

ADTECH ANALYTICS PROJECT (SQL + PYTHON + POWER BI)

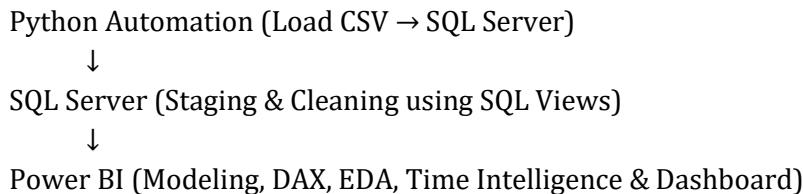
❖ PROJECT OVERVIEW

This AdTech analytics project delivers an end-to-end data pipeline and BI solution covering automated data ingestion, SQL-based data cleaning, modeling, DAX measures, and visually rich Power BI dashboards for media-performance insights.

📊 DATASET OVERVIEW

- 120,000 ad-event records (Impressions, Clicks, Conversions)
- 50+ fields covering campaign, creative, publisher, geography, device, and timestamps
- Contains cost, revenue, fraud flags, viewability, and engagement metrics
- Multiple inconsistent advertiser/publisher names
- Timestamp and URL cleaning required

🏗 PROJECT ARCHITECTURE



🐍 PYTHON AUTOMATION – CSV TO SQL

Automated ingestion loads all CSV files into SQL Server tables using SQLAlchemy and Windows Authentication.

```
from sqlalchemy import create_engine
import pandas as pd
```

```
import os


# -----
# Database Connection Settings (Windows Auth)
# -----


DB_SERVER = "DESKTOP-EN2BGT5\SQLEXPRESS" # or "localhost" for default instance
DB_NAME = "Adtech" # your database name


# Create SQLAlchemy engine using trusted connection
engine = create_engine(
    f"mssql+pyodbc://@{DB_SERVER}/{DB_NAME}?driver=ODBC+Driver+17+for+SQL+Server&trusted_connection=yes"
)


# -----
# Folder containing CSV files
# -----


DATA_FOLDER = r"C:\Users\dell\Pictures\Adtech_project" # change to your folder path


# Loop through all CSV files and upload
for file in os.listdir(DATA_FOLDER):
    if file.lower().endswith(".csv"):
        file_path = os.path.join(DATA_FOLDER, file)
        table_name = os.path.splitext(file)[0] # remove '.csv'
        print(f"📦 Uploading {file} → Table: {table_name}")
        df = pd.read_csv(file_path, encoding="utf-8")
        df.to_sql(table_name, con=engine, if_exists="replace", index=False)
        print(f"✓ Uploaded {file} ({len(df)} rows)\n")
```

```
print("🎉 All CSV files uploaded successfully!")
```

✍️ SQL DATA CLEANING (VIEW LOGIC)

A SQL View (vw_Adtech_Cleaned) standardizes advertiser/publisher names, removes special characters, fixes timestamps, flags duplicates, normalizes device & geo fields, and prepares metrics for Power BI modeling.

```
***** VIEW OF CLEANED DATA *****
```

```
USE Adtech;
```

```
GO
```

```
CREATE VIEW vw_Adtech_Cleaned AS
```

```
SELECT
```

```
ImpressionID,
```

```
-- NEW Duplicate flag logic using COUNT() OVER()
```

```
CASE
```

```
WHEN COUNT(ImpressionID) OVER (PARTITION BY ImpressionID) > 1 THEN 'Duplicate'
```

```
ELSE 'Unique'
```

```
END AS Clean_ImpressionID,
```

```
-- Clean advertiser and publisher
```

```
CASE
```

```
WHEN AdvertiserName IN ('ACME_CORP','ACME COIP','ACMETCORP','ACME_RORP','CCME CORP','CCME CORP LTD','ADME CORP','ACMK CORP') THEN 'ACME CORP'
```

```
WHEN AdvertiserName IN ('BLUESK VTD','BLUESKY_LJD') THEN 'BLUESKY'
```

```
WHEN AdvertiserName IN ('MARKETIFYH','MARKETIFYT','MARKETIFYY','MARKETIFYS','MARKETIFYZ') THEN 'MARKETIFY'
```

```
ELSE REPLACE(REPLACE(UPPER(TRIM(REPLACE(TRANSLATE(AdvertiserName, CHAR(9)+CHAR(10)+CHAR(13), ' '), CHAR(160), ")))), '_LTD',''), 'LTD','')
```

```
END AS Clean_AdvertiserName,
```

```
REPLACE(UPPER(TRIM(REPLACE(TRANSLATE(PublisherName, CHAR(9)+CHAR(10)+CHAR(13), ' '),  
CHAR(160), "))),'!','') AS Clean_PublisherName,
```

```
-- Campaign
```

```
REPLACE(UPPER(TRIM(REPLACE(TRANSLATE(CampaignName, CHAR(9)+CHAR(10)+CHAR(13), ' '),  
CHAR(160), "))),'!','') AS Clean_CampaignName,
```

```
-- Ad details
```

```
UPPER(TRIM(AdFormat)) AS AdFormat,  
UPPER(TRIM(CreativeID)) AS CreativeID,  
UPPER(TRIM(CreativeSize)) AS CreativeSize,
```

```
-- Device & Geo
```

```
CASE  
WHEN DeviceType IN ('mobile','Mobile','MOBILE') THEN 'Mobile'  
WHEN DeviceType IN ('desktop','Desktop','DESKTOP') THEN 'Desktop'  
ELSE 'Other'  
END AS DeviceType,
```

```
REPLACE(TRIM(UPPER(COALESCE(Browser,'NA'))),'CHROME','CHROME') AS Browser,  
TRIM(UPPER(COALESCE(Country,'NA'))) AS Country,  
UPPER(TRIM(City)) AS City,  
COALESCE(GeoLat,0) AS GeoLat,  
COALESCE(GeoLong,0) AS GeoLong,
```

```
-- Time details
```

```
EventTime_Clean,  
CAST(EventTime_Clean AS DATE) AS EventDate,  
YEAR(EventTime_Clean) AS EventYear,  
MONTH(EventTime_Clean) AS EventMonth,
```

```
DATEPART(WEEK, EventTime_Clean) AS EventWeek,  
DATEPART(HOUR, EventTime_Clean) AS EventHour,  
  
-- Numeric metrics  
Revenue_Num,  
Cost_Num,  
ViewabilityPct,  
Viewability_Flag,  
  
-- URL & UTM  
LOWER(LTRIM(RTRIM(REPLACE(LandingPageURL, 'http://', 'http://')))) AS LandingPageURL,  
LandingDomain,  
UTM_Source,  
UTM_Medium,  
  
-- Engagements  
ClickToView_ms,  
TimeToConversion_s,  
PageLoad_s,  
  
-- Fraud flags  
IsFraud,  
SessionID,  
UserAgent  
  
FROM adtech_dataset_main_120k;  
GO
```

POWER BI MODELING & DAX

Key DAX measures include:

- CTR, CVR, eCPM, CPC, ROAS, ROI
- Cumulative Revenue, YTD/MTD Revenue
- Fraud % detection & segmentation
- DayPart classification using SWITCH()
- Publisher/Campaign rankings using RANKX

***** DAX SUMMARY *****

Total_impression

```
Dynamic_Total_Impressions = CALCULATE(
    COUNTROWS(adserver_feed_20k),
    adserver_feed_20k[EventType] = "impression"
)
```

Total_conversion

```
Dynamic_Total_Conversions = CALCULATE(
    COUNTROWS(adserver_feed_20k),
    adserver_feed_20k[EventType] = "conversion"
)
```

Total_click

```
Dynamic_Total_Clicks = CALCULATE(
    COUNTROWS(adserver_feed_20k),
    adserver_feed_20k[EventType] = "click"
)
```

CTR (click through rate) = showing how effective your content is at getting users to click

```
CTR = DIVIDE([Dynamic_Total_Clicks], [Dynamic_Total_Impressions], 0)
```

👉 CVR (conversion rate) = which is the percentage of users who complete a desired action

```
CVR = DIVIDE([Total_Conversion], [Total_Click], 0)
```

👉 eCPM(Revenue per thousand impression)

```
eCPM(Revenue per thousand impression) =  
DIVIDE(SUM(vw_Adtech_Cleaned[Revenue_Num]) * 1000, [Total_Impressions], 0)
```

👉 Cost per click

```
Cost_per_click =  
DIVIDE(SUM(vw_Adtech_Cleaned[Cost_Num]), [Total_Click], 0)
```

👉 ROI (return on investment)

```
ROI (Return on investement) =  
DIVIDE((SUM(vw_Adtech_Cleaned[Revenue_Num]) -  
SUM(vw_Adtech_Cleaned[Cost_Num])), SUM(vw_Adtech_Cleaned[Cost_Num]))
```

👉 Cumulative Revenue

```
Cumulative Revenue =  
CALCULATE(  
    [Total_revenue],  
    FILTER(  
        ALL(vw_Adtech_Cleaned),  
        vw_Adtech_Cleaned[EventDate] <=  
        MAX(vw_Adtech_Cleaned[EventDate])  
    )  
)
```

👉 Revenue YTD

```
Revenue YTD =  
TOTALYTD([Total_revenue], vw_Adtech_Cleaned[EventDate].[Date])
```

👉 Revenue MTD

```
Revenue MTD = TOTALMTD([Total_revenue],  
vw_Adtech_Cleaned[EventDate].[Date])
```

👉 Revenue Previous Year

```
Revenue_Prev_Year = CALCULATE([Total_revenue],  
SAMEPERIODLASTYEAR(vw_Adtech_Cleaned[EventDate].[Date]))
```

Revenue last 30 days

```
Revenue_last_30_Days =  
CALCULATE([Total_revenue],DATESINPERIOD(vw_Adtech_Cleaned[EventDate].[Date],MAX(vw_Adtech_Cleaned[EventDate]),-30,DAY))
```

ROAS (Return on Ad Spend)

```
ROAS = DIVIDE([Total_revenue],[Total Cost])
```

Top N Conversion

```
TopN_conversion =  
RANKX(ALL(vw_Adtech_Cleaned[Clean_CampaignName]),[Avg_TimeToConversion_s],,,DESC,Dense)
```

Top N eCPM

```
TopN_eCPM =  
RANKX(  
    ALL(vw_Adtech_Cleaned[Clean_PublisherName]),  
    [eCPM(Revenue per thousand impression)],  
    ,  
    DESC  
)
```

Avg Click_views

```
Avg_ClickToView_ms =  
CALCULATE(  
    AVERAGE(vw_Adtech_Cleaned[ClickToView_ms]),  
    FILTER(  
        vw_Adtech_Cleaned,  
        vw_Adtech_Cleaned[ClickToView_ms] > 0  
    )  
)
```

Avg Time_To_conversion

```
Avg_TimeToConversion_s =  
CALCULATE(  
    AVERAGE(vw_Adtech_Cleaned[TimeToConversion_s]),  
    FILTER(  
        vw_Adtech_Cleaned,  
        vw_Adtech_Cleaned[TimeToConversion_s] > 0  
    )  
)
```

👉 Fraud_Impression

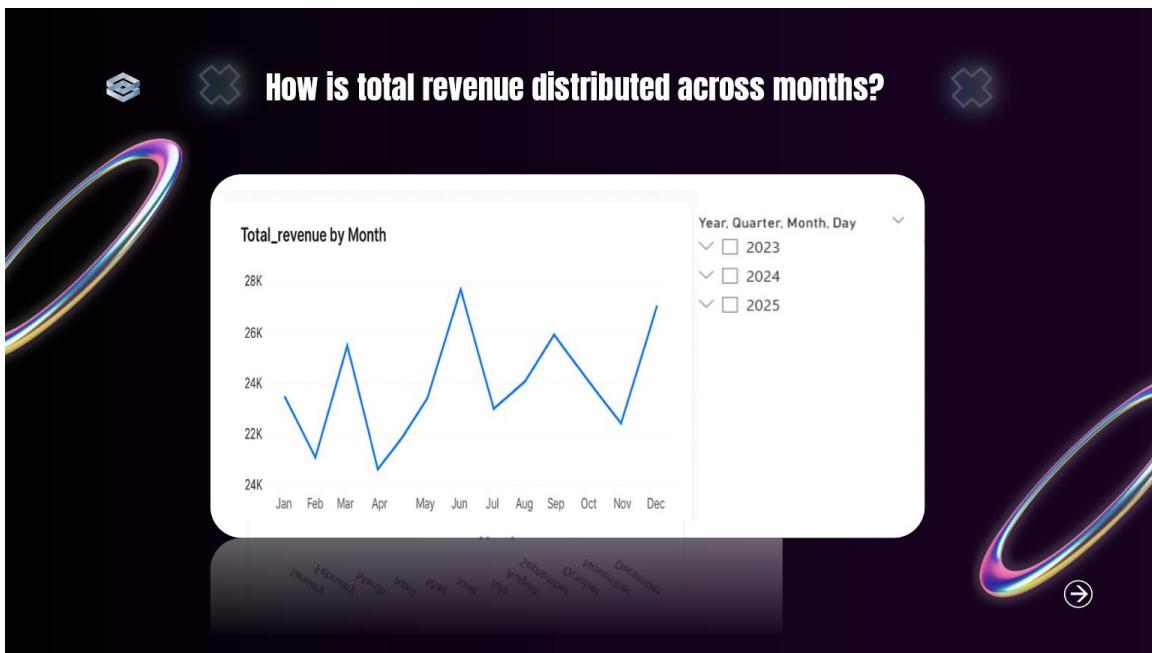
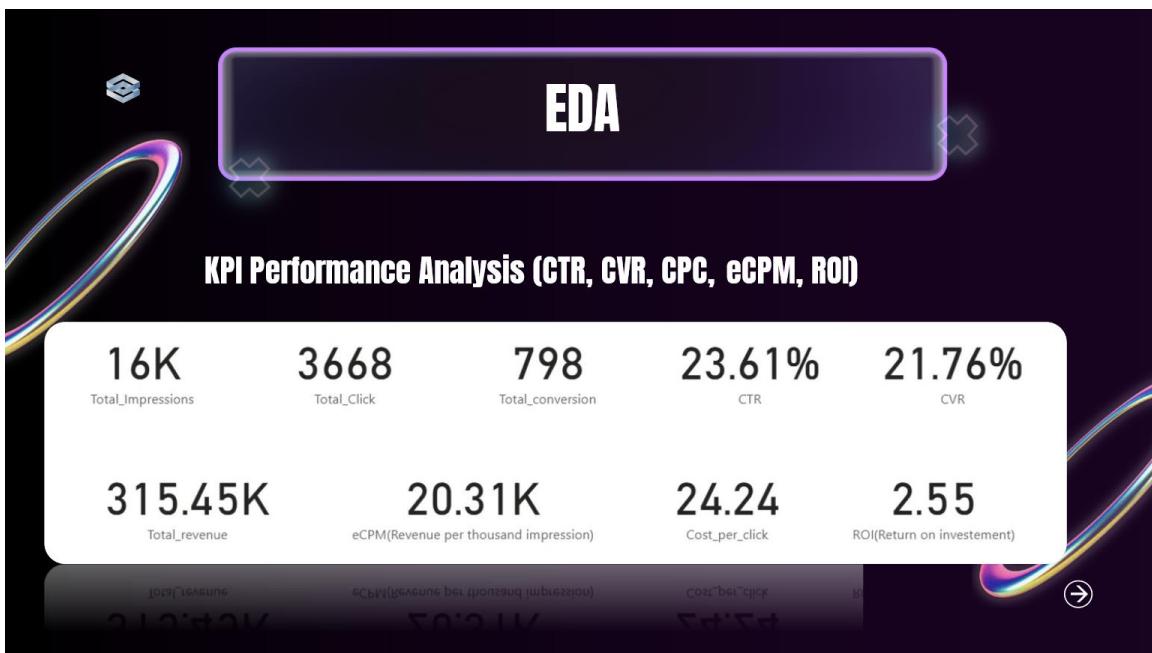
```
Fraud_Impressions =  
CALCULATE([Total_Impressions], vw_Adtech_Cleaned[IsFraud] = 1)
```

👉 Fraud_Percentage

```
Fraud_Percentage =  
DIVIDE(  
    CALCULATE(COUNTROWS(vw_Adtech_Cleaned), vw_Adtech_Cleaned[IsFraud]  
= 1),  
    COUNTROWS(vw_Adtech_Cleaned),  
    0  
)
```

👉 Day part

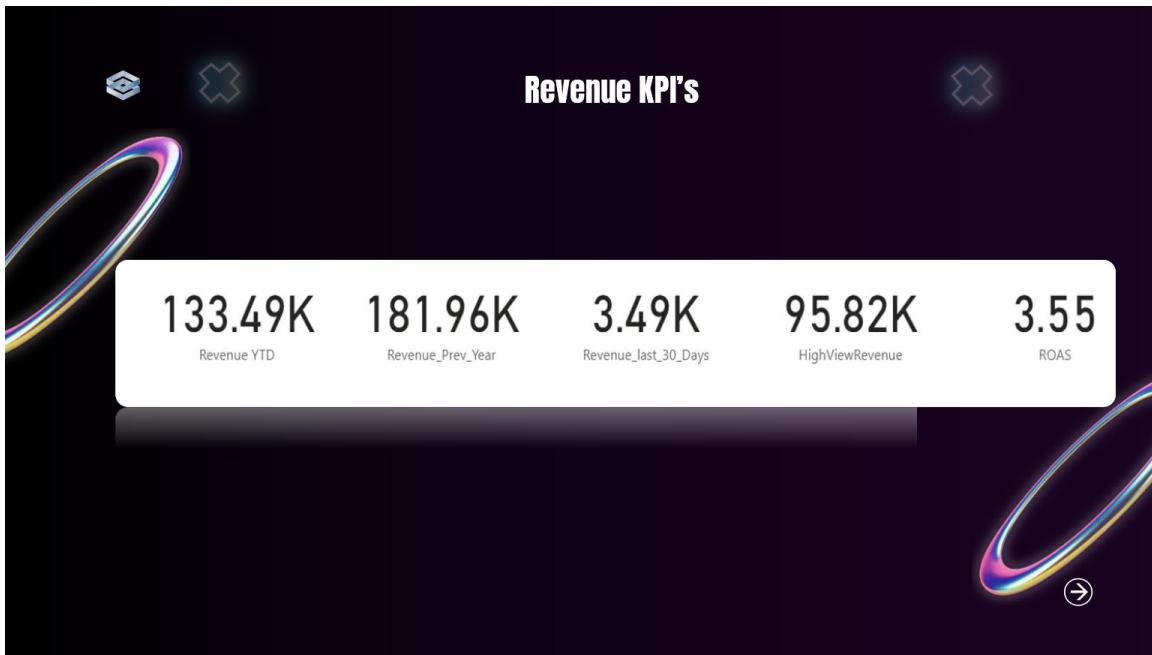
```
DayPart =  
SWITCH(  
    TRUE(),  
    vw_Adtech_Cleaned[EventHour] >= 0 && vw_Adtech_Cleaned[EventHour] <  
6, "Late Night (12AM-6AM)",  
    vw_Adtech_Cleaned[EventHour] >= 6 && vw_Adtech_Cleaned[EventHour] <  
12, "Morning (6AM-12PM)",  
    vw_Adtech_Cleaned[EventHour] >= 12 && vw_Adtech_Cleaned[EventHour]  
< 18, "Afternoon (12PM-6PM)",  
    "Evening (6PM-12AM)"  
)
```

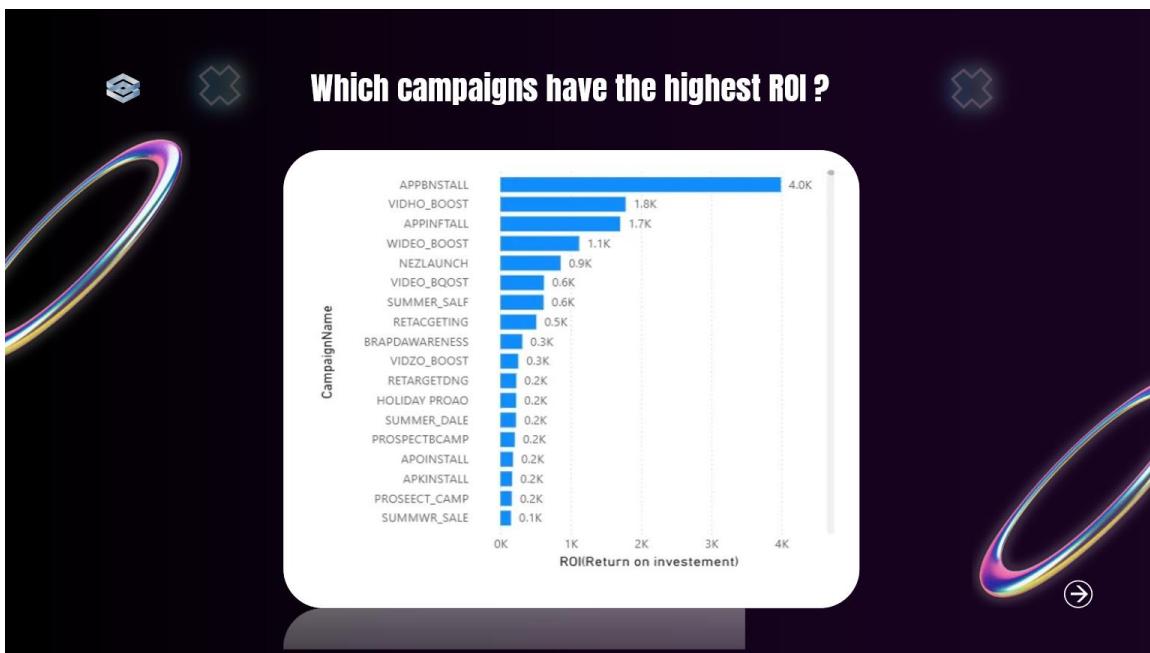


How is total revenue distributed across months?

Clean_CampaignName	Desktop	Mobile	Other	Total
AAPINSTALL	1.80	1.80	0.00	1.80
AAPINSTALL	0.00	0.00	0.00	0.00
AEPINSTALL	0.00	0.23	0.23	0.23
AETARGETING	0.00		1.23	1.23
AGPINSTALL		0.00		0.00
AIDEO_BOOST	0.00		0.00	0.00
AKPINSTALL	0.00	0.00	0.00	0.00
ALPINSTALL	0.00		0.00	0.00
AMPINSTALL	0.00	0.00	0.00	0.00
AOLIDAY PROMO	0.00			0.98
Total	1,00,560.80	1,59,858.14	55,032.65	3,15,451.59

Below the table, there is a dark bar with some text and numbers that are mostly illegible due to the low contrast.





Which campaigns show zero or negative ROI & ROAS?

Clean_CampaignName	Publishername	Cost	ROI	ROAS
APINSTALL	NEWSTODAY	0.61	-1.00	0.00
APINSTALL	QOVIEHUB	1.11	-1.00	0.00
APINSTALL	COOKINGBLOG	2.15	-1.00	0.00
APINSTALL	FASHIONMAG	0.52	-1.00	0.00
APINSTALL	COOKINGBLOG	1.40	-1.00	0.00
APINSTALL	SPORTSDAILY	0.14	-1.00	0.00
APINSTALL	COOKINGBLOG	0.33	-1.00	0.00
APINSTALL	FASHIONMAG	1.25	-1.00	0.00
APINSTALL	TECHFORUM	0.67	-1.00	0.00
APINSTALL	SPORTSDAILY	0.17	-1.00	0.00
VIDEO_BOOST	MOVIEHUB	1.25	-1.00	0.00
APINSTALL	FASHIONMAG	1.31	-1.00	0.00
APINSTALL	SPORTSDAILY	1.21	-1.00	0.00
total		88,925.63	2.55	3.55

Is there a significant CTR -CVR imbalance across publishers?

CTR Percentage: 23.61%

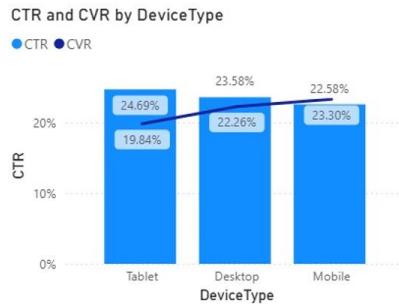
CVR Percentage: 21.76%

CTR and CVR by PublisherName

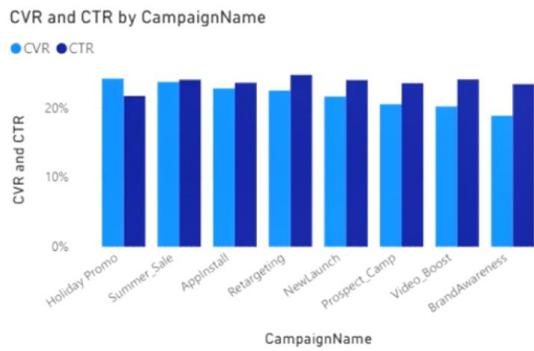
Legend: CTR (Blue Line), CVR (Purple Line)

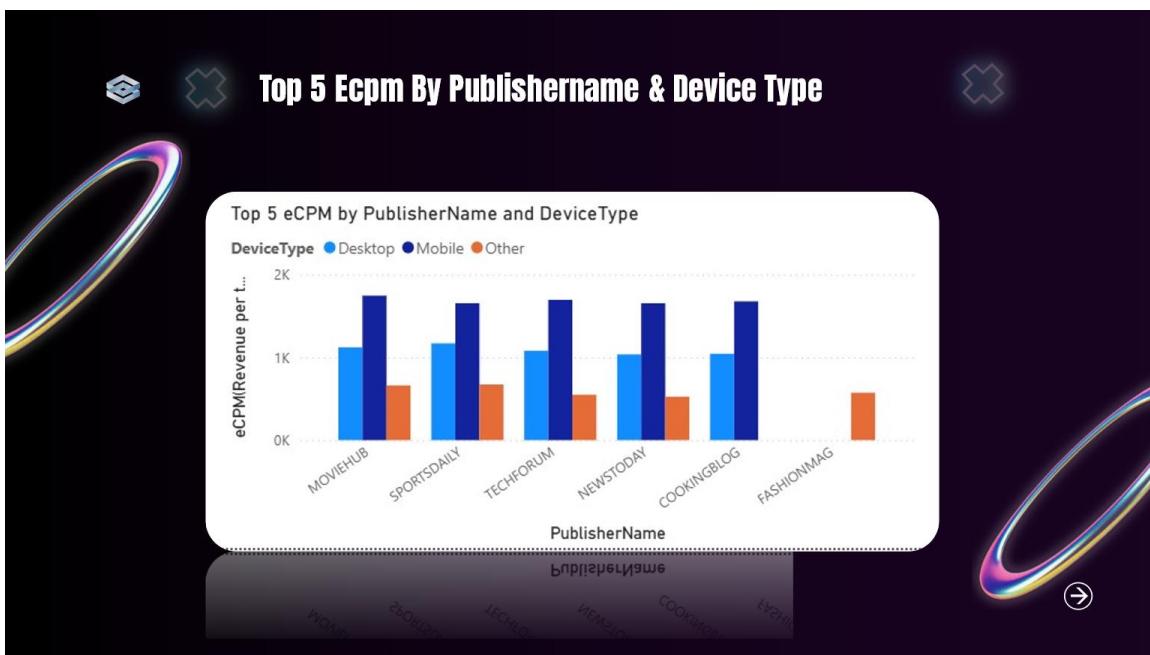
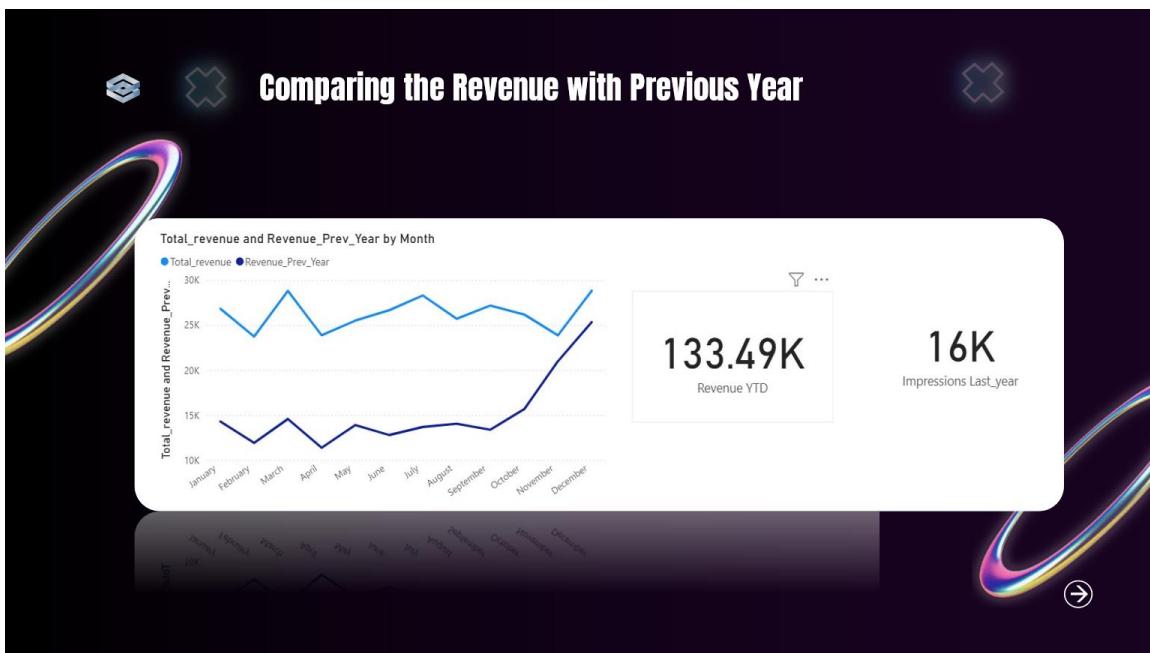
PublisherName	CTR (%)	CVR (%)
CookingBlog	24.5	22.5
MovieHub	22.5	21.5
NewsToday	23.5	22.5
TechForum	24.0	22.0
FashionMag	23.5	21.0
SportsDaily	22.5	21.5

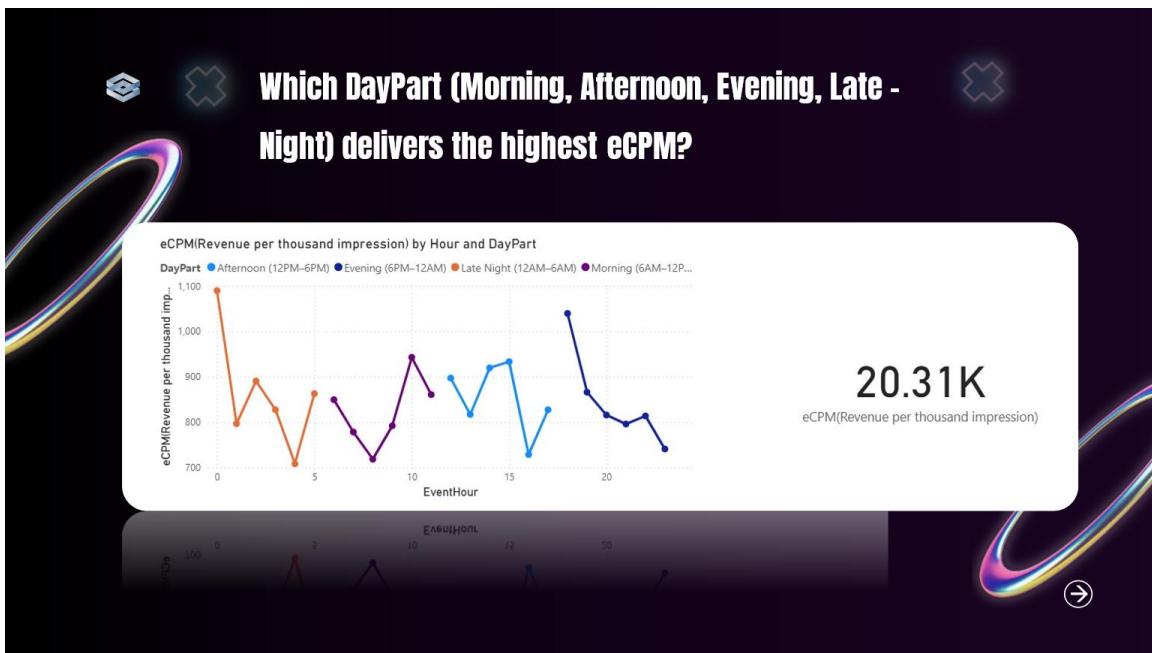
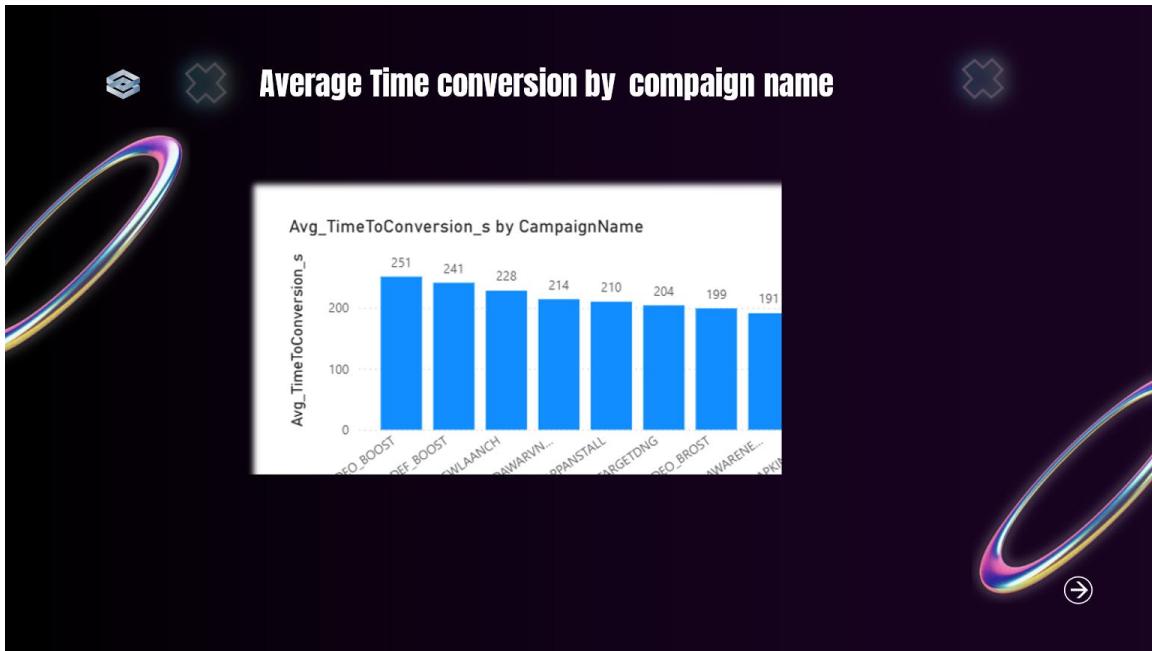
Is there a noticeable device preference trend for engagements (CTR) vs conversions (CVR)?



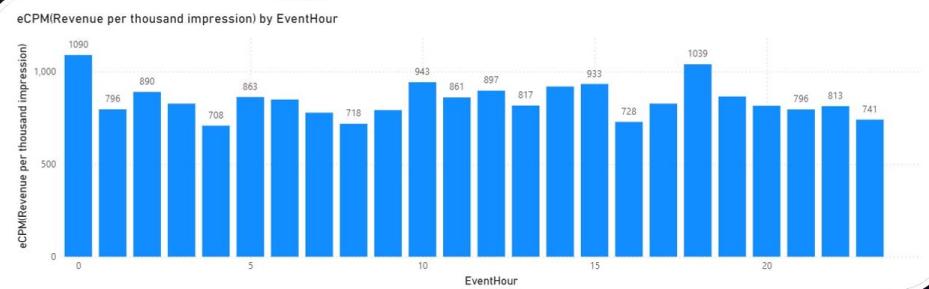
Which campaign generates the highest CTR OR CVR?







Does Morning (6AM -12PM) show noticeable dips in certain hours?



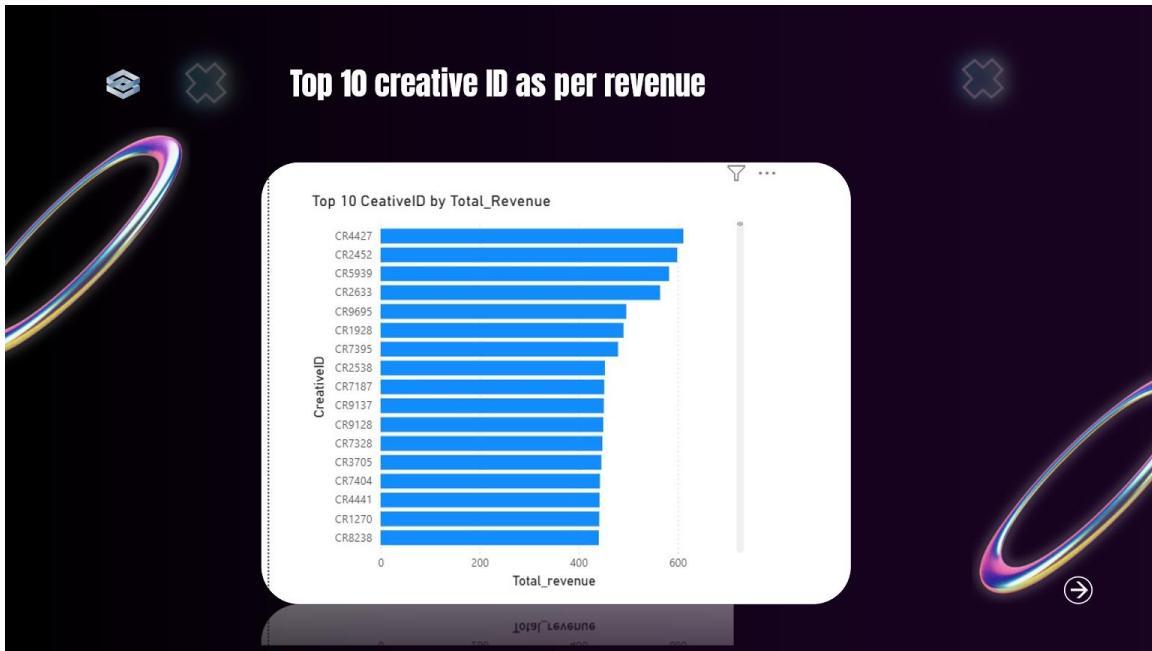
What is the percentage of frauds?

0.49250%

Fraud_Percentage

591

Total Fraud_entry



🎨 VISUAL WORKFLOW ILLUSTRATION

📁 CSV Files → 🐍 Python Loader → 🗂️ SQL Cleaned View → 📈 Power BI Dashboard

💡 BUSINESS RECOMMENDATIONS

- Google Ads delivered the highest ROAS (3.4x)
- LinkedIn campaigns showed 22% lower CTR, reduce spend on low-performing ads
- Peak engagement observed on evenings (7-12 PM)
- Need to work on Festival season [July- October]
- Cookingblog and moviehub have high difference in CTR vs CVR
- Recommend shifting at least 30% of creatives to short-form video