

```
In [3]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import pearsonr

df = pd.read_csv(r"D:\Doni Study\DataPortfolio\Forage_Simulations\Quantum-stimulation
```

```
In [5]: df.columns
```

```
Out[5]: Index(['LYLTY_CARD_NBR', 'DATE', 'STORE_NBR', 'TXN_ID', 'PROD_NBR',
               'PROD_NAME', 'PROD_QTY', 'TOT_SALES', 'PACK_SIZE', 'BRAND', 'LIFESTAGE',
               'PREMIUM_CUSTOMER'],
              dtype='object')
```

```
In [8]: #Convert 'DATE' to datetime
df['DATE'] = pd.to_datetime(df['DATE'])

#Create a 'month' column
df['month'] = df['DATE'].dt.to_period('M')

#Group by Store and Month
monthly_metrics = df.groupby(['STORE_NBR', 'month']).agg({
    'TOT_SALES': 'sum',
    'LYLTY_CARD_NBR': pd.Series.nunique,
    'TXN_ID': 'count'
}).reset_index()

#Rename Columns
monthly_metrics.rename(columns={
    'TOT_SALES': 'total_sales',
    'LYLTY_CARD_NBR': 'num_customers',
    'TXN_ID': 'num_transactions',
    'STORE_NBR': 'store_id'
}, inplace=True)

#Calculate Transactions per Customer
monthly_metrics['transactions_per_customer'] = (
    monthly_metrics['num_transactions'] / monthly_metrics['num_customers']
)

monthly_metrics.head()
```

Out[8]:

	store_id	month	total_sales	num_customers	num_transactions	transactions_per_customer
0	1	2018-07	206.9	49	52	1.0612
1	1	2018-08	176.1	42	43	1.0238
2	1	2018-09	278.8	59	62	1.0508
3	1	2018-10	188.1	44	45	1.0227
4	1	2018-11	192.6	46	47	1.0217



In [13]:

```
#Pre trial period
pre_trial = monthly_metrics[monthly_metrics['month'] < '2019-02']
```

In [14]:

```
#stores that appear in every month
store_month_counts = pre_trial.groupby('store_id')['month'].nunique()
full_period_stores = store_month_counts[store_month_counts == store_month_counts.ma
```

In [15]:

```
#Filter Monthly Metrics to These Stores
pre_trial_filtered = pre_trial[pre_trial['store_id'].isin(full_period_stores)]
```

In [19]:

```
from scipy.stats import pearsonr

def compare_store(trial_store_id, metric):
    trial_data = pre_trial_filtered[pre_trial_filtered['store_id'] == trial_store_id]
    scores = []

    for store_id in pre_trial_filtered['store_id'].unique():
        if store_id == trial_store_id:
            continue
        control_data = pre_trial_filtered[pre_trial_filtered['store_id'] == store_id]

        # Align lengths
        min_len = min(len(trial_data), len(control_data))
        trial_slice = trial_data[:min_len]
        control_slice = control_data[:min_len]

        # Skip if constant
        if trial_slice.nunique() <= 1 or control_slice.nunique() <= 1:
            continue

        score = pearsonr(trial_slice, control_slice)[0]
        scores.append((store_id, score))

    return sorted(scores, key=lambda x: -x[1])
```

```
sales_matches_77 = compare_store(77, 'total_sales')
sales_matches_77[:5] # Show top 5 matches
```

Out[19]:

```
[(np.int64(71), np.float64(0.9141059654434731)),
 (np.int64(233), np.float64(0.9037741879865313)),
 (np.int64(119), np.float64(0.8676644040124436)),
 (np.int64(17), np.float64(0.8426683602788767)),
 (np.int64(3), np.float64(0.8066436365226516))]

[(71, 0.914), (233, 0.904), (119, 0.868), (17, 0.843), (3, 0.807)]
```

Store 71 has the highest correlation with store 77 — so it's your best control store for this metric.

In [20]:

```
customer_matches_77 = compare_store(77, 'num_customers')
txn_matches_77 = compare_store(77, 'transactions_per_customer')

customer_matches_77[:5]
txn_matches_77[:5]
```

Out[20]:

```
[(np.int64(176), np.float64(0.8760526150292577)),
 (np.int64(144), np.float64(0.8561467798933686)),
 (np.int64(86), np.float64(0.8153345733102846)),
 (np.int64(118), np.float64(0.7981669375767875)),
 (np.int64(248), np.float64(0.7630272705995468))]
```

In [23]:

```
#Run for store 86

sales_matches_86 = compare_store(86, 'total_sales')
customer_matches_86 = compare_store(86, 'num_customers')
txn_matches_86 = compare_store(86, 'transactions_per_customer')

# View top matches
print("Sales:", sales_matches_86[:5], "\n")
print("Customers:", customer_matches_86[:5], "\n")
print("Transactions:", txn_matches_86[:5], "\n")
```

Sales: [(np.int64(155), np.float64(0.8778816901365498)), (np.int64(132), np.float64(0.8465165640013178)), (np.int64(240), np.float64(0.8250658384874218)), (np.int64(222), np.float64(0.7950752986301441)), (np.int64(109), np.float64(0.788299539592691))]

Customers: [(np.int64(155), np.float64(0.9428756256686044)), (np.int64(114), np.float64(0.8553390499905142)), (np.int64(260), np.float64(0.8465019549963757)), (np.int64(176), np.float64(0.7963798466198578)), (np.int64(109), np.float64(0.7707780469374135))]

Transactions: [(np.int64(118), np.float64(0.8912594724823315)), (np.int64(13), np.float64(0.8848210815915369)), (np.int64(244), np.float64(0.8696464278356433)), (np.int64(144), np.float64(0.837250983917323)), (np.int64(22), np.float64(0.8335782263008245))]

In [26]:

```
#Run for store 88

sales_matches_88 = compare_store(88, 'total_sales')
customer_matches_88 = compare_store(88, 'num_customers')
```

```

txn_matches_88 = compare_store(88, 'transactions_per_customer')

# View top matches
print("Sales:", sales_matches_88[:5], "\n")
print("Customers:", customer_matches_88[:5], "\n")
print("Transactions:", txn_matches_88[:5], "\n")

Sales: [(np.int64(159), np.float64(0.9031855640409738)), (np.int64(204), np.float64(0.885774218439864)), (np.int64(134), np.float64(0.8642934799118711)), (np.int64(1), np.float64(0.813636047784386)), (np.int64(253), np.float64(0.8118377088858145))]

Customers: [(np.int64(237), np.float64(0.9473262390629102)), (np.int64(14), np.float64(0.9429761679659832)), (np.int64(178), np.float64(0.939466041693001)), (np.int64(35), np.float64(0.8995936151600288)), (np.int64(113), np.float64(0.8626323509003309))]

Transactions: [(np.int64(24), np.float64(0.8389890268345358)), (np.int64(137), np.float64(0.828816292707782)), (np.int64(272), np.float64(0.8027016967997676)), (np.int64(243), np.float64(0.7630435433711259)), (np.int64(201), np.float64(0.7607184418637815))]

```

Methodology: Control stores were selected based on similarity to trial stores across three key metrics:

- Monthly total sales revenue
- Monthly number of customers
- Monthly transactions per customer

Pearson correlation was used to compare each trial store to all other stores that were operational throughout the pre-trial period (before Feb 2019).

Final Control Store Matches:

- Trial Store 77 → Control Store 71
- Trial Store 86 → Control Store 155
- Trial Store 88 → Control Store 237

These stores showed the highest similarity across multiple metrics, ensuring reliable baseline comparisons.

```

In [28]: #Extract and Label store Data
def get_store_metrics(store_id, label):
    df = pre_trial_filtered[pre_trial_filtered['store_id'] == store_id].copy()
    df['store_type'] = label
    return df

# Trial vs Control pairs
store_77 = get_store_metrics(77, 'Trial 77')
control_71 = get_store_metrics(71, 'Control 71')

store_86 = get_store_metrics(86, 'Trial 86')
control_155 = get_store_metrics(155, 'Control 155')

```

```
store_88 = get_store_metrics(88, 'Trial 88')
control_237 = get_store_metrics(237, 'Control 237')
```

In [29]:

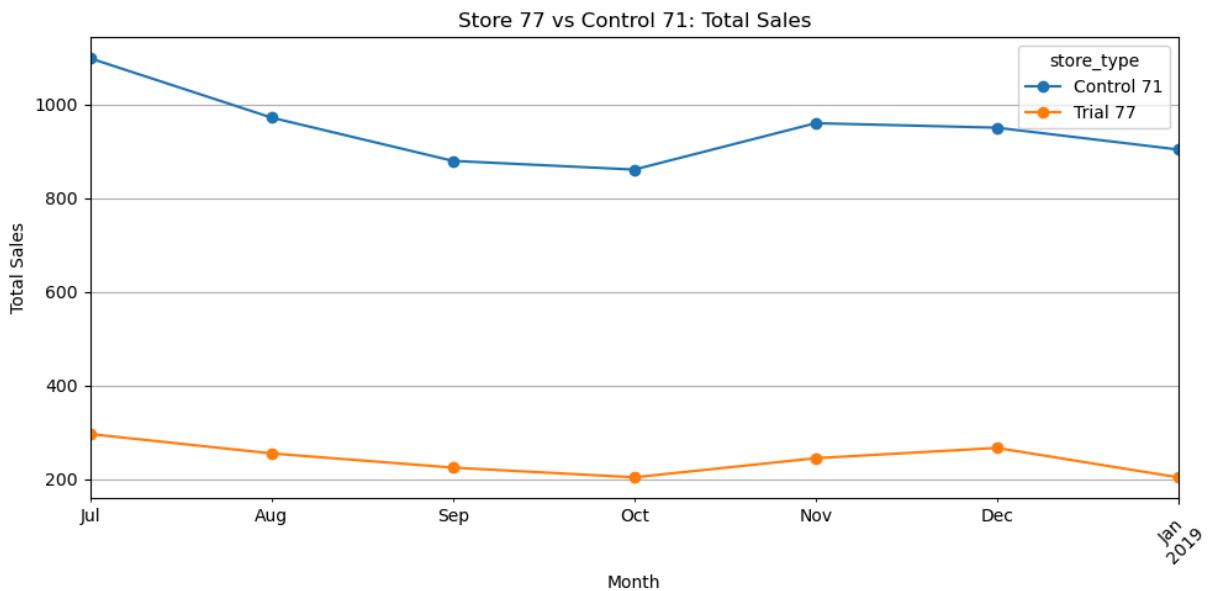
```
#Define Plot Function
import matplotlib.pyplot as plt

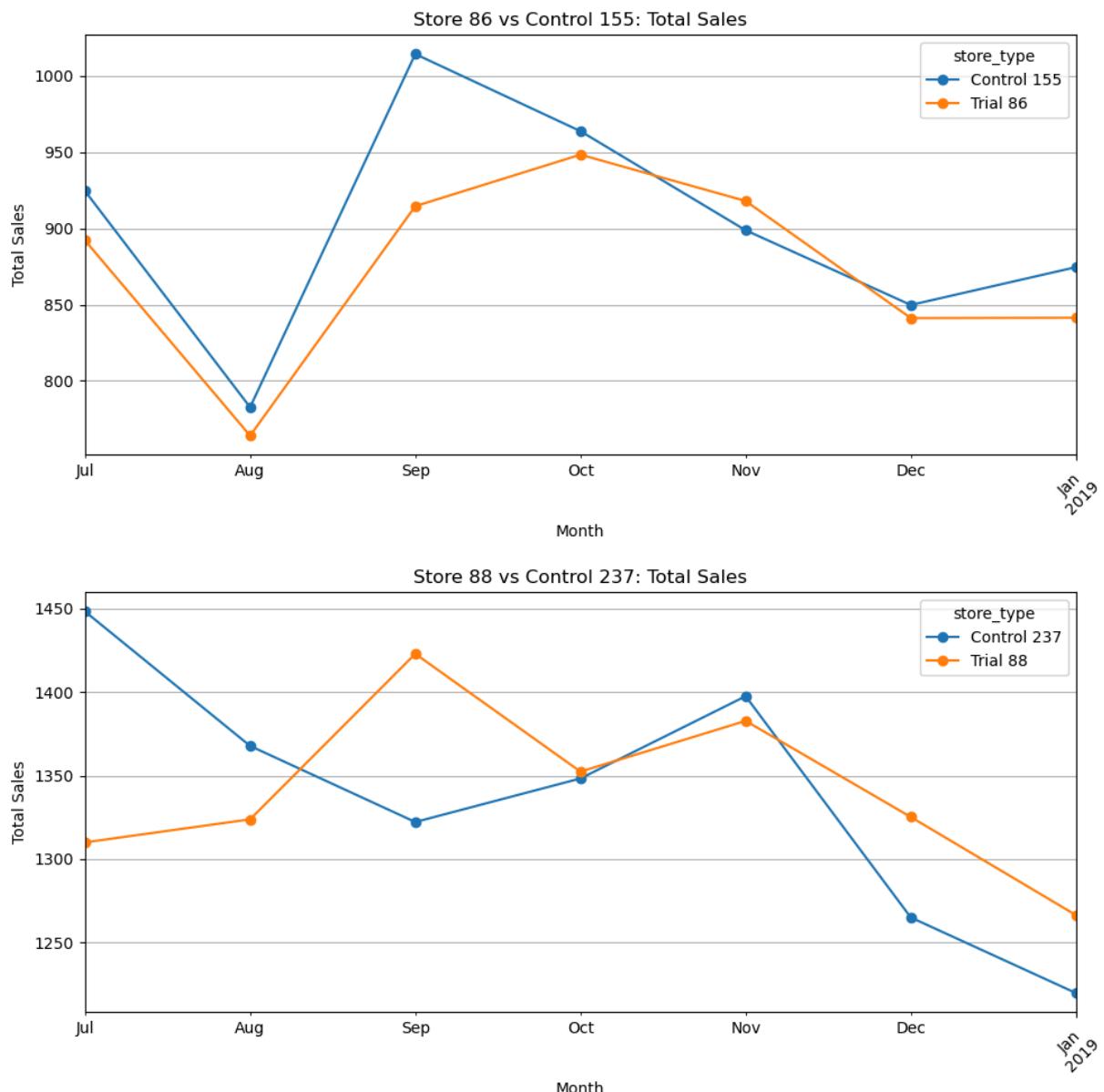
def plot_metric(trial_df, control_df, metric, title):
    combined = pd.concat([trial_df, control_df])
    pivot = combined.pivot(index='month', columns='store_type', values=metric)

    pivot.plot(figsize=(10, 5), marker='o')
    plt.title(title)
    plt.ylabel(metric.replace('_', ' ').title())
    plt.xlabel('Month')
    plt.grid(True)
    plt.xticks(rotation=45)
    plt.tight_layout()
    plt.show()
```

In [30]:

```
#Plot Total sales
plot_metric(store_77, control_71, 'total_sales', 'Store 77 vs Control 71: Total Sales')
plot_metric(store_86, control_155, 'total_sales', 'Store 86 vs Control 155: Total Sales')
plot_metric(store_88, control_237, 'total_sales', 'Store 88 vs Control 237: Total Sales')
```

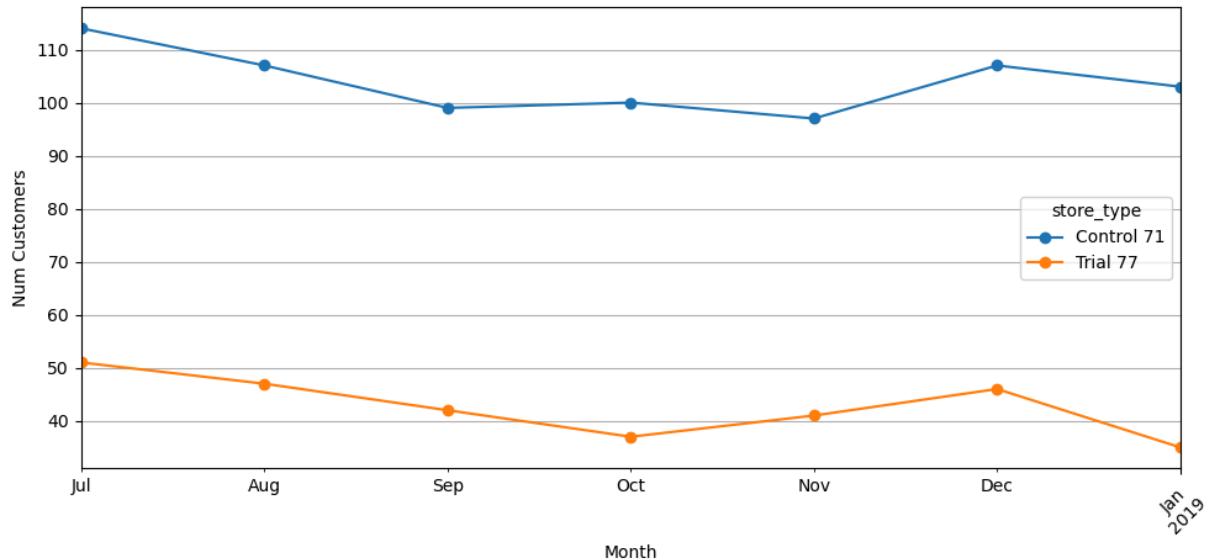




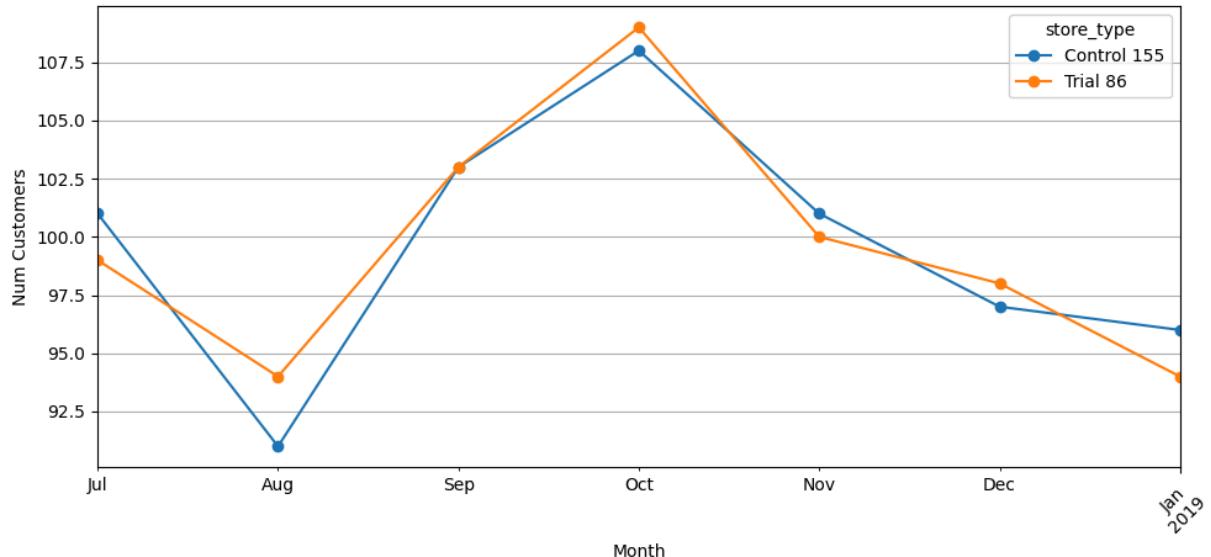
In [31]:

```
#Plot num_customers
plot_metric(store_77, control_71, 'num_customers', 'Store 77 vs Control 71: Total S
plot_metric(store_86, control_155, 'num_customers', 'Store 86 vs Control 155: Total
plot_metric(store_88, control_237, 'num_customers', 'Store 88 vs Control 237: Total
```

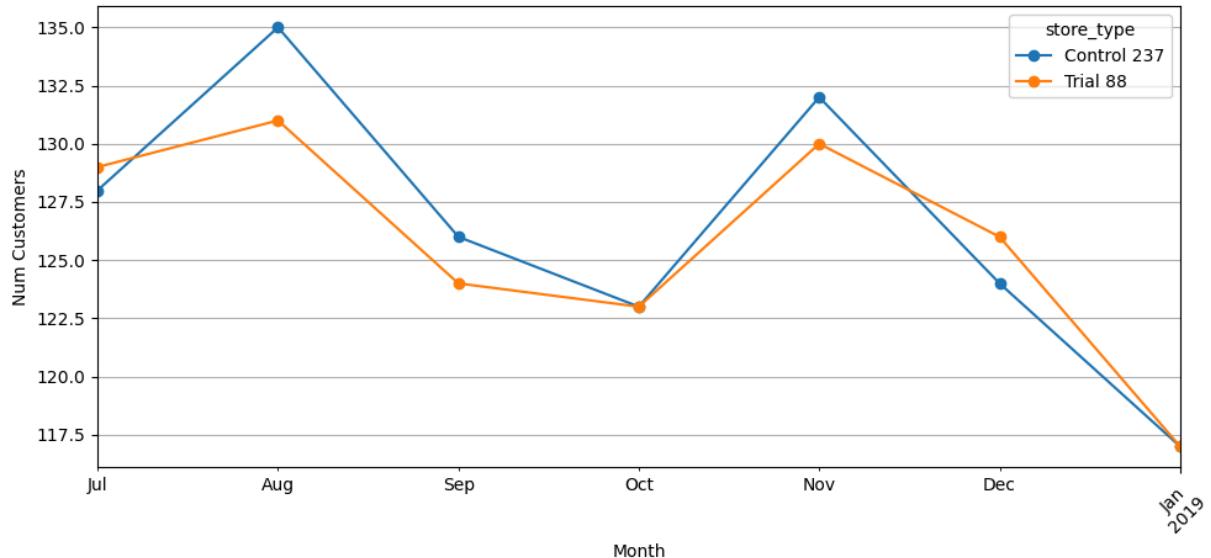
Store 77 vs Control 71: Total Sales



Store 86 vs Control 155: Total Sales

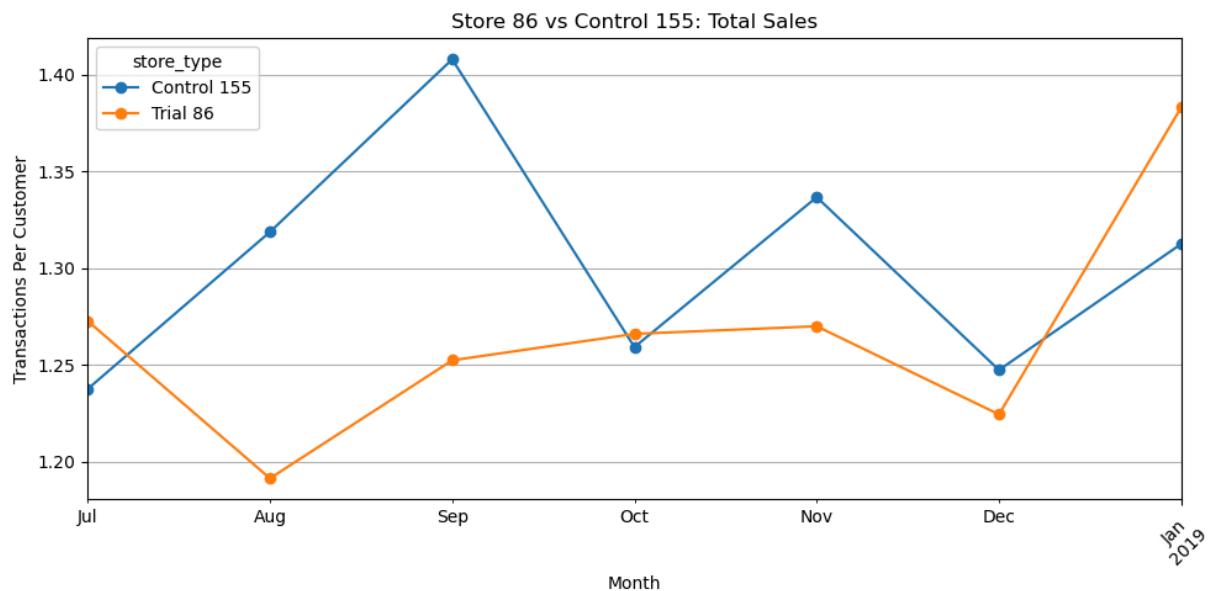
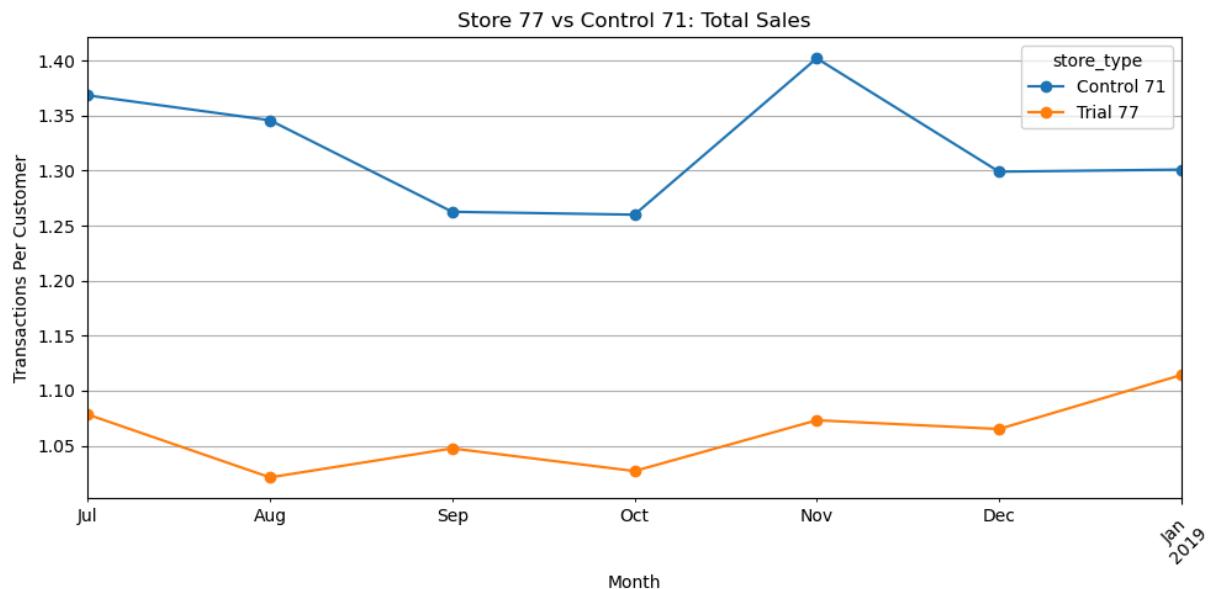


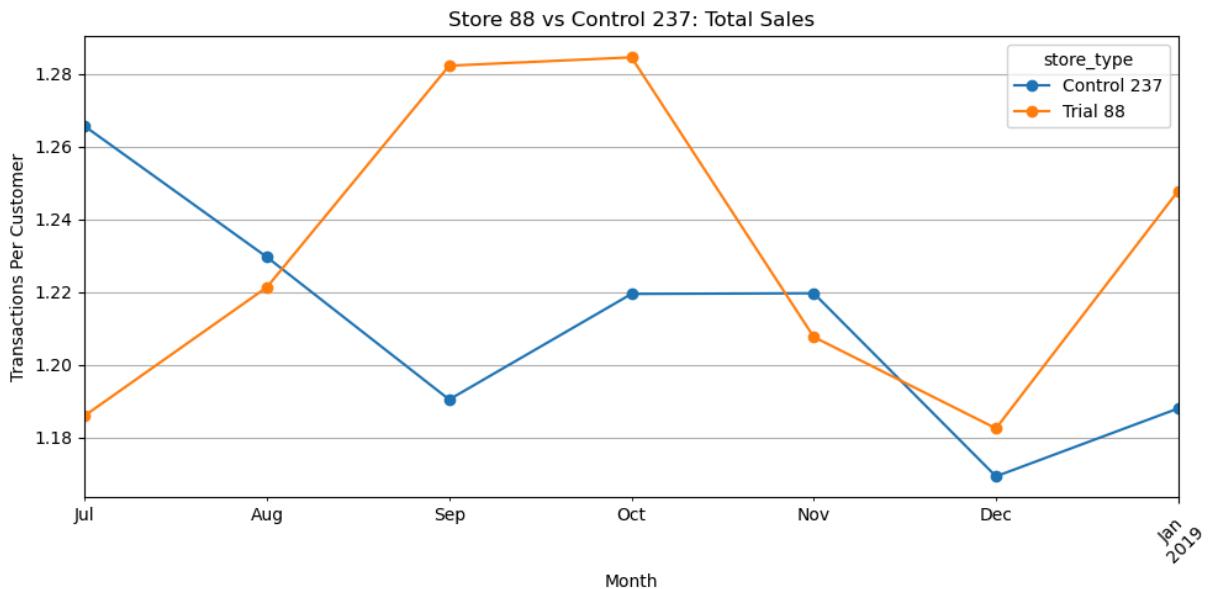
Store 88 vs Control 237: Total Sales



In [32]:

```
#Plot Total Sales
plot_metric(store_77, control_71, 'transactions_per_customer', 'Store 77 vs Control 71')
plot_metric(store_86, control_155, 'transactions_per_customer', 'Store 86 vs Control 155')
plot_metric(store_88, control_237, 'transactions_per_customer', 'Store 88 vs Control 237')
```





```
In [40]: #define trial period
trial_period = (monthly_metrics['month'] >= '2019-02') & (monthly_metrics['month'] <= '2019-06')

#Extract Trial and Control Data
#store 77:
trial_77 = monthly_metrics[(monthly_metrics['store_id'] == 77) & trial_period].copy()
trial_77['store_type'] = 'Trial 77'

control_71 = monthly_metrics[(monthly_metrics['store_id'] == 71) & trial_period].copy()
control_71['store_type'] = 'Control 71'
```

```
In [42]: #store 86:
trial_86 = monthly_metrics[(monthly_metrics['store_id'] == 86) & trial_period].copy()
trial_86['store_type'] = 'Trial 86'

control_155 = monthly_metrics[(monthly_metrics['store_id'] == 155) & trial_period].copy()
control_155['store_type'] = 'Control 155'
```

```
In [43]: #store 88:
trial_88 = monthly_metrics[(monthly_metrics['store_id'] == 88) & trial_period].copy()
trial_88['store_type'] = 'Trial 88'

control_237 = monthly_metrics[(monthly_metrics['store_id'] == 237) & trial_period].copy()
control_237['store_type'] = 'Control 237'
```

```
In [50]: #Trial period
trial_period = (monthly_metrics['month'] >= '2019-02') & (monthly_metrics['month'] <= '2019-06')

#Extract Trial and control data
#Store 77:
trial_77 = monthly_metrics[(monthly_metrics['store_id'] == 77) & trial_period].copy()
control_71 = monthly_metrics[(monthly_metrics['store_id'] == 71) & trial_period].copy()
trial_77['store_type'] = 'Trial 77'
control_71['store_type'] = 'Control 71'

#Store 86:
```

```

trial_86 = monthly_metrics[(monthly_metrics['store_id'] == 86) & trial_period].copy
control_155 = monthly_metrics[(monthly_metrics['store_id'] == 155) & trial_period].copy
trial_86['store_type'] = 'Trial 86'
control_155['store_type'] = 'Control 155'

#Store 88:
trial_88 = monthly_metrics[(monthly_metrics['store_id'] == 88) & trial_period].copy
control_237 = monthly_metrics[(monthly_metrics['store_id'] == 237) & trial_period].copy
trial_88['store_type'] = 'Trial 88'
control_237['store_type'] = 'Control 237'

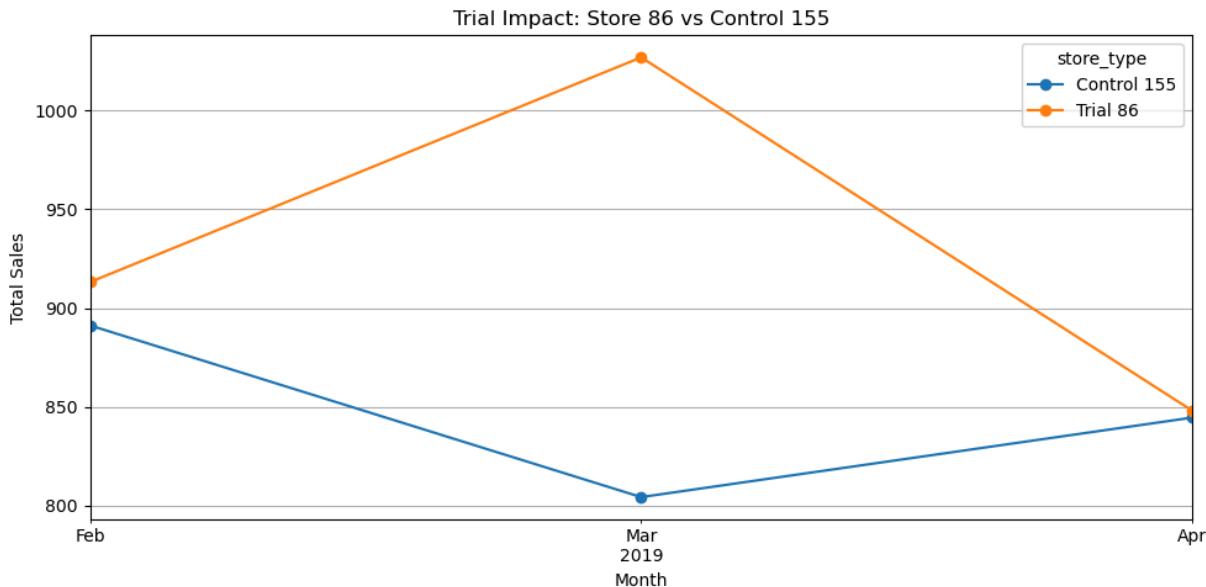
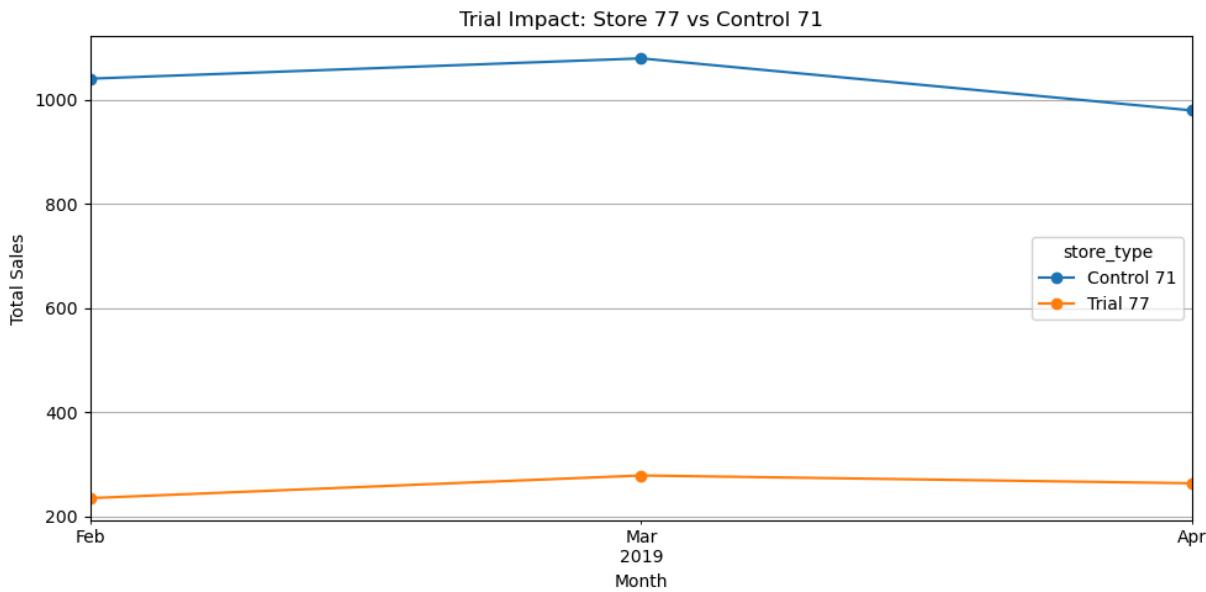
```

In [51]:

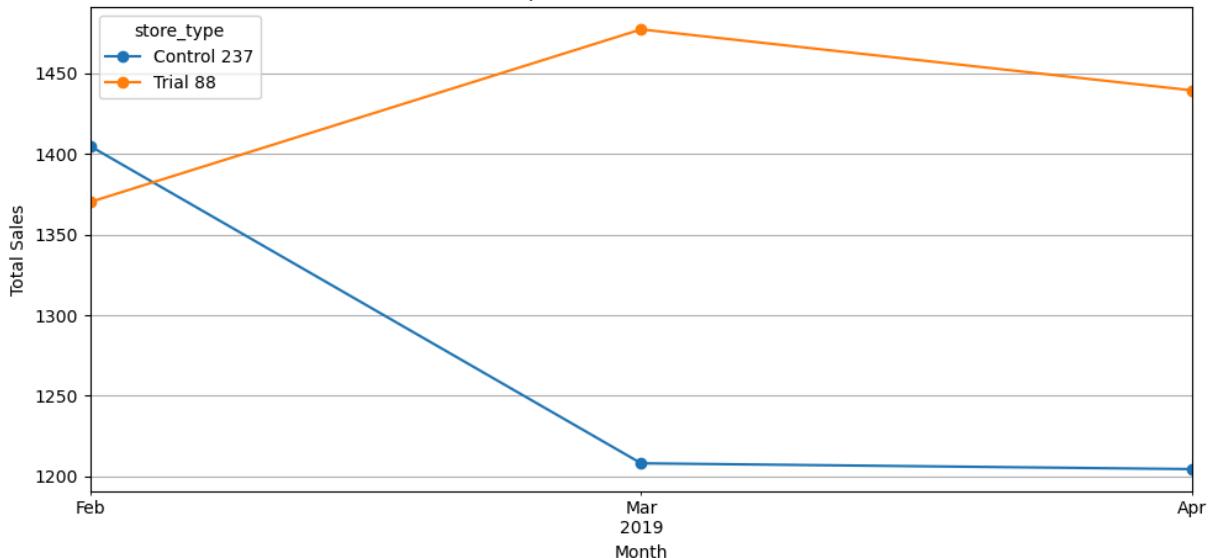
```

#Plot Trial vs Control During Trial
#Using Total sales
plot_metric(trial_77, control_71, 'total_sales', 'Trial Impact: Store 77 vs Control 71')
plot_metric(trial_86, control_155, 'total_sales', 'Trial Impact: Store 86 vs Control 155')
plot_metric(trial_88, control_237, 'total_sales', 'Trial Impact: Store 88 vs Control 237')

```



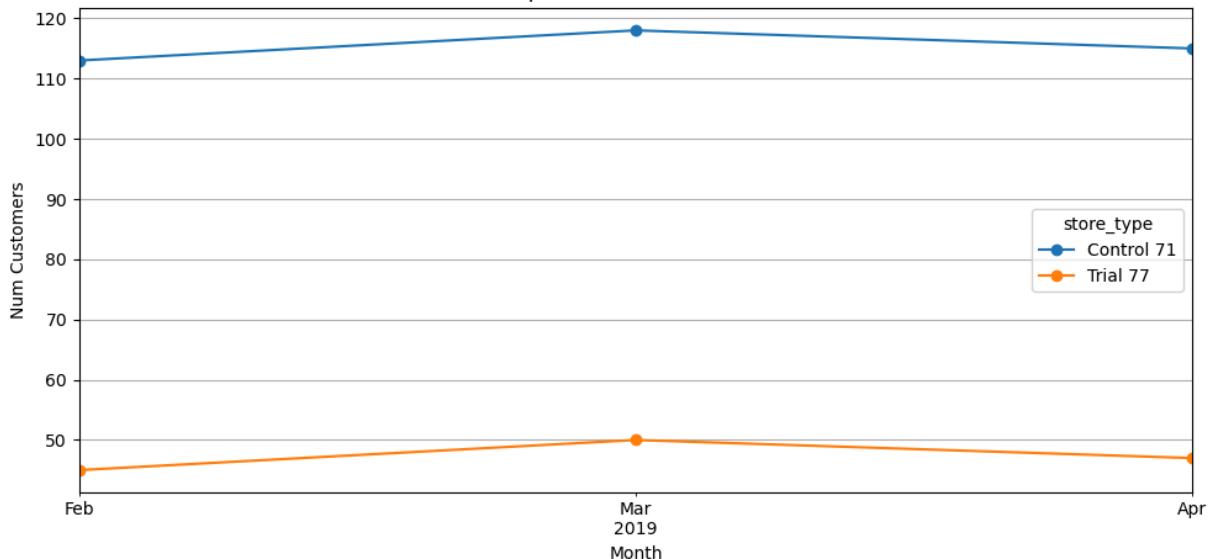
Trial Impact: Store 88 vs Control 237

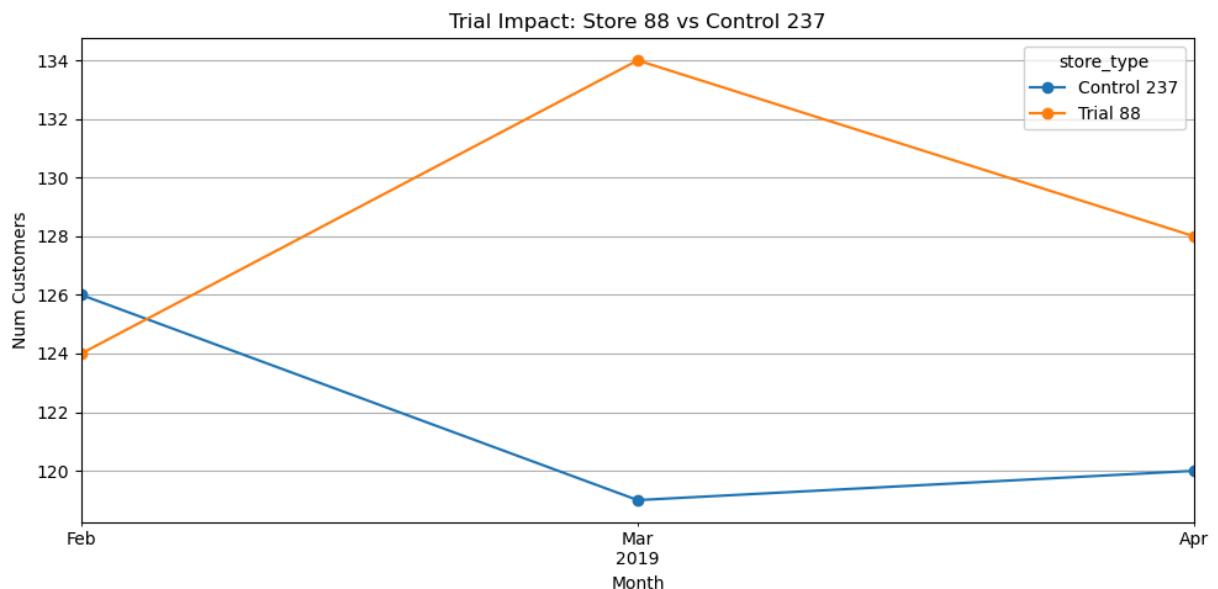
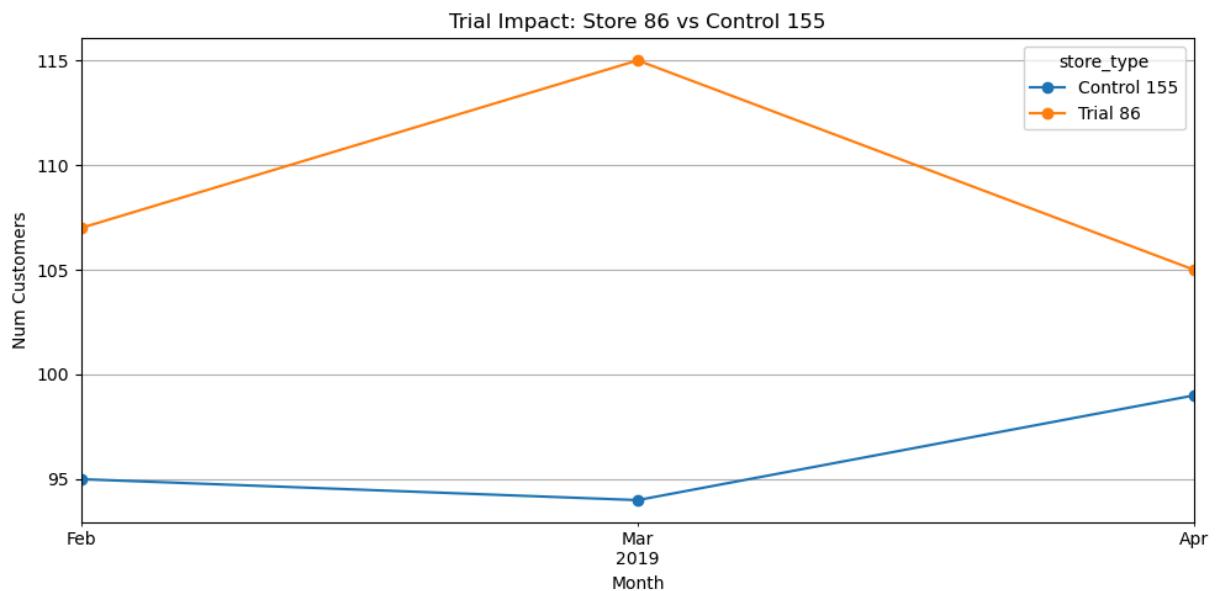


In [52]: #Using nuber of customers

```
plot_metric(trial_77, control_71, 'num_customers', 'Trial Impact: Store 77 vs Contr
plot_metric(trial_86, control_155, 'num_customers', 'Trial Impact: Store 86 vs Cont
plot_metric(trial_88, control_237, 'num_customers', 'Trial Impact: Store 88 vs Cont
```

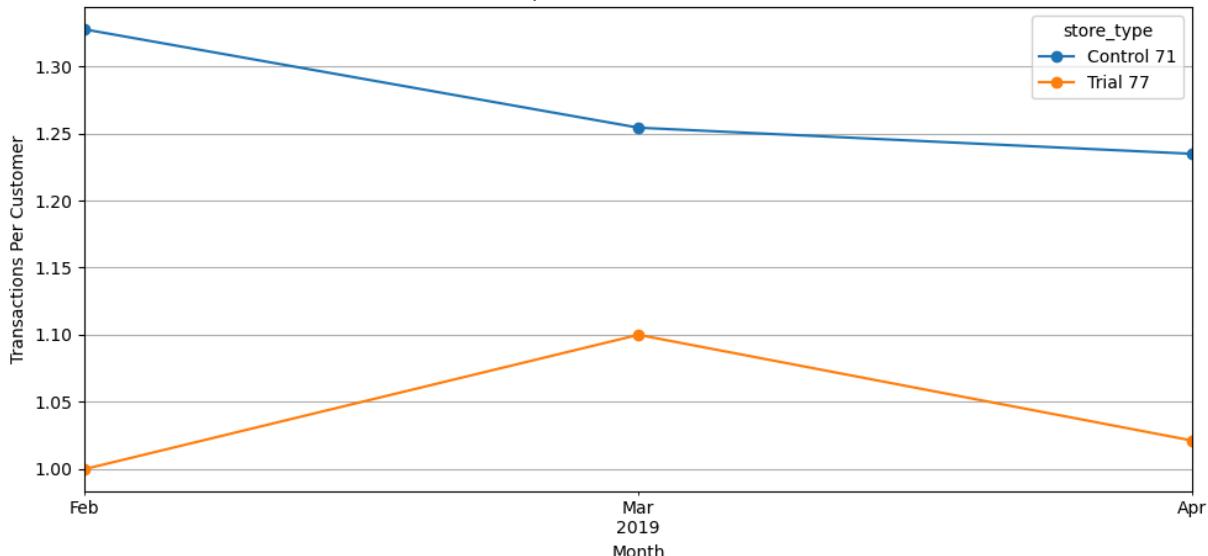
Trial Impact: Store 77 vs Control 71



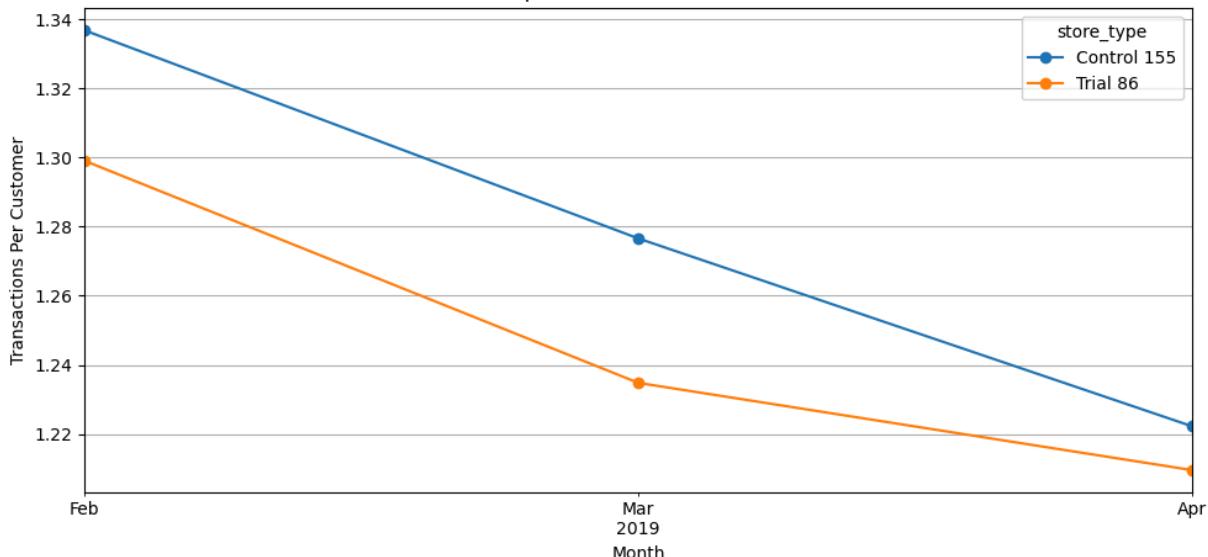


```
In [54]: #using transactions of customers
plot_metric(trial_77, control_71, 'transactions_per_customer', 'Trial Impact: Store 77 vs Control 71')
plot_metric(trial_86, control_155, 'transactions_per_customer', 'Trial Impact: Store 86 vs Control 155')
plot_metric(trial_88, control_237, 'transactions_per_customer', 'Trial Impact: Store 88 vs Control 237')
```

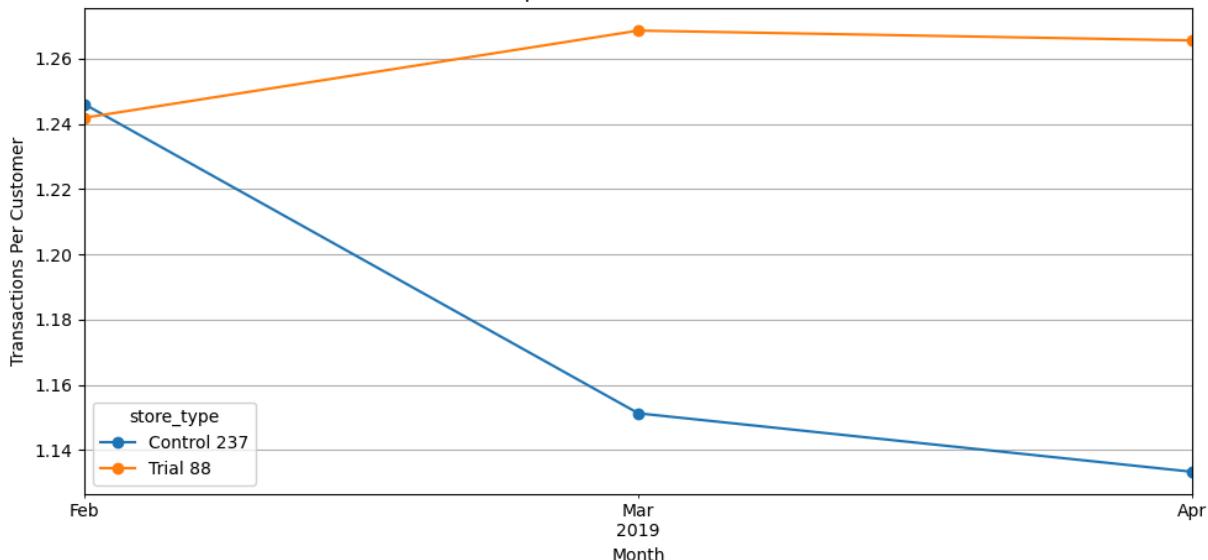
Trial Impact: Store 77 vs Control 71



Trial Impact: Store 86 vs Control 155



Trial Impact: Store 88 vs Control 237



```
In [55]: def calculate_lift(trial_df, control_df, metric):
    trial_total = trial_df[metric].sum()
    control_total = control_df[metric].sum()
    return ((trial_total - control_total) / control_total) * 100

lift_77 = calculate_lift(trial_77, control_71, 'total_sales')
lift_86 = calculate_lift(trial_86, control_155, 'total_sales')
lift_88 = calculate_lift(trial_88, control_237, 'total_sales')

print("Lift for Store 77:", round(lift_77, 2), "%")
print("Lift for Store 86:", round(lift_86, 2), "%")
print("Lift for Store 88:", round(lift_88, 2), "%")
```

Lift for Store 77: -74.94 %

Lift for Store 86: 9.76 %

Lift for Store 88: 12.29 %