### "A/B TESTING INSIGHTS: SQL-BASED ANALYSIS OF MARKETING STRATEGIES"

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#### A/B Testing Intro

A/B testing is a simple yet powerful way to compare two versions of something—like marketing campaigns or product features—to see which performs better. By dividing users into two groups (A and B) and analyzing how they respond, we can make data-backed decisions that improve performance, engagement, and ultimately, business outcomes.

#### **Dataset Used**

We sourced our dataset from Kaggle: A/B Testing Marketing Campaign Dataset. The dataset includes two CSV files, each containing 30 rows of data. Our primary analysis was conducted on the file, having two campaign groups, control\_group.csv and test\_group.csv.

#### **Dataset Overview**

Re	esult Grid   III 🛟 Filter Rows: Export: III   Wrap Cell Content: IA   Fetch rows:									
	Campaign Name	Date	Spend [USD]	# of Impressions	Reach	# of Website Clicks	# of Searches	# of View Content	# of Add to Cart	# of Purchase
<b>)</b>	Control Campaign	1.08.2019	2280	82702	56930	7016	2290	2159	1819	618
	Control Campaign	2.08.2019	1757	121040	102513	8110	2033	1841	1219	511
	Control Campaign	3.08.2019	2343	131711	110862	6508	1737	1549	1134	372
	Control Campaign	4.08.2019	1940	72878	61235	3065	1042	982	1183	340
	Control Campaign	5.08.2019	1835							
	Control Campaign	6.08.2019	3083	109076	87998	4028	1709	1249	784	764
	Control Campaign	7.08.2019	2544	142123	127852	2640	1388	1106	1166	499
	Control Campaign	8.08.2019	1900	90939	65217	7260	3047	2746	930	462
	Control Campaign	9.08.2019	2813	121332	94896	6198	2487	2179	645	501
	Control Composion	10.09.2010	2140	117624	01257	2277	2475	1004	1620	724

#### Data Cleaning

Combining two campaign data to one by UNION ALL.

```
CREATE TABLE ab_campaign_data AS
SELECT
  'control' AS group_type,
  `Campaign Name`,
  `Date`,
  `Spend [USD]`,
  `# of Impressions`,
  `Reach`,
  `# of Website Clicks`,
  `# of Searches`,
  `# of View Content`,
  `# of Add to Cart`,
  `# of Purchase`
FROM control_group
UNION ALL
SELECT
  'test' AS group_type,
  `Campaign Name`,
  `Date`,
  `Spend [USD]`,
  `# of Impressions`,
  `Reach`,
  `# of Website Clicks`,
  `# of Searches`,
  `# of View Content`,
  `# of Add to Cart`,
  `# of Purchase`
FROM test_group;
```

Replacing Blank Spaces to O, for better calculation.

```
SELECT DISTINCT `# of Purchase`
FROM ab_campaign_data;
UPDATE ab_campaign_data
SET `# of Purchase` = O
WHERE TRIM(`# of Purchase`) = "OR `# of Purchase` IS NULL;
UPDATE ab_campaign_data
SET `# of Impressions` = 0
WHERE TRIM(`# of Impressions`) = " OR `# of Impressions` IS NULL;
UPDATE ab_campaign_data
SET `# of Website Clicks` = 0
WHERE TRIM(`# of Website Clicks`) = "OR `# of Website Clicks` IS NULL;
UPDATE ab_campaign_data
SET `# of Searches` = O
WHERE TRIM(`# of Searches`) = "OR `# of Searches` IS NULL;
UPDATE ab_campaign_data
SET `# of View Content` = 0
WHERE TRIM(`# of View Content`) = "OR `# of View Content` IS NULL;
UPDATE ab_campaign_data
SET `# of Add to Cart` = O
WHERE TRIM(`# of Add to Cart`) = " OR `# of Add to Cart` IS NULL;
UPDATE ab_campaign_data
SET `Reach` = O
WHERE TRIM(`Reach`) = " OR `Reach` IS NULL;
```

Redefining datatypes for all columns.

ALTER TABLE ab\_campaign\_data
MODIFY `# of Purchase` INT,
MODIFY `# of Impressions` INT,
MODIFY `# of Website Clicks` INT,
MODIFY `# of Searches` INT,
MODIFY `# of View Content` INT,
MODIFY `# of Add to Cart` INT;

ALTER TABLE ab\_campaign\_data MODIFY `Reach` INT;

ALTER TABLE ab\_campaign\_data MODIFY `Spend [USD]` FLOAT;

Rename columns to proper names.

```
CHANGE `Campaign_Name` campaign_name VARCHAR(255),
CHANGE `Date` campaign_date DATE,
CHANGE `Spend [USD]` spend_usd FLOAT,
CHANGE `# of Impressions` impressions INT,
CHANGE `Reach` reach INT,
CHANGE `# of Website Clicks` website_clicks INT,
CHANGE `# of Searches` searches INT,
CHANGE `# of View Content` view_content INT,
CHANGE `# of Add to Cart` add_to_cart INT,
CHANGE `# of Purchase` purchase INT;
```

#### Final Cleaned Dataset Overview

Result Grid Filter Rows:			Export: Wrap Cell Content: TA Fetch rows:			Fetch rows:					
	group_type	campaign_name	campaign_date	spend_usd	impressions	reach	website_clicks	searches	view_content	add_to_cart	purchase
<b>&gt;</b>	control	Control Campaign	2001-08-20	2280	82702	56930	7016	2290	2159	1819	618
	control	Control Campaign	2002-08-20	1757	121040	102513	8110	2033	1841	1219	511
	control	Control Campaign	2003-08-20	2343	131711	110862	6508	1737	1549	1134	372
	control	Control Campaign	2004-08-20	1940	72878	61235	3065	1042	982	1183	340
	control	Control Campaign	2005-08-20	1835	0	0	0	0	0	0	0
	control	Control Campaign	2006-08-20	3083	109076	87998	4028	1709	1249	784	764
	control	Control Campaign	2007-08-20	2544	142123	127852	2640	1388	1106	1166	499
	control	Control Campaign	2008-08-20	1900	90939	65217	7260	3047	2746	930	462
	control	Control Campaign	2009-08-20	2813	121332	94896	6198	2487	2179	645	501
	control	Control Campaign	2010-08-20	2149	117624	91257	2277	2475	1984	1629	734

## Calculating Metrics / Performing Action on Data

Campaign Metrics Aggregation: Summarize total spend, reach, clicks, and purchases by group.

```
-- Common Table Expression to prepare aggregated data
WITH campaign_metrics AS (
  SELECT
    group_type,
               COUNT(DISTINCT campaign_name) AS
total_campaigns,
    COUNT(*) AS total_days,
    SUM(spend_usd) AS total_spend,
    SUM(impressions) AS total_impressions,
    SUM(reach) AS total_reach,
    SUM(website_clicks) AS total_clicks,
    SUM(searches) AS total_searches,
    SUM(view_content) AS total_views,
    SUM(add_to_cart) AS total_carts,
    SUM(purchase) AS total_purchases,
     SUM(purchase) / NULLIF(SUM(website_clicks), O) AS
purchase_per_click,
      SUM(add_to_cart) / NULLIF(SUM(view_content), 0)
AS cart_per_view
  FROM ab_campaign_data
  GROUP BY group_type
```

**Conversion Rate:** 

Percentage of clicks that led to purchases.

Click-Through Rate (CTR):
Percentage of impressions that
turned into clicks.

Return on Ad Spend (ROAS): Total purchase value per dollar spent.

```
-- Metric 3: Return on Ad Spend (ROAS)

roas AS (
    SELECT
        group_type,
        ROUND(SUM(purchase) / NULLIF(SUM(spend_usd), 0),
2) AS roas
    FROM ab_campaign_data
    GROUP BY group_type
),
```

Cost Per Purchase (CPP):

Avg. amount spent per purchase.

```
-- Metric 4: Cost per Purchase (CPP)

cpp AS (
    SELECT
        group_type,
        ROUND(SUM(spend_usd) / NULLIF(SUM(purchase), 0),

2) AS cost_per_purchase
    FROM ab_campaign_data
    GROUP BY group_type
),
```

Engagement Rate:
Percentage of users engaging after seeing the ad.

Average Daily Spend: Total spend divided by campaign days.

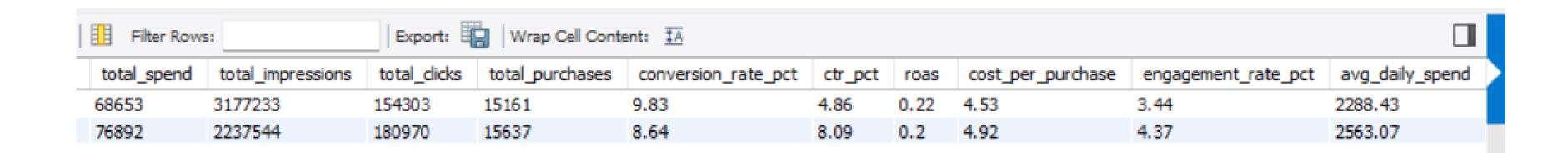
```
-- Metric 6: Daily Average Spend
daily_spend AS (
    SELECT
    group_type,
        ROUND(SUM(spend_usd) / COUNT(DISTINCT
campaign_date), 2) AS avg_daily_spend
    FROM ab_campaign_data
    GROUP BY group_type
),
```

Combined Output:

Merge all metrics for comparison between A/B groups using Inner Join.

```
-- Final Output: Combine All Metrics
SELECT
  cm.group_type,
  cm.total_campaigns,
  cm.total_days,
  cm.total_spend,
  cm.total_impressions,
  cm.total_clicks,
  cm.total_purchases,
  conv.conversion_rate_pct,
  ctr.ctr_pct,
  roas.roas,
  cpp.cost_per_purchase,
  engagement.engagement_rate_pct,
  daily_spend.avg_daily_spend
FROM campaign_metrics cm
JOIN conversion_rate conv ON cm.group_type =
conv.group_type
JOIN ctr ON cm.group_type = ctr.group_type
JOIN roas ON cm.group_type = roas.group_type
JOIN cpp ON cm.group_type = cpp.group_type
JOIN
         engagement
                         ON
                                 cm.group_type
engagement.group_type
JOIN
         daily_spend
                                 cm.group_type
                         ON
daily_spend.group_type;
```

#### Final Cleaned Output



#### Conclusion

After analyzing both campaigns, we observed that the Control Campaign outperformed the Test Campaign in several key areas. It recorded a higher conversion rate of 9.83% compared to 8.64%, and delivered a slightly better Return on Ad Spend (ROAS) of 0.22 against 0.20. Additionally, the cost per purchase was lower at \$4.53 versus \$4.92, making it more cost-efficient. On the other hand, the Test Campaign showed better engagement, with a Click-Through Rate (CTR) of 8.09% (vs 4.86%) and an engagement rate of 4.37% (vs 3.44%), along with a higher number of total clicks and purchases despite having fewer impressions. These findings suggest that while the Test Campaign succeeded in grabbing user attention, the Control Campaign was more effective in converting those interactions into meaningful purchases with better budget utilization. So, our final verdict is that the Control Campaign remains the more efficient and conversion-optimized strategy, making it the better choice for driving cost-effective results.

# Thankyou very much!