

```

# IMPORTANT: RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES
# TO THE CORRECT LOCATION (/kaggle/input) IN YOUR NOTEBOOK,
# THEN FEEL FREE TO DELETE THIS CELL.
# NOTE: THIS NOTEBOOK ENVIRONMENT DIFFERS FROM KAGGLE'S PYTHON
# ENVIRONMENT SO THERE MAY BE MISSING LIBRARIES USED BY YOUR
# NOTEBOOK.

import os
import sys
from tempfile import NamedTemporaryFile
from urllib.request import urlopen
from urllib.parse import unquote, urlparse
from urllib.error import HTTPError
from zipfile import ZipFile
import tarfile
import shutil

CHUNK_SIZE = 40960
DATA_SOURCE_MAPPING = 'chips-customer-analysis-plan-forage:https%3A%2F%2Fstorage.googleapis.com%2Fkaggle-data-sets%2F5818142'

KAGGLE_INPUT_PATH='/kaggle/input'
KAGGLE_WORKING_PATH='/kaggle/working'
KAGGLE_SYMLINK='kaggle'

!umount /kaggle/input/ 2> /dev/null
shutil.rmtree('/kaggle/input', ignore_errors=True)
os.makedirs(KAGGLE_INPUT_PATH, 0o777, exist_ok=True)
os.makedirs(KAGGLE_WORKING_PATH, 0o777, exist_ok=True)

try:
    os.symlink(KAGGLE_INPUT_PATH, os.path.join(".", 'input'), target_is_directory=True)
except FileExistsError:
    pass
try:
    os.symlink(KAGGLE_WORKING_PATH, os.path.join(".", 'working'), target_is_directory=True)
except FileExistsError:
    pass

for data_source_mapping in DATA_SOURCE_MAPPING.split(','):
    directory, download_url_encoded = data_source_mapping.split(':')
    download_url = unquote(download_url_encoded)
    filename = urlparse(download_url).path
    destination_path = os.path.join(KAGGLE_INPUT_PATH, directory)
    try:
        with urlopen(download_url) as fileres, NamedTemporaryFile() as tfile:
            total_length = fileres.headers['content-length']
            print(f'Downloading {directory}, {total_length} bytes compressed')
            dl = 0
            data = fileres.read(CHUNK_SIZE)
            while len(data) > 0:
                dl += len(data)
                tfile.write(data)
                done = int(50 * dl / int(total_length))
                sys.stdout.write(f"\r[{ '=' * done }{' ' * (50 - done)}] {dl} bytes downloaded")
                sys.stdout.flush()
                data = fileres.read(CHUNK_SIZE)
            if filename.endswith('.zip'):
                with ZipFile(tfile) as zfile:
                    zfile.extractall(destination_path)
            else:
                with tarfile.open(tfile.name) as tarfile:
                    tarfile.extractall(destination_path)
            print(f'\nDownloaded and uncompressed: {directory}')
    except HTTPError as e:
        print(f'Failed to load (likely expired) {download_url} to path {destination_path}')
        continue
    except OSError as e:
        print(f'Failed to load {download_url} to path {destination_path}')
        continue

print('Data source import complete.')
```

```

🔄 Downloading chips-customer-analysis-plan-forage, 2668722 bytes compressed
[=====] 2668722 bytes downloaded
Downloaded and uncompressed: chips-customer-analysis-plan-forage
Downloading chips-customer-analysis-plan-forage-trailcontrol, 3080231 bytes compressed
[=====] 3080231 bytes downloaded
Downloaded and uncompressed: chips-customer-analysis-plan-forage-trailcontrol
Data source import complete.
```

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load

import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import pearsonr
from scipy.spatial.distance import euclidean

# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session

↵ /kaggle/input/chips-customer-analysis-plan-forage/QVI_purchase_behaviour.csv
/kaggle/input/chips-customer-analysis-plan-forage/QVI_transaction_data.csv
/kaggle/input/chips-customer-analysis-plan-forage-trailcontrol/QVI_data.csv

# Load datasets
qvi_data = pd.read_csv('/kaggle/input/chips-customer-analysis-plan-forage-trailcontrol/QVI_data.csv')
```

```
qvi_data.info()
```

```
↵ <class 'pandas.core.frame.DataFrame'>
RangeIndex: 264834 entries, 0 to 264833
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   LYLTY_CARD_NBR        264834 non-null int64
1   DATE                  264834 non-null object
2   STORE_NBR             264834 non-null int64
3   TXN_ID                264834 non-null int64
4   PROD_NBR              264834 non-null int64
5   PROD_NAME             264834 non-null object
6   PROD_QTY              264834 non-null int64
7   TOT_SALES             264834 non-null float64
8   PACK_SIZE             264834 non-null int64
9   BRAND                 264834 non-null object
10  LIFESTAGE              264834 non-null object
11  PREMIUM_CUSTOMER      264834 non-null object
dtypes: float64(1), int64(6), object(5)
memory usage: 24.2+ MB
```

```
# Displayng the first few rows of dataset
```

```
qvi_data_head = qvi_data.head()
```

```
qvi_data_shape = qvi_data.shape
```

```
qvi_data_head, qvi_data_shape
```

```
↵ (  LYLTY_CARD_NBR      DATE  STORE_NBR  TXN_ID  PROD_NBR  \
0      1000  2018-10-17      1      1      5
1      1002  2018-09-16      1      2     58
2      1003  2019-03-07      1      3     52
3      1003  2019-03-08      1      4    106
4      1004  2018-11-02      1      5     96

      PROD_NAME  PROD_QTY  TOT_SALES  PACK_SIZE  \
0  Natural Chip      Compny SeaSalt175g      2      6.0      175
1  Red Rock Deli Chikn&Garlic Aioli 150g      1      2.7      150
2  Grain Waves Sour Cream&Chives 210G      1      3.6      210
3  Natural ChipCo      Honey Soy Chckn175g      1      3.0      175
4      WW Original Stacked Chips 160g      1      1.9      160

      BRAND      LIFESTAGE  PREMIUM_CUSTOMER
0  NATURAL  YOUNG  SINGLES/COUPLES      Premium
1  RRD      YOUNG  SINGLES/COUPLES      Mainstream
2  GRNWVES      YOUNG  FAMILIES      Budget
3  NATURAL      YOUNG  FAMILIES      Budget
4  WOOLWORTHS  OLDER  SINGLES/COUPLES      Mainstream ,
(264834, 12))
```

✓ APPROACH:-

1. Control Store Selection: Selecting control stores based on pre-trial metrics.
2. Trial vs Control Comparison: Comparing trial stores with their control stores during the trial period.
3. Visualizations and Statistical Tests: Generating visualizations and conducting significance tests.

Control Store Selection: Selecting control stores based on pre-trial metrics.

```
# Convert 'DATE' column to datetime if not already
qvi_data['DATE'] = pd.to_datetime(qvi_data['DATE'])

# Convert trial start and end dates to datetime objects
trial_start_date = pd.to_datetime('2019-02-01')
trial_end_date = pd.to_datetime('2019-04-30')

# Defining pre-trial period (adjust based on actual trial start date)
pre_trial_period = qvi_data[qvi_data['DATE'] < trial_start_date]

# Calculate monthly metrics: total sales, customers, transactions per customer
def calculate_metrics(data):
    metrics = data.groupby(['STORE_NBR', data['DATE'].dt.to_period('M')]).agg(
        total_sales=('TOT_SALES', 'sum'),
        num_customers=('LYLTY_CARD_NBR', 'nunique'),
        transactions_per_customer=('TXN_ID', 'count')
    ).reset_index()
    return metrics

# Calculate metrics for pre-trial period
pre_trial_metrics = calculate_metrics(pre_trial_period)

# Function to calculate Euclidean distance between trial and control store metrics
def find_control_store(trial_store_metrics, all_store_metrics):
    distances = {}

    # Loop through all stores to calculate distances
    for store in all_store_metrics['STORE_NBR'].unique():
        if store != trial_store_metrics['STORE_NBR'].unique()[0]:
            # Filter metrics for the control store
            control_store_metrics = all_store_metrics[all_store_metrics['STORE_NBR'] == store]

            # Align both trial and control store metrics by month (matching DATE period)
            merged_metrics = pd.merge(trial_store_metrics, control_store_metrics, on='DATE', suffixes=('_trial', '_control'))

            # Calculate the Euclidean distance across months for the relevant metrics
            distance = 0
            for _, row in merged_metrics.iterrows():
                trial_vector = [row['total_sales_trial'], row['num_customers_trial'], row['transactions_per_customer_trial']]
                control_vector = [row['total_sales_control'], row['num_customers_control'], row['transactions_per_customer_control']]

                # Sum the Euclidean distances across each month
                distance += euclidean(trial_vector, control_vector)

            # Store the distance for the control store
            distances[store] = distance

    # Return store with the smallest aggregated distance
    control_store = min(distances, key=distances.get)
    return control_store

# Finding control store for trial store 86
trial_store_86 = pre_trial_metrics[pre_trial_metrics['STORE_NBR'] == 86]
control_store_86 = find_control_store(trial_store_86, pre_trial_metrics)

print(f"Control store for trial store 86: {control_store_86}")

# Example: Find control store for trial store 77
trial_store_77 = pre_trial_metrics[pre_trial_metrics['STORE_NBR'] == 77]
control_store_77 = find_control_store(trial_store_77, pre_trial_metrics)

print(f"Control store for trial store 77: {control_store_77}")
```

```
# Example: Find control store for trial store 88
trial_store_88 = pre_trial_metrics[pre_trial_metrics['STORE_NBR'] == 88]
control_store_88 = find_control_store(trial_store_88, pre_trial_metrics)
```

```
print(f"Control store for trial store 88: {control_store_88}")
```

```
→ Control store for trial store 86: 155
   Control store for trial store 77: 233
   Control store for trial store 88: 237
```

Trial vs Control Comparison: Comparing trial stores with their control stores during the trial period.

```
# Function to compare trial and control store metrics
def compare_trial_and_control(trial_store, control_store, data):
    # Filter the data for the trial and control stores
    trial_store_data = data[data['STORE_NBR'] == trial_store]
    control_store_data = data[data['STORE_NBR'] == control_store]
    #trial_period = data[(data['DATE'] >= trial_start_date) & (data['DATE'] <= trial_end_date)]

    # Calculate metrics for both stores
    trial_metrics = calculate_metrics(trial_store_data)
    control_metrics = calculate_metrics(control_store_data)

    # Convert 'DATE' to string to avoid 'Period' issues
    trial_metrics['DATE'] = trial_metrics['DATE'].astype(str)
    control_metrics['DATE'] = control_metrics['DATE'].astype(str)

    # Plot comparison
    plt.figure(figsize=(10,6))
    plt.plot(trial_metrics['DATE'], trial_metrics['total_sales'], label=f'Trial Store {trial_store}')
    plt.plot(control_metrics['DATE'], control_metrics['total_sales'], label=f'Control Store {control_store}', linestyle='--')
    plt.title(f'Total Sales: Trial Store {trial_store} vs Control Store {control_store}')
    plt.xlabel('Month')
    plt.ylabel('Total Sales')
    plt.legend()
    plt.show()

    return trial_metrics, control_metrics

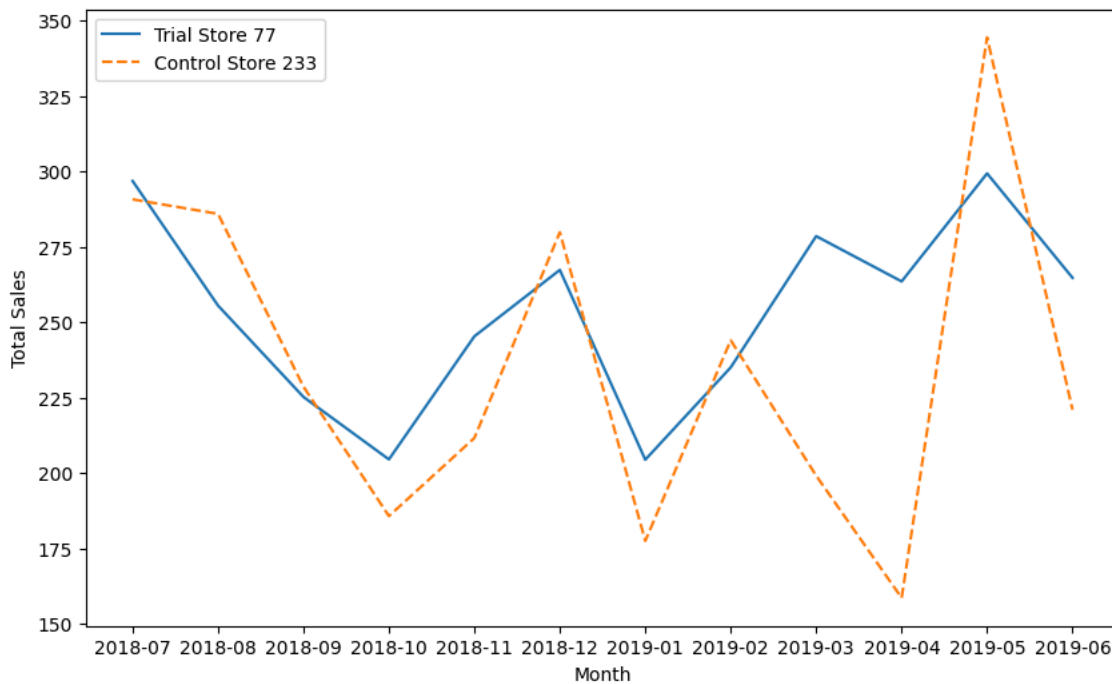
# Compare trial store 77 with its control store
trial_metrics_77, control_metrics_77 = compare_trial_and_control(77, control_store_77, qvi_data)

# Compare trial store 86 with its control store
trial_metrics_86, control_metrics_86 = compare_trial_and_control(86, control_store_86, qvi_data)

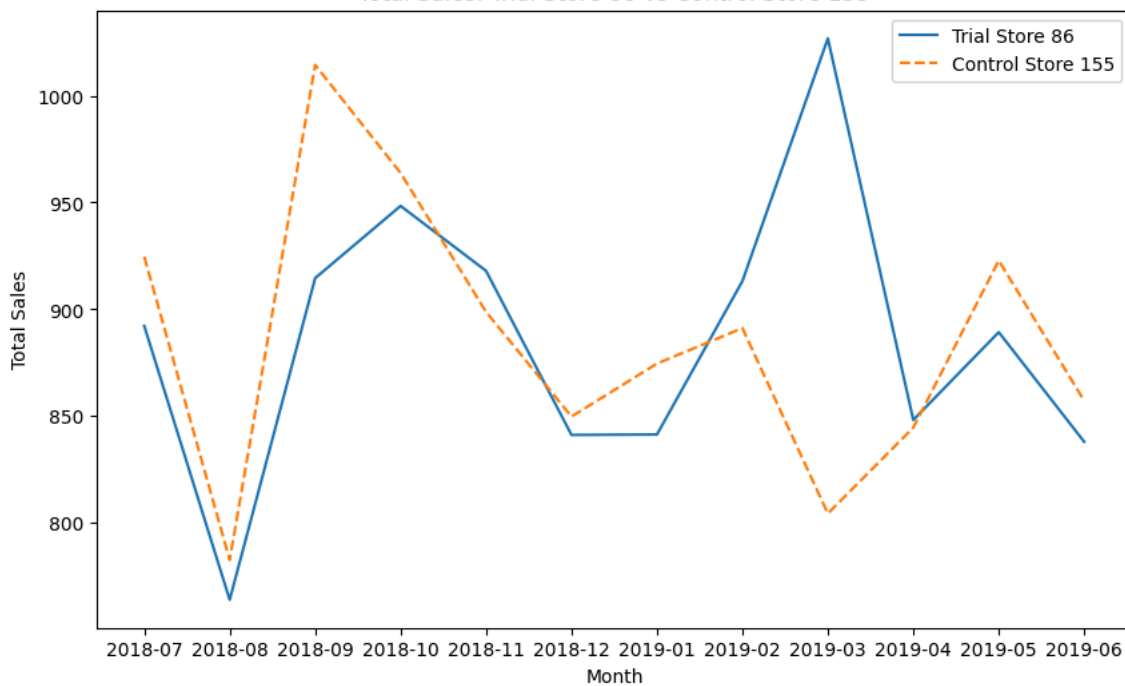
# Compare trial store 88 with its control store
trial_metrics_88, control_metrics_88 = compare_trial_and_control(88, control_store_88, qvi_data)
```



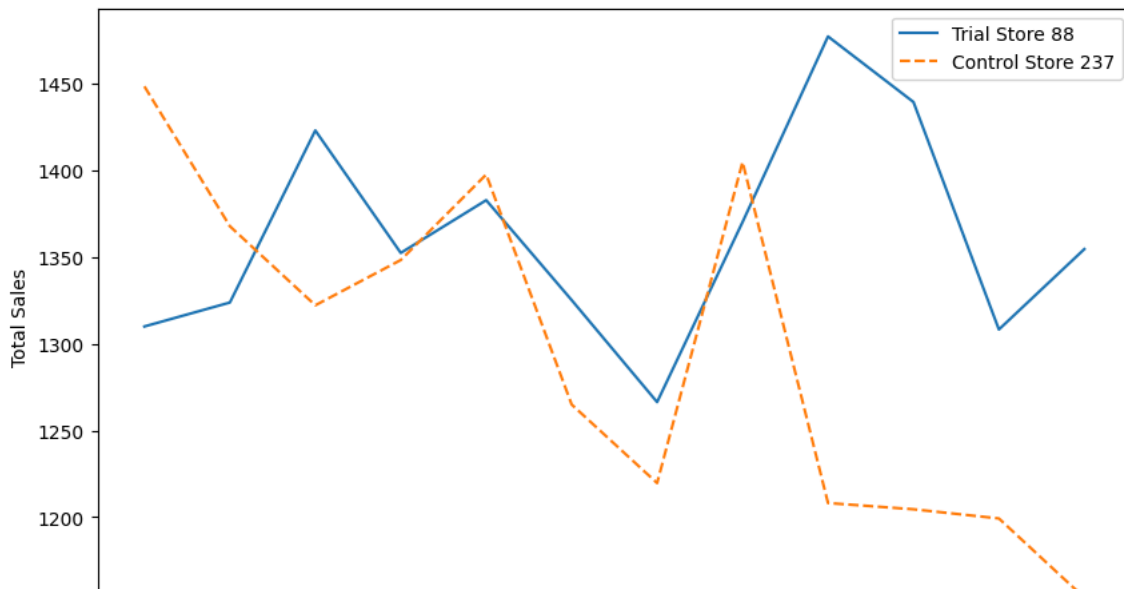
Total Sales: Trial Store 77 vs Control Store 233

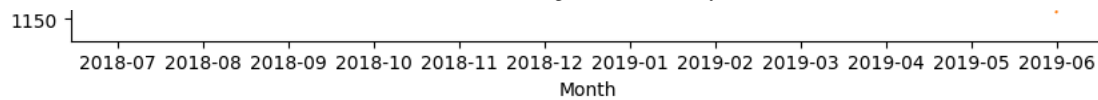


Total Sales: Trial Store 86 vs Control Store 155



Total Sales: Trial Store 88 vs Control Store 237





Visualizations and Statistical Tests: Generating visualizations and conducting significance tests.

```
# TRIAL STORE 77
import matplotlib.pyplot as plt
from scipy.stats import ttest_ind

# Perform t-test to check if sales during the trial are significantly different
t_stat, p_value = ttest_ind(trial_metrics_77['total_sales'], control_metrics_77['total_sales'])

# Print the t-statistic and p-value
print(f"T-test result: t-statistic = {t_stat}, p-value = {p_value}")

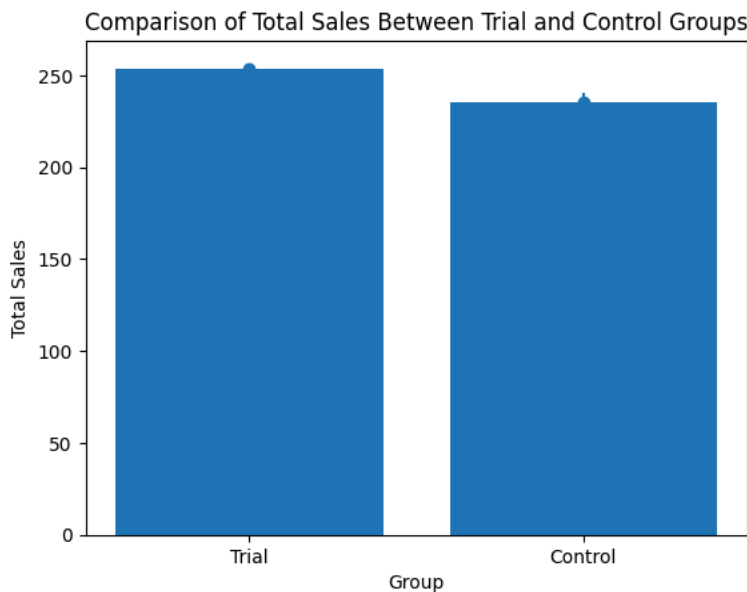
# Check the significance of the p-value
if p_value < 0.05:
    print("The difference in total sales is statistically significant.")
else:
    print("The difference in total sales is not statistically significant.")

# Create a bar plot to visualize the mean sales for each group
plt.bar(['Trial', 'Control'], [trial_metrics_77['total_sales'].mean(), control_metrics_77['total_sales'].mean()])
plt.xlabel('Group')
plt.ylabel('Total Sales')
plt.title('Comparison of Total Sales Between Trial and Control Groups')

# Add error bars representing standard errors
plt.errorbar(['Trial', 'Control'], [trial_metrics_77['total_sales'].mean(), control_metrics_77['total_sales'].mean()],
             yerr=[trial_metrics_77['total_sales'].std() / len(trial_metrics_77), control_metrics_77['total_sales'].std() / len(control_metrics_77)],
             fmt='o')

# Show the plot
plt.show()
```

→ T-test result: t-statistic = 0.9679236368390388, p-value = 0.34360373343122186
The difference in total sales is not statistically significant.



```
# TRIAL STORE 86
# Perform t-test to check if sales during the trial are significantly different
t_stat, p_value = ttest_ind(trial_metrics_86['total_sales'], control_metrics_86['total_sales'])

# Print the t-statistic and p-value
print(f"T-test result: t-statistic = {t_stat}, p-value = {p_value}")

# Check the significance of the p-value
if p_value < 0.05:
    print("The difference in total sales is statistically significant.")
else:
    print("The difference in total sales is not statistically significant.")

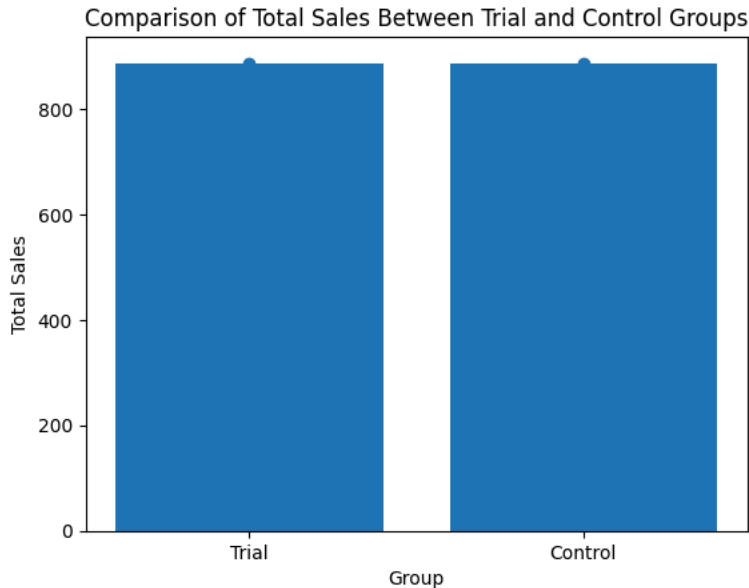
# Create a bar plot to visualize the mean sales for each group
plt.bar(['Trial', 'Control'], [trial_metrics_86['total_sales'].mean(), control_metrics_86['total_sales'].mean()])
plt.xlabel('Group')
plt.ylabel('Total Sales')
```

```
plt.title('Comparison of Total Sales Between Trial and Control Groups')

# Add error bars representing standard errors
plt.errorbar(['Trial', 'Control'], [trial_metrics_86['total_sales'].mean(), control_metrics_86['total_sales'].mean()],
            yerr=[trial_metrics_86['total_sales'].std() / len(trial_metrics_86), control_metrics_86['total_sales'].std() / len(control_metrics_86)],
            fmt='o')

# Show the plot
plt.show()
```

➡ T-test result: t-statistic = 0.019793022667677837, p-value = 0.9843869025850827
The difference in total sales is not statistically significant.



```
# TRIAL STORE 88
# Perform t-test to check if sales during the trial are significantly different
t_stat, p_value = ttest_ind(trial_metrics_88['total_sales'], control_metrics_88['total_sales'])

# Print the t-statistic and p-value
print(f"T-test result: t-statistic = {t_stat}, p-value = {p_value}")

# Check the significance of the p-value
if p_value < 0.05:
    print("The difference in total sales is statistically significant.")
else:
    print("The difference in total sales is not statistically significant.")

# Create a bar plot to visualize the mean sales for each group
plt.bar(['Trial', 'Control'], [trial_metrics_88['total_sales'].mean(), control_metrics_88['total_sales'].mean()])
plt.xlabel('Group')
plt.ylabel('Total Sales')
plt.title('Comparison of Total Sales Between Trial and Control Groups')

# Add error bars representing standard errors
plt.errorbar(['Trial', 'Control'], [trial_metrics_88['total_sales'].mean(), control_metrics_88['total_sales'].mean()],
            yerr=[trial_metrics_88['total_sales'].std() / len(trial_metrics_88), control_metrics_88['total_sales'].std() / len(control_metrics_88)],
            fmt='o')

# Show the plot
plt.show()
```

➡ T-test result: t-statistic = 1.9781599338138325, p-value = 0.06056801623928565
The difference in total sales is not statistically significant.

