

# ADTECH ANALYTICS PROJECT

## (SQL + PYTHON + POWER BI)

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### 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 (Load CSV → SQL Server)



SQL Server (Staging & Cleaning using SQL Views)



Power BI (Modeling, DAX, EDA, Time Intelligence & Dashboard)

### 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_co
nnection=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,
```

```
--  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

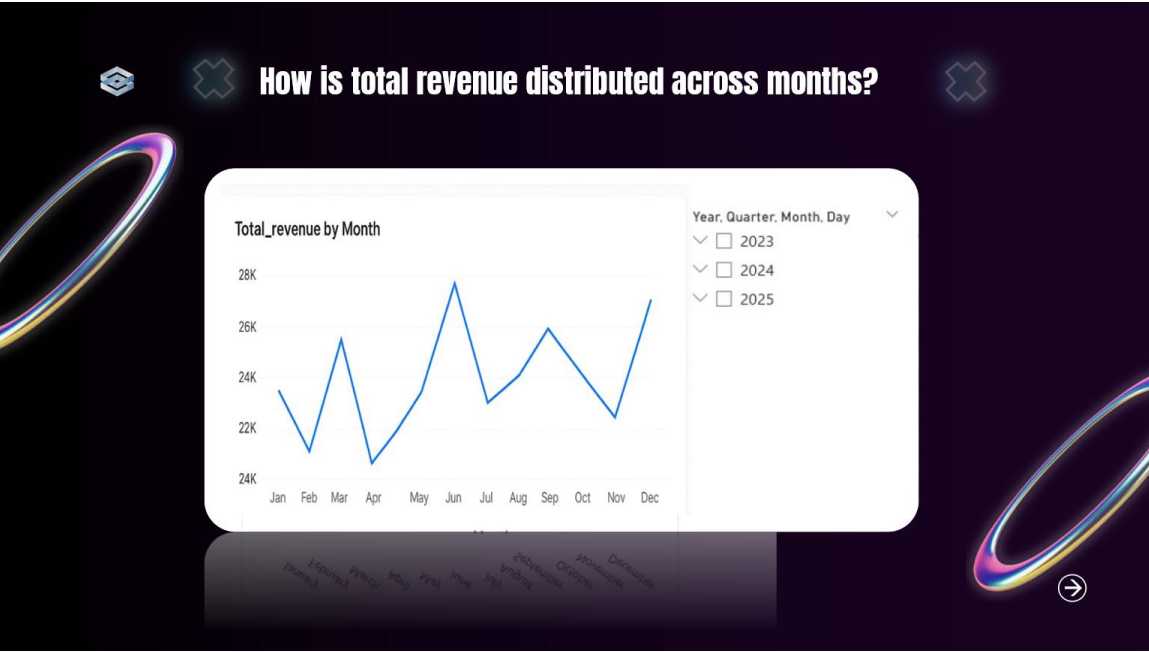
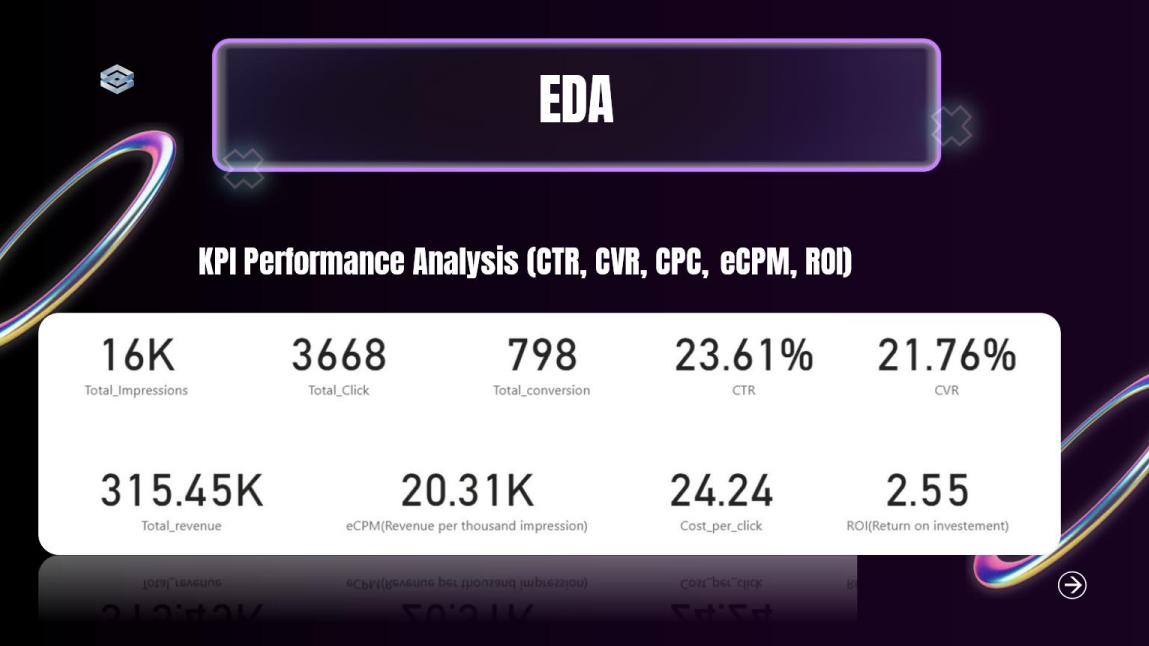
```
Fraud_Impressions =  
CALCULATE([Total_Impressions], vw_Adtech_Cleaned[IsFraud] = 1)
```

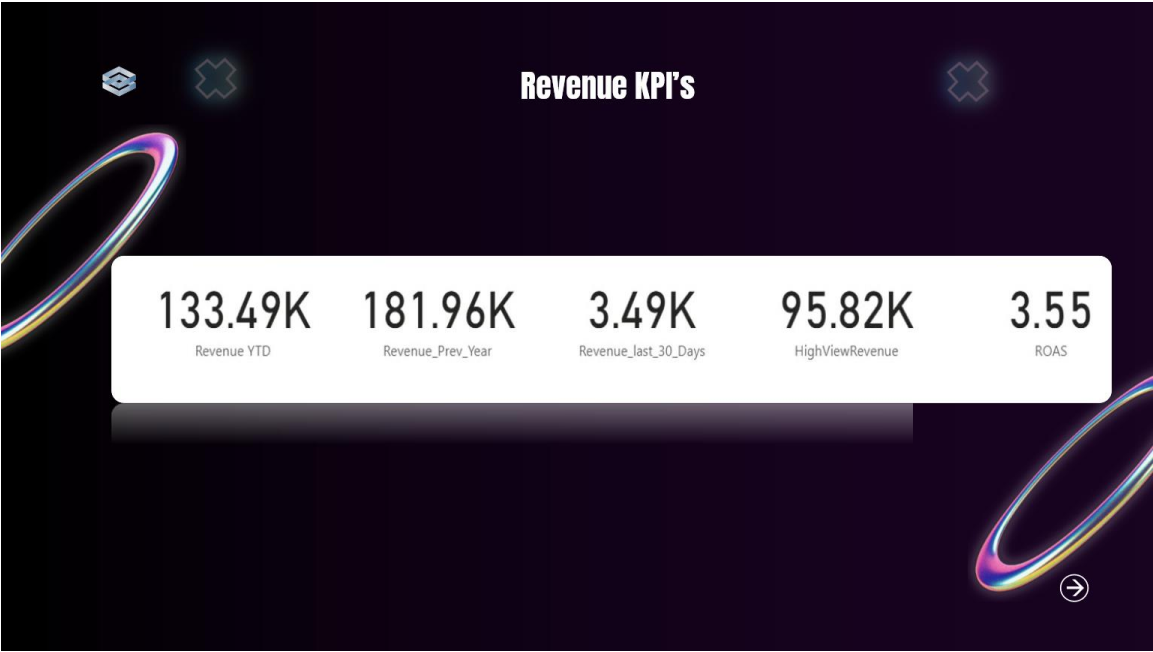
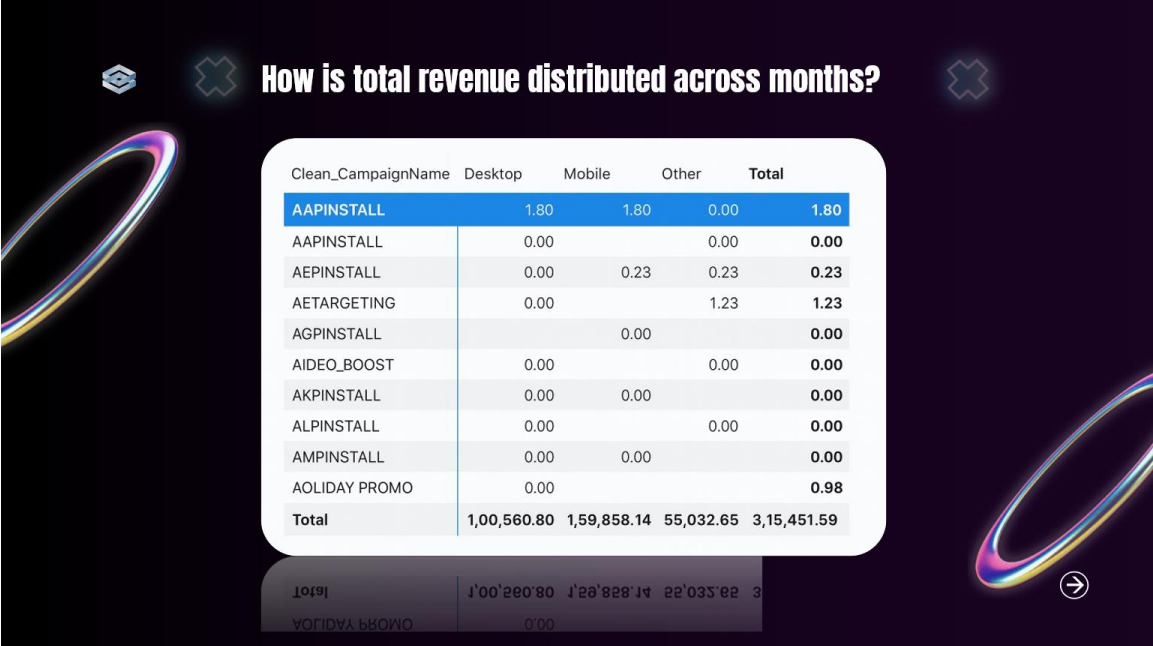
## Fraud\_percentage

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

## Day part

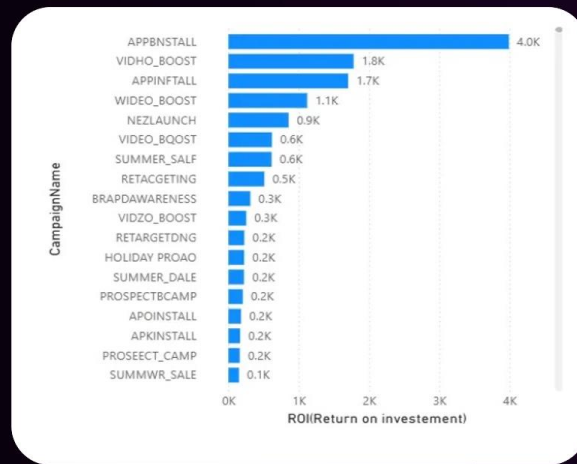
```
DayPart =  
  
SWITCH(  
  
    TRUE(),  
  
    vw_Adtech_Cleaned[EventHour] >= 0 && vw_Adtech_Cleaned[EventHour] <  
6, "Late Night (12AM-6AM)",  
  
    vw_Adtech_Cleaned[EventHour] >= 6 && vw_Adtech_Cleaned[EventHour] <  
12, "Morning (6AM-12PM)",  
  
    vw_Adtech_Cleaned[EventHour] >= 12 &&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







## Which campaigns have the highest ROI ?



## How is revenue distributed across 2023, 2024, and 2025?



Year	Cumulative Revenue
2025	3,15,451.59
2024	1,81,964.72
2023	25,887.27
Total	3,15,451.59





Which campaigns show zero or negative ROI & ROAS?



Clean_CampaignName	PublisherName	Cost	ROI	ROAS
APIINSTALL	NEWTODAY	0.61	-1.00	0.00
APIINSTALL	QOVIEHUB	1.11	-1.00	0.00
APIINSTALL	COOKINGBLOG	2.15	-1.00	0.00
APIINSTALL	FASHIONMAG	0.52	-1.00	0.00
APIINSTALL	COOKINGBLOG	1.40	-1.00	0.00
APIINSTALL	SPORTSDAILY	0.14	-1.00	0.00
APIINSTALL	COOKINGBLOG	0.33	-1.00	0.00
APIINSTALL	FASHIONMAG	1.25	-1.00	0.00
APIINSTALL	TECHFORUM	0.67	-1.00	0.00
APIINSTALL	SPORTSDAILY	0.17	-1.00	0.00
VIDEO_BOOST	MOVIEHUB	1.25	-1.00	0.00
APIINSTALL	FASHIONMAG	1.31	-1.00	0.00
APIINSTALL	SPORTSDAILY	1.21	-1.00	0.00
Total		88,925.63	2.55	3.55

Cost	88,925.63
ROI	2.55
ROAS	3.55



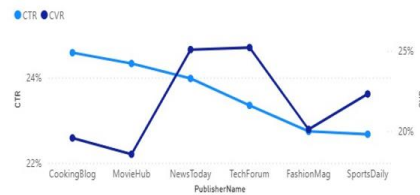
Is there a significant CTR-CVR imbalance across publishers?



23.61%  
CTR Percentage

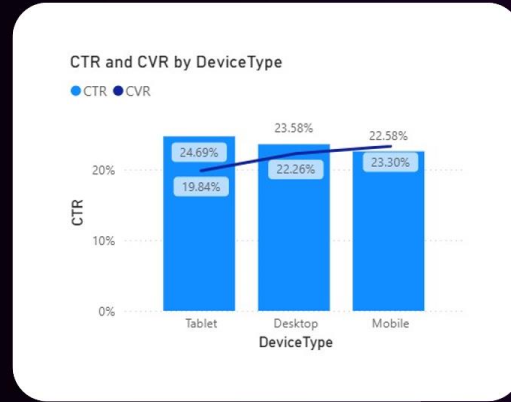
21.76%  
CVR Percentage

CTR and CVR by PublisherName

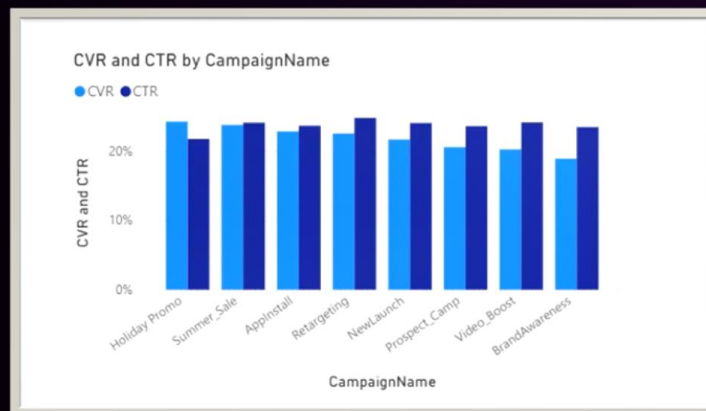




Is there a noticeable device preference trend for engagements (CTR) vs conversions (CVR)?



Which campaign generates the highest CTR OR CVR?



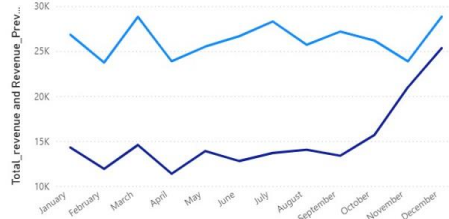


## Comparing the Revenue with Previous Year



Total\_revenue and Revenue\_Prev\_Year by Month

● Total\_revenue ● Revenue\_Prev\_Year



133.49K  
Revenue YTD

16K  
Impressions Last\_year

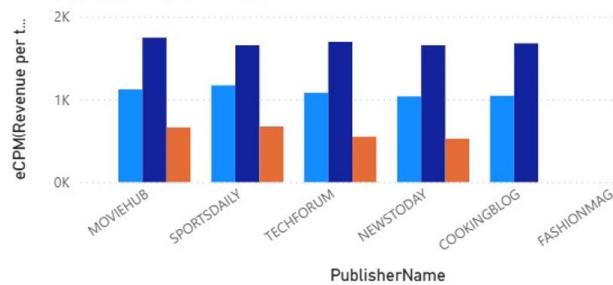


## Top 5 Ecpm By Publishername & Device Type



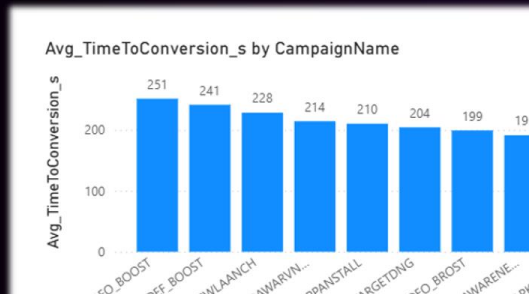
Top 5 eCPM by PublisherName and DeviceType

DeviceType ● Desktop ● Mobile ● Other

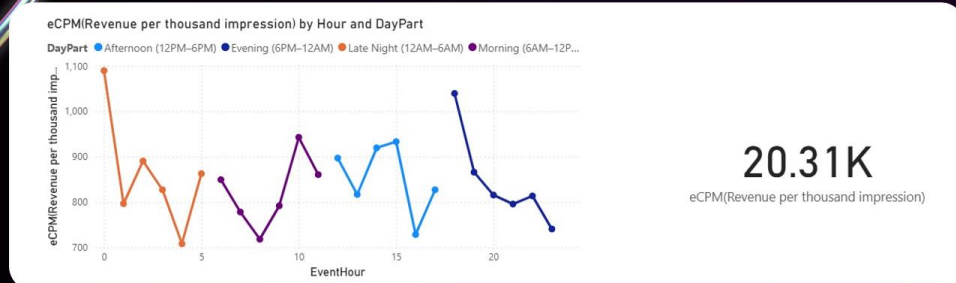




## Average Time conversion by campaign name



## Which DayPart (Morning, Afternoon, Evening, Late - Night) delivers the highest eCPM?



20.31K

eCPM(Revenue per thousand impression)







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



What is the percentage of frauds?



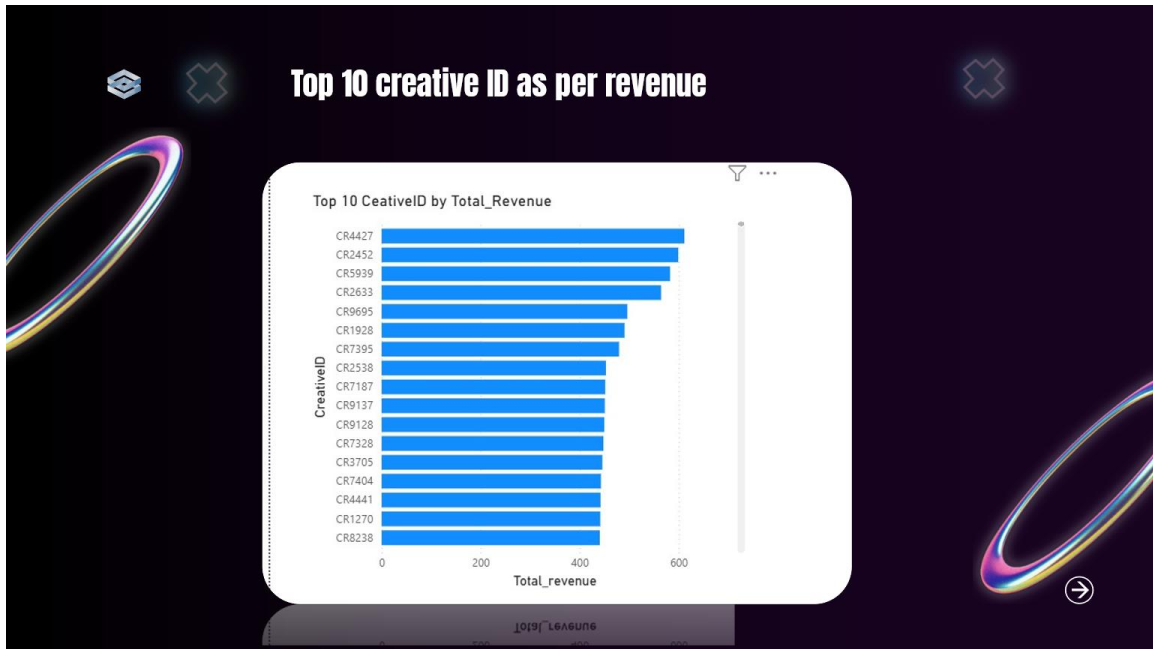
0.49250%

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