full customer history for better support.
- Speech Analytics Understand customer sentiment and improve communication.
- > Real-Time Dashboards Monitor agent performance and call center efficiency.

# Q9. What metrics should be included in the final dashboard to comprehensively view call centre performance and guide investment decisions?

## Ans: Daily Call Volume

To track call centre load and identify peak periods.

> Average Calls Handled per Agent per Day

To measure agent workload and efficiency.

> Repeat Caller Count & Percentage

Indicates loyalty or unresolved issues; important for resource planning.

> Call Distribution by Guru ID and User ID

Helps evaluate both agent and customer engagement patterns.

> Monthly Call Volume Trends

Useful for understanding seasonality and planning agent availability.

Customer Satisfaction Ratings

Critical for service quality evaluation.

> Correlation between Call Duration and Satisfaction

Even if low, include to show you've tested the relationship.

> Agent-Wise Satisfaction Rating

Identifies top and low-performing agents for training or rewards.

> Top 10 Gurus by Satisfaction Score

Helps in investment decisions like promotions or training.

> Call Status Breakdown (Completed, Incomplete, Failed)

Reveals quality and efficiency of operations.

Q10. How would you allocate a 1 crore rupee investment to optimize operational efficiency, enhance customer satisfaction, and boost profitability, and what analysis-based recommendations would you offer to support this?

**Ans**: To answer this question, I first conducted a key analysis on **sales by consultation type**.

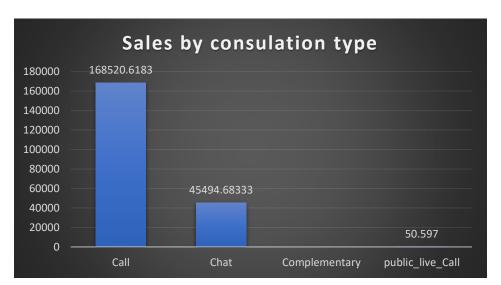
#### **Importance of This Analysis:**

- > Understanding which consultation types (e.g., calls, chats) contribute the most to revenue is essential to **prioritize investments** that yield higher returns.
- > This analysis helps in **identifying high-performing channels** that should be enhanced and low-performing ones that may need improvement or reconsideration.

It directly supports the objective of optimizing efficiency and boosting profitability by revealing where to allocate more resources strategically.

## Approach I have taken:

- > I created a **pivot table** with:
  - Rows: Consultation Type
  - Values: Sum of Amount
- This setup allowed me to compute the total sales for each consultation type and identify the primary revenue drivers.



Consulation_typ *	Sum of amount
Call	168520.6183
Chat	45494.68333
Complementary	
public_live_Call	50.597
Grand Total	214065.8987

## **Based on My Complete Analysis:**

- > Calls are the primary revenue driver, with chats contributing significantly.
- > Agent-level information is incomplete, so tracking performance is crucial.
- There's no data on customer wait time, which is an opportunity for improvement.
- Some agents underperform, suggesting a need for structured training & evaluation.

## **Investment Allocation Strategy:**

- 1. Technology & Application Upgrades (₹40 lakhs)
  - Improve call quality, reduce technical issues.
  - Upgrade user interface for chat to make it more engaging.
  - Integrate call tracking, wait time monitoring, and performance dashboards.

## 2. Agent Monitoring & Analytics Tools (₹15 lakhs)

- Track real-time performance, wait time, resolution rate, and call quality.
- Use analytics to identify top and low performers.

## 3. Agent Training & Quality Programs (₹20 lakhs)

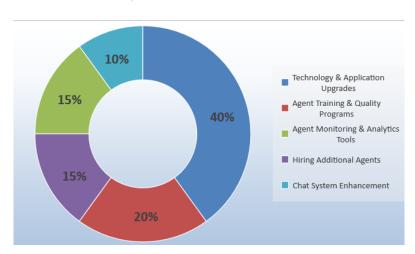
- Train new and underperforming agents on communication, product knowledge, and FCR.
- > Introduce monthly performance reviews to ensure training ROI.

## 4. Hiring Additional Agents (₹15 lakhs)

- Onboard skilled agents to manage peak-hour traffic and reduce waiting time.
- Prioritize agents with prior experience in customer handling.

## 5. Chat System Enhancement (₹10 lakhs)

- ➤ Make chat features more responsive, personalized, and multilingual.
- Automate FAQs via chatbots to reduce manual load.



Reference: This analysis is provided on sheet name "pivot\_2"