

## **Problem Statement**

The primary objective is to analyze the "Social Media and Entertainment Dataset" to uncover insights into user engagement and content popularity across various social media platforms. This analysis aims to inform strategies for enhancing user interaction and optimizing content delivery

## **Key Performance Indicators (KPIs)**

I measured the effectiveness of social media and entertainment content, by considering the following KPIs:

- 1. Average Daily Screen Time: Understanding the daily screen time of users to assess their digital consumption.**
- 2. Social Media Fatigue Levels: Identifying factors contributing to fatigue and trends among users.**
- 3. Correlation Between Average SleepTime\_hrs and Screen Time\_hrs: Measuring the impact of screen time on users' sleep quality.**
- 4. Average Monthly Expenditure on Entertainment by occupation: Identifying AND calculating the most used platforms and devices for entertainment and communication alongside their occupation and their monthly expenses.**
- 5. Digital Well-being Awareness: COUNT how many users' awareness of digital well-being tools from moderate, high, and low**
- 6. Revenue from Subscription Platforms: Calculating the average monthly expenditure on entertainment platforms.**
- 7. Impact of Tech Savviness on Content Preferences: Exploring how tech-savvy users engage with content.**

## Query 1: Average Daily Screen Time by age:

/\* Daily ScreenTime, this is for each age group\*/

```
SELECT Age, ROUND(AVG(ScreenTime_hrs), 2) AS avg_daily_ScreenTime
```

```
FROM social_media.social_media
```

```
GROUP BY Age;
```

Result Grid	Filter Rows:	Export:
Age	avg_daily_ScreenTime	
32	6.98	
62	7.1	
51	7.04	
44	6.93	
21	7.03	
16	6.99	
58	6.97	
49	6.94	
14	6.91	
63	6.9	
56	6.99	
24	6.95	
22	6.99	
28	7.01	
40	6.93	
35	7.05	
20	6.98	
36	6.95	

## Query2: Social Media Fatigue Levels

/\*THIS QUERY LOOKED AT THE TOP PLATFORMS THAT CAUSES FATIGUE LEVELS With a scale 1-10\*/

```
SELECT Primary_Platform, SocialMediaFatigueLevel_scale, COUNT(*) AS  
total_Primary_Platform
```

```
FROM social_media.social_media
```

```
GROUP BY SocialMediaFatigueLevel_scale, Primary_Platform
```

```
ORDER BY SocialMediaFatigueLevel_scale, Primary_Platform DESC
```

	Primary_Platform	SocialMediaFatigueLevel_scale	total_Primary_Platform
▶	YouTube	1	2865
	Twitter	1	2834
	TikTok	1	2827
	Instagram	1	2836
	Facebook	1	2930
	YouTube	2	2800
	Twitter	2	2788
	TikTok	2	2884
	Instagram	2	2866
	Facebook	2	2917
	YouTube	3	2853
	Twitter	3	2863
	TikTok	3	2930
	Instagram	3	2905
	Facebook	3	2905
	YouTube	4	2865

### QUERY 3: Correlation Between AverageSleepTime\_hrs and Screen Time\_hrs

/\*Correlation between sleep quality scale and Screen time,

I used AI to help me with this query since the MYSQL workbench doesn't have the CORR function\*/

SELECT

(COUNT(\*) \* SUM(Sleep\_Quality\_scale \* ScreenTime\_hrs) - SUM(Sleep\_Quality\_scale)  
\* SUM(ScreenTime\_hrs)) /

SQRT(

(COUNT(\*) \* SUM(POW(Sleep\_Quality\_scale, 2)) - POW(SUM(Sleep\_Quality\_scale),  
2)) \*

(COUNT(\*) \* SUM(POW(ScreenTime\_hrs, 2)) - POW(SUM(ScreenTime\_hrs), 2))

) AS correlation\_coefficient

FROM social\_media.social\_media;

Result Grid		Filter Rows:	Export:	Wrap Cell Content:
	correlation_coefficient			
▶	0.00807943376210403			

This shows positive correlation as its headed more to a whole number or its above 0.5

SELECT Sleep\_Quality\_scale, ScreenTime\_hrs

FROM social\_media.social\_media;

Result Grid		Filter Rows:	Export:	Wrap Cell Co
	Sleep_Quality_scale	ScreenTime_hrs		
▶	7	7.33		
	8	4.23		
	5	4.45		
	9	11.49		
	5	7.14		
	6	6.35		
	2	2.57		
	9	7.22		
	5	9.51		
	6	4.51		
	1	9.47		
	1	2.47		
	4	3.27		
	2	4.84		
	9	4		

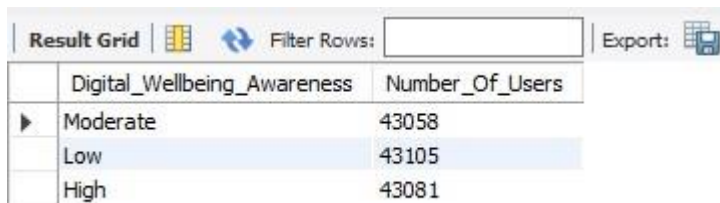
#### Query 4: Average Monthly Expenditure on Entertainment by Occupation

```
SELECT PreferredEntertainment_Platform, Occupation,  
ROUND(AVG(Monthly_Expenditure_on_Entertainment_USD), 2) AS Expenditure  
FROM social_media.social_media  
GROUP BY Occupation, PreferredEntertainment_Platform;
```

Result Grid			
Filter Rows:		Export:	Wrap Cell Content:
PreferredEntertainment_Platform	Occupation	Expenditure	
Netflix	Professional	251.3	
Spotify	Student	246.04	
Spotify	Retired	249.83	
Amazon Prime	Student	251.49	
Amazon Prime	Retired	250.91	
Amazon Prime	Unemployed	249.27	
YouTube	Student	247.32	
Amazon Prime	Professional	247.85	
Netflix	Retired	251.95	
Spotify	Professional	251.31	
Netflix	Unemployed	253.17	
YouTube	Retired	246.64	
YouTube	Professional	249.4	
YouTube	Unemployed	249.19	

## QUERY 5: Digital Well-being Awareness

```
SELECT Digital_Wellbeing_awareness, COUNT(*) AS Users
FROM social_media.social_media
WHERE Digital_Wellbeing_awareness IN ('Moderate', 'high', 'low')
GROUP BY Digital_Wellbeing_awareness
```

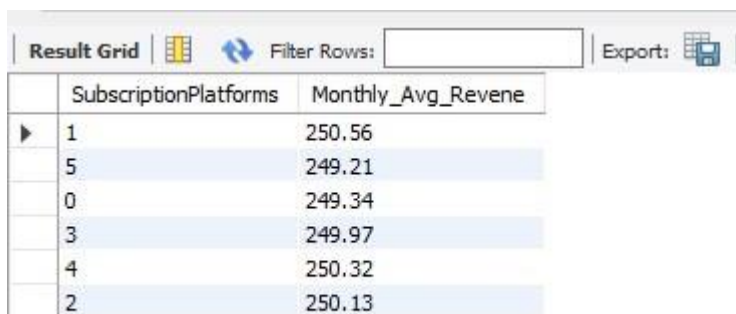


The screenshot shows a database interface with a 'Result Grid' tab. It includes a 'Filter Rows' search bar and an 'Export' button. The data is presented in a table with two columns: 'Digital\_Wellbeing\_Awareness' and 'Number\_Of\_Users'.

	Digital_Wellbeing_Awareness	Number_Of_Users
▶	Moderate	43058
	Low	43105
	High	43081

## QUERY 6: AVG Revenue from Subscription Platforms

```
SELECT SubscriptionPlatforms,
Round(AVG(Monthly_Expenditure_on_Entertainment_USD), 2) AS Monthly_Avg_Revene
FROM social_media.social_media
GROUP BY SubscriptionPlatforms;
```



The screenshot shows a database interface with a 'Result Grid' tab. It includes a 'Filter Rows' search bar and an 'Export' button. The data is presented in a table with two columns: 'SubscriptionPlatforms' and 'Monthly\_Avg\_Revene'.

	SubscriptionPlatforms	Monthly_Avg_Revene
▶	1	250.56
	5	249.21
	0	249.34
	3	249.97
	4	250.32
	2	250.13

## QUERY 7: Impact of Tech Savviness on Content Preferences

```
SELECT PreferredContent_Type, AVG(Tech_Savviness_Level_scale) AS  
AVG_Tech_savviness
```

```
FROM social_media.social_media
```

```
GROUP BY PreferredContent_Type
```

```
ORDER BY AVG_Tech_savviness
```

Result Grid			Filter Rows:	Export:
	PreferredContent_Type	AVG_Tech_savviness		
▶	Movies	4.9973		
	Series	4.9977		
	News	5.0046		
	Short Videos	5.0225		

## **Project Conclusion**

This project provided valuable insights into digital consumption patterns. By leveraging SQL for data exploration and Tableau and Power BI for visualization, we delivered actionable recommendations to improve users' digital well-being.



## Reference:

[Social media dataset](#)