

PCard-Expenditures-Analysis

August 20, 2023

1 Analysis of Toronto City Employee's Expense Card Transactions

1.1 Introduction

The City of Toronto is transparent about its employee expense card transactions, producing monthly reports available to the public. These transactions are accessible through the City's Open Data Portal.

1.2 Objectives

The core aims of this analysis include:

1. **Expenditure Characterization:** Understand the expenditure patterns of the City's Divisions or Cost Centres by exploring metrics like the number and frequency of transactions, their usual amounts, and their dispersion.
2. **Group Identification:** Determine significant groups of Divisions or Cost Centres with similar expense behaviours.
3. **Anomaly Spotting:** Uncover anomalies in the data that warrant further scrutiny.

1.3 Procedures

To accomplish these objectives, the following steps will be undertaken:

1.3.1 1. Data Collection and Preprocessing

1.1 Download the data from [Toronto's Open Data Portal](#).

```
[ ]: #1.1.1 Import required packages
```

```
from io import BytesIO
import matplotlib as mpl
import matplotlib.pyplot as plt

import os
import numpy as np
import pandas as pd
import re
import requests
import seaborn as sns
```

```

import zipfile

mpl.rcParams['figure.dpi'] = 300

# Shows plots in jupyter notebook
%matplotlib inline

# Set plot style
sns.set(color_codes=True)

import warnings

# Suppress the specific warnings
warnings.filterwarnings("ignore")

```

```

[ ]: #1.1.2
# We need to load th PCard Expenditure data into one dataframe so that we can
↳work with them in Python.
# We have used API for accessing this data. Consuming an API is more efficient
↳and less prone to errors and thus preferred over other ways of loading data.

base_url = "https://ckan0.cf.opendata.inter.prod-toronto.ca"
url = base_url + "/api/3/action/package_show"
params = {"id": "pcard-expenditures"}

package = requests.get(url, params=params).json()

all_dataframes = [] # List to hold individual DataFrames

# Folder in the same directory as the Jupyter notebook
download_folder = "./"

for idx, resource in enumerate(package["result"]["resources"]):
    if not resource["datastore_active"]:
        resource_url = base_url + "/api/3/action/resource_show?id=" +
↳resource["id"]
        resource_metadata = requests.get(resource_url).json()
        download_url = resource_metadata["result"]["url"]
        resource_format = resource["format"].lower()
        filename = download_url.split('/')[-1] # Extract filename from the URL

        print(f"Resource {idx} - URL: {download_url} - Format:
↳{resource_format}")

# Check if the resource format is zip
if resource_format == "zip":
    response = requests.get(download_url)

```

```

with zipfile.ZipFile(BytesIO(response.content)) as z:
    # Count the number of .xlsx files within the zip
    xlsx_files_count = sum(1 for file in z.namelist() if file.endswith('.
↳xlsx'))

    for file in z.namelist():
        # Only load xlsx files within the zip
        if file.endswith('xlsx'):
            try:
                df = pd.read_excel(z.open(file), engine='openpyxl')
                all_dataframes.append(df)
            except Exception as e:
                print(f"Error processing file {file} in {download_url}:
↳{e}")

        # Check if the number of .xlsx files matches the length of
↳all_dataframes
        if xlsx_files_count != len(all_dataframes):
            print(f"Warning: Number of xlsx files in the zip
↳({xlsx_files_count}) doesn't match the number of DataFrames loaded
↳({len(all_dataframes)}).")

        else:
            # Download and save (overwrite if exists) any other file
            response = requests.get(download_url)
            with open(os.path.join(download_folder, filename), 'wb') as file:
                file.write(response.content)

```

Resource 0 - URL: https://ckan0.cf.opendata.inter.prod-toronto.ca/dataset/ebc3f9c2-2f80-4405-bf4f-5fb309581485/resource/070bdbd3-9bae-4269-b096-e3a8bd7460c8/download/pcard_expenditures_readme.xls - Format: xls

Resource 1 - URL: <https://ckan0.cf.opendata.inter.prod-toronto.ca/dataset/ebc3f9c2-2f80-4405-bf4f-5fb309581485/resource/d83a5249-fb07-4c38-9145-9e12a32ce1d4/download/expenditures.zip> - Format: zip

1.2 Clean and consolidate the data into a single dataframe (combined_pcard)

```

[ ]: #1.2.1
# Make a copy of all_dataframes that will not affect the original loaded
↳dataframe list.
pcard_dataframe_list = [df.copy() for df in all_dataframes]

# Get columns of the first dataframe as the reference column for consolidation
↳into a single dataset.
reference_columns = pcard_dataframe_list[0].columns

# Flag to indicate if structures are the same

```

```

all_same_structure = True

# Iterate through all DataFrames in the list and Check whether the dataframes
↳in `all_dataframes` list have the same structure and column names
for idx, df in enumerate(pcard_dataframe_list[1:], start=1): # Start from the
↳second dataframe
    # If columns of current dataframe do not match reference columns
    if not df.columns.equals(reference_columns):
        print(f"DataFrame at index {idx} does not have the same structure as
↳the first DataFrame.")
        all_same_structure = False

```

```

DataFrame at index 3 does not have the same structure as the first DataFrame.
DataFrame at index 5 does not have the same structure as the first DataFrame.
DataFrame at index 6 does not have the same structure as the first DataFrame.
DataFrame at index 7 does not have the same structure as the first DataFrame.
DataFrame at index 8 does not have the same structure as the first DataFrame.
DataFrame at index 9 does not have the same structure as the first DataFrame.
DataFrame at index 10 does not have the same structure as the first DataFrame.
DataFrame at index 11 does not have the same structure as the first DataFrame.
DataFrame at index 13 does not have the same structure as the first DataFrame.
DataFrame at index 14 does not have the same structure as the first DataFrame.
DataFrame at index 15 does not have the same structure as the first DataFrame.
DataFrame at index 16 does not have the same structure as the first DataFrame.
DataFrame at index 17 does not have the same structure as the first DataFrame.
DataFrame at index 18 does not have the same structure as the first DataFrame.
DataFrame at index 19 does not have the same structure as the first DataFrame.
DataFrame at index 20 does not have the same structure as the first DataFrame.
DataFrame at index 21 does not have the same structure as the first DataFrame.
DataFrame at index 22 does not have the same structure as the first DataFrame.
DataFrame at index 23 does not have the same structure as the first DataFrame.
DataFrame at index 24 does not have the same structure as the first DataFrame.
DataFrame at index 26 does not have the same structure as the first DataFrame.
DataFrame at index 27 does not have the same structure as the first DataFrame.
DataFrame at index 28 does not have the same structure as the first DataFrame.
DataFrame at index 30 does not have the same structure as the first DataFrame.
DataFrame at index 31 does not have the same structure as the first DataFrame.
DataFrame at index 32 does not have the same structure as the first DataFrame.
DataFrame at index 33 does not have the same structure as the first DataFrame.
DataFrame at index 34 does not have the same structure as the first DataFrame.
DataFrame at index 35 does not have the same structure as the first DataFrame.
DataFrame at index 36 does not have the same structure as the first DataFrame.
DataFrame at index 37 does not have the same structure as the first DataFrame.
DataFrame at index 38 does not have the same structure as the first DataFrame.
DataFrame at index 39 does not have the same structure as the first DataFrame.
DataFrame at index 40 does not have the same structure as the first DataFrame.
DataFrame at index 41 does not have the same structure as the first DataFrame.

```

```
[ ]: #1.2.2
#check which columns in the other dataframes that differ from the first
↳ dataframe(reference_columns) in the `all_dataframes` list.

# Get columns of the first dataframe as the reference
reference_columns = set(pcard_dataframe_list[0].columns)

# Print the reference columns
print("Reference columns:", ', '.join(reference_columns))
print("")

# Flag to indicate if structures are the same
all_same_structure = True

# Iterate through all DataFrames in the list
for idx, df in enumerate(pcard_dataframe_list[1:], start=1): # Start from the
↳ second dataframe
    current_columns = set(df.columns)

    # Find columns that are in the reference but not in the current dataframe
    missing_in_current = reference_columns - current_columns
    # Find columns that are in the current dataframe but not in the reference
    additional_in_current = current_columns - reference_columns

    if missing_in_current or additional_in_current:
        all_same_structure = False
        print(f"DataFrame at index {idx} differs from the first DataFrame:")

        if missing_in_current:
            print(f"    Missing columns: {'', '.join(missing_in_current)}")
        if additional_in_current:
            print(f"    Additional columns: {'', '.join(additional_in_current)}")
        print("")
```

Reference columns: Merchant Type, Division, Cost Centre / WBS Element / Order No., Batch Transaction ID, Transaction Date, G/L Account, Transaction Amt., Original Amount, Merchant Name, Trx Currency, G/L Account Description, Original Currency, Purpose, Merchant Type Description, Card Posting Dt, Cost Centre / WBS Element / Order No. Description

DataFrame at index 3 differs from the first DataFrame:
 Missing columns: Cost Centre / WBS Element / Order No.
 Additional columns: Cost Centre / WBS Element / Order No

DataFrame at index 5 differs from the first DataFrame:
 Missing columns: Trx Currency
 Additional columns: Tr Currency

DataFrame at index 6 differs from the first DataFrame:

Missing columns: Division, Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Divison, Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 7 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Decription

DataFrame at index 8 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS element / Order Description, Cost Centre / WBS element / Order

DataFrame at index 9 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description, G/L Account Description

Additional columns: Cost Centre / WBS Element / Order , Cost Centre / WBS Element / Order Description, Exp Type Desc

DataFrame at index 10 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description, G/L Account Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description, Exp Type Desc

DataFrame at index 11 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 13 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 14 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 15 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 16 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 17 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 18 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 19 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order, Cost Center / WBS Element / Order Description

DataFrame at index 20 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 21 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 22 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description, G/L Account Description

Additional columns: Cost Centre/ WBS Element / Order, Cost Centre/ WBS Element / Order Description, Exp Type Desc

DataFrame at index 23 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 24 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 26 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 27 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 28 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre / WBS Element / Order, Cost Centre / WBS Element / Order Description

DataFrame at index 30 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description, G/L Account Description

Additional columns: Cost Center / WBS Element / Order, Exp Type Desc, Cost Center / WBS Element / Order Description

DataFrame at index 31 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order, Cost Center / WBS Element / Order Description

DataFrame at index 32 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order, Cost Center / WBS Element / Order Description

DataFrame at index 33 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Batch Transaction ID,
Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Centre /WBS Element / Order, Cost Centre /WBS Element
/ Order Description, Batch-Transaction ID

DataFrame at index 34 differs from the first DataFrame:

Missing columns: Division, Cost Centre / WBS Element / Order No., Cost Centre
/ WBS Element / Order No. Description

Additional columns: Division , Cost Center / WBS Element / Order, Cost Center
/ WBS Element / Order Description

DataFrame at index 35 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS
Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order, Cost Center / WBS
Element / Order Description

DataFrame at index 36 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS
Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order, Cost Center / WBS
Element / Order Description

DataFrame at index 37 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS
Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order Description, Cost
Center / WBS Element / Order

DataFrame at index 38 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS
Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order, Cost Center / WBS
Element / Order Description

DataFrame at index 39 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS
Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order #, Cost Center / WBS
Element / Order # Description

DataFrame at index 40 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS
Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order, Cost Center / WBS
Element / Order Description

DataFrame at index 41 differs from the first DataFrame:

Missing columns: Cost Centre / WBS Element / Order No., Cost Centre / WBS Element / Order No. Description

Additional columns: Cost Center / WBS Element / Order, Cost Center / WBS Element / Order Description

```
[ ]: #1.2.3
#By examining the output in 1.2.2 above it looks like the differences are in
↳the column names. Some have spaces, others have fullstop etc.
#Thus we need to normalize the column names and use consistent names through
↳out all the dataframes so that we can merge the data into one combined
↳dataframe.

# `pcard_dataframe_list` is a list of your dataframes
reference_columns = pcard_dataframe_list[0].columns.tolist()

def normalize_column_name(col_name):
    col_name = col_name.lower() # Convert to lowercase
    col_name = re.sub(r'[^a-z0-9]', '', col_name) # Remove non-alphanumeric
    ↳characters
    col_name = col_name.strip() # Remove leading and trailing spaces
    return col_name

column_mapping = {
    'glaccountdescription': 'gl_account_description',
    'purpose': 'purpose',
    'cardpostingdt': 'card_posting_dt',
    'batchtransactionid': 'batch_transaction_id',
    'trxcurrency': 'trx_currency',
    'trcurrency': 'trx_currency',
    'originalcurrency': 'original_currency',
    'costcentrewbselementordernodesdescription':
    ↳'cost_centre_wbselement_ordernodesdescription',
    'merchantname': 'merchant_name',
    'transactionamt': 'transaction_amt',
    'costcentrewbselementorderno': 'cost_centre_wbselement_orderno',
    'transactiondate': 'transaction_date',
    'originalamount': 'original_amount',
    'glaccount': 'gl_account',
    'merchantsdescription': 'merchant_type_description',
    'division': 'division',
    'divison': 'division',
    'merchants': 'merchant_type',
    'exptypedesc': 'gl_account_description',
    'costcentrewbselementorder': 'cost_centre_wbselement_orderno',

```

```

        'costcentrewbselementorderdescription':␣
↪ 'cost_centre_wbselement_ordernodesdescription',
        'costcentrewbselementorderdescription':␣
↪ 'cost_centre_wbselement_ordernodesdescription',
        'costcenterwbselementorder': 'cost_centre_wbselement_orderno',
        'costcenterwbselementorderdescription':␣
↪ 'cost_centre_wbselement_ordernodesdescription',
        'costcenterwblselementorderdescription':␣
↪ 'cost_centre_wbselement_ordernodesdescription',
        'costcenterwbselementorder#': 'cost_centre_wbselement_orderno',
        'costcenterwbselementorder#description':␣
↪ 'cost_centre_wbselement_ordernodesdescription',
    }

# Normalize and map columns in each dataframe
for i, df in enumerate(pcard_dataframe_list):
    df.columns = [column_mapping.get(normalize_column_name(col), col) for col
↪ in df.columns]

# Adjusting the reference dataframe's columns as well
reference_columns = [column_mapping.get(normalize_column_name(col), col) for
↪ col in reference_columns]
pcard_dataframe_list[0].columns = reference_columns

# Combine all dataframes into one dataframe
combined_df= pd.concat(pcard_dataframe_list, ignore_index=True, sort=False)

```

```

[ ]: #1.2.4
#Inspect the structure of the combined dataframe and see if everything checks
↪ out.
combined_df.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 220542 entries, 0 to 220541
Data columns (total 16 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   division                             184088 non-null  object
1   batch_transaction_id                 184088 non-null  object
2   transaction_date                     184087 non-null  datetime64[ns]
3   card_posting_dt                     184088 non-null  datetime64[ns]
4   merchant_name                       184088 non-null  object
5   transaction_amt                     218701 non-null  float64
6   trx_currency                        218701 non-null  object
7   original_amount                     220542 non-null  float64
8   original_currency                   220542 non-null  object
9   gl_account                          184088 non-null  object

```

```

10  gl_account_description          184088 non-null object
11  cost_centre_wbselement_orderno 183456 non-null object
12  cost_centre_wbselement_ordernodesdescription 183847 non-null object
13  merchant_type                  184068 non-null float64
14  merchant_type_description      184088 non-null object
15  purpose                        182692 non-null object
dtypes: datetime64[ns](2), float64(3), object(11)
memory usage: 26.9+ MB

```

```

[ ]: #1.2.5
      # Drop duplicate rows and reset index
      combined_df = combined_df.drop_duplicates().reset_index(drop=True)

```

```

[ ]: #1.2.6
      #Check for missing/null values
      combined_df.isnull().sum()

```

```

[ ]: division          31042
      batch_transaction_id 31042
      transaction_date    31043
      card_posting_dt     31042
      merchant_name       31042
      transaction_amt      1316
      trx_currency        1316
      original_amount      0
      original_currency    0
      gl_account          31042
      gl_account_description 31042
      cost_centre_wbselement_orderno 31674
      cost_centre_wbselement_ordernodesdescription 31283
      merchant_type       31062
      merchant_type_description 31042
      purpose             32428
      dtype: int64

```

```

[ ]: #1.2.7
      #Treat null values i.e. drop rows where division, batch_transaction_id,
      ↪merchant_type, card_posting_dt, merchant_name are null
      #We are dropping the null values since our analysis and investigation relies
      ↪heavily on the batch ID and division hence dropping the rows without
      ↪batch_transaction id
      # will make the analysis more useful

      mask = combined_df[['division','batch_transaction_id','transaction_date',
      ↪'merchant_type', 'card_posting_dt','merchant_name']].isnull().all(axis=1)
      combined_pcard = combined_df[~mask]
      combined_pcard.isnull().sum()

```

```
[ ]: division                                0
    batch_transaction_id                    0
    transaction_date                        1
    card_posting_dt                        0
    merchant_name                          0
    transaction_amt                        0
    trx_currency                          0
    original_amount                       0
    original_currency                     0
    gl_account                            0
    gl_account_description                 0
    cost_centre_wbselement_orderno       632
    cost_centre_wbselement_ordernodesdescription 241
    merchant_type                         20
    merchant_type_description             0
    purpose                              1386
    dtype: int64
```

```
[ ]: #1.2.8
    # Let us do further clean up by dropping observations with null transaction_
    ↪ dates.
    combined_pcard = combined_pcard.dropna(subset=['transaction_date'])
    combined_pcard.isnull().sum()
```

```
[ ]: division                                0
    batch_transaction_id                    0
    transaction_date                        0
    card_posting_dt                        0
    merchant_name                          0
    transaction_amt                        0
    trx_currency                          0
    original_amount                       0
    original_currency                     0
    gl_account                            0
    gl_account_description                 0
    cost_centre_wbselement_orderno       632
    cost_centre_wbselement_ordernodesdescription 241
    merchant_type                         20
    merchant_type_description             0
    purpose                              1386
    dtype: int64
```

```
[ ]: #1.2.9
    #Let us Inspect the strucutre and shape of our cleaned dataframe
    #Note: we have not treated the null values for purpose,
    ↪ cost_centre_wbs_elment_order_no, and
    ↪ cost_centre_wbs_element_order_no_description
```

```
# as this will not affect the outcome of our clustering analysis.
combined_pcard.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 183736 entries, 0 to 214775
Data columns (total 16 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   division                             183736 non-null object
1   batch_transaction_id                 183736 non-null object
2   transaction_date                     183736 non-null datetime64[ns]
3   card_posting_dt                     183736 non-null datetime64[ns]
4   merchant_name                       183736 non-null object
5   transaction_amt                     183736 non-null float64
6   trx_currency                        183736 non-null object
7   original_amount                     183736 non-null float64
8   original_currency                   183736 non-null object
9   gl_account                          183736 non-null object
10  gl_account_description               183736 non-null object
11  cost_centre_wbselement_orderno      183104 non-null object
12  cost_centre_wbselement_ordernodesdescription 183495 non-null object
13  merchant_type                       183716 non-null float64
14  merchant_type_description            183736 non-null object
15  purpose                             182350 non-null object
dtypes: datetime64[ns](2), float64(3), object(11)
memory usage: 23.8+ MB
```

```
[ ]: combined_pcard.nunique()
```

```
[ ]: division                             58
batch_transaction_id                 183720
transaction_date                     1290
card_posting_dt                     951
merchant_name                       15217
transaction_amt                     48568
trx_currency                        1
original_amount                     47885
original_currency                   22
gl_account                          303
gl_account_description               303
cost_centre_wbselement_orderno      9621
cost_centre_wbselement_ordernodesdescription 2612
merchant_type                       321
merchant_type_description            314
purpose                             94061
dtype: int64
```

```
[ ]: #1.2.10
#Since we will be focusing mostly on the the Division and GL Account features,
↳let us start by cleaning up the Division Column.

# Strip leading and trailing spaces
combined_pcard['division'] = combined_pcard['division'].str.strip()

# Replace & with AND and then remove special characters (but first check if the
↳value is a string)
combined_pcard['division'] = combined_pcard['division'].apply(
    lambda x: re.sub(r'[^a-zA-Z0-9\s]', '', re.sub(r'&', 'AND', x))
)

# Convert to uppercase
combined_pcard['division'] = combined_pcard['division'].str.upper()

sorted(combined_pcard['division'].unique())
```

```
[ ]: ['311 TORONTO',
      'ACCOUNTING SERVICES',
      'AFFORDABLE HOUSING OFFICE',
      'CFO',
      'CHILDRENS SERVICES',
      'CITY CLERKS OFFICE',
      'CITY MANAGER',
      'CITY MANAGERS OFFICE',
      'CITY PLANNING',
      'CORPORATE CONTRACTS',
      'CORPORATE FINANCE',
      'CORPORATE SECURITY',
      'COURT SERVICES',
      'DEPUTY CITY MGR AND CFO',
      'DEPUTY CITY MGR INTERNAL SERVICES',
      'ECONOMIC DEVELOPMENT AND CULTURE',
      'EMERGENCY MEDICAL SERVICES',
      'EMPLOYMENT AND SOCIAL SERVICES',
      'ENGINEERING AND CONSTRUCTION SERVICES',
      'ENVIRONMENT AND ENERGY',
      'ENVIRONMENT AND ENERGY OFFICE',
      'EXECUTIVE MANAGEMENT',
      'FACILITIES MANAGEMENT',
      'FACILITIES MANAGEMENT DIVISON',
      'FINANCE AND ADMINISTRATION',
      'FINANCIAL PLANNING',
      'FIRE SERVICES',
      'FLEET SERVICES',
      'HUMAN RESOURCES',
```

```

'INFORMATION AND TECHNOLOGY',
'INTERNAL AUDIT',
'LEGAL SERVICES',
'LONG TERM CARE HOMES',
'LONG TERM CARE HOMES AND SERVICES',
'MUNICIPAL LICENSING AND STANDARDS',
'OFFICE OF EMERGENCY MANAGEMENT',
'PARKS FORESTRY AND RECREATION',
'PENSION PAYROLL AND EMPLOYEE BENEFITS',
'POLICY PLANNING FINANCE AND ADMINISTRATION',
'PUBLIC HEALTH',
'PURCHASING AND MATERIALS MANAGEMENT',
'REAL ESTATE SERVICES',
'REVENUE SERVICES',
'SHELTER SUPPORT AND HOUSING ADMINISTRATION',
'SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION',
'SOLID WASTE MANAGEMENT',
'STRATEGIC AND CORPORATE POLICY',
'STRATEGIC COMMUNICATIONS',
'TORONTO BUILDING',
'TORONTO PARAMEDIC SERVICES',
'TORONTO WATER',
'TRANSPORTATION SERVICES',
'TREASURER']

```

```

[ ]: #1.2.11
#For divisions with minimal data or significance, we can group them together
↳for clarity.
#This will require us to get more information from the Toronto website on how
↳we can best group the divisions.

#1.2.11.1
#For this analysis we have used the latest Dec 2022 Divisions as highlighted in
↳the Business Expense report.

import requests
import pdfplumber

url = "https://www.toronto.ca/wp-content/uploads/2023/05/
↳9833-Business-Expense-Division-Dec-2022-vFinal-FINAL.pdf"
local_filename = "temp_division_pdf_file.pdf"

# Download the file
response = requests.get(url, stream=True)
with open(local_filename, 'wb') as f:
    for chunk in response.iter_content(chunk_size=8192):

```



```

        f.write(chunk)

# Open the downloaded file with pdfplumber
with pdfplumber.open(local_filename) as pdf:
    # Extract tables from all pages
    all_tables = [page.extract_table() for page in pdf.pages if page.
↪extract_table() is not None]

# Convert the tables to DataFrames
dfs = [pd.DataFrame(table[1:], columns=table[0]) for table in all_tables]

# Remove the temporary PDF file
os.remove(local_filename)

unique_divisions = [x for x in dfs[0]['Division'].str.upper().unique() if x is_
↪not None]
sorted_divisions = sorted(unique_divisions)
sorted_divisions

```

```

[ ]: ['ACCOUNTING SERVICES',
      'CAPITAL MARKET',
      "CHILDREN'S SERVICES",
      "CITY CLERK'S OFFICE",
      'CITY PLANNING',
      'CORPORATE REAL ESTATE MANAGEMENT',
      'COURT SERVICES',
      'CUSTOMER EXPERIENCE (311 TORONTO)',
      'ECONOMIC DEVELOPMENT AND CULTURE',
      'ENGINEERING & CONSTRUCTION SERVICES',
      'ENVIRONMENT & CLIMATE',
      'EXECUTIVE ADMINISTRATION',
      'FINANCIAL CONTROL AND PROCESS IMPROVEMENT',
      'FLEET SERVICES',
      'HOUSING SECRETARIAT',
      'INDIGENOUS AFFAIRS OFFICE',
      'INSURANCE & RISK MANAGEMENT',
      'INTERNAL AUDIT',
      'LEGAL SERVICES',
      'MUNICIPAL LICENSING AND STANDARDS',
      'OFFICE OF EMERGENCY MANAGEMENT',
      'OFFICE OF THE CHIEF OF STAFF',
      'OFFICE OF THE CISO',
      'OFFICE OF THE CONTROLLER',
      'OFFICE OF THE DEPUTY CITY MANAGER',
      'PARKS, FORESTRY & RECREATION',
      'PENSION, PAYROLL & EMPLOYEE BENEFITS',
      'PEOPLE & EQUITY',

```

```
'POLICY, PLANNING, FINANCE & ADMINISTRATION',
'PURCHASING & MATERIALS MANAGEMENT',
'REVENUE SERVICES',
'SENIORS SERVICES AND LONG-TERM CARE',
'SHELTER, SUPPORT AND HOUSING ADMINISTRATION',
'SOCIAL DEVELOPMENT, FINANCE AND ADMINISTRATION',
'SOLID WASTE MANAGEMENT SERVICES',
'STRATEGIC PARTNERSHIPS',
'STRATEGIC PUBLIC & EMPLOYEE COMMUNICATIONS',
'TECHNOLOGY SERVICES',
'TORONTO BUILDING',
'TORONTO EMPLOYMENT & SOCIAL SERVICES',
'TORONTO FIRE SERVICES',
'TORONTO PARAMEDIC SERVICES',
'TORONTO PUBLIC HEALTH',
'TORONTO WATER',
'TOTAL',
'TRANSIT EXPANSION OFFICE']
```

[]: #1.2.11.2

*#Based on the output of 1.2.11.1 let us do mapping of names to standardize the
↪division column.*

```
name_mapping = {
    '311 TORONTO': 'CUSTOMER EXPERIENCE (311 TORONTO)',
    'AFFORDABLE HOUSING OFFICE': 'SHELTER SUPPORT AND HOUSING ADMINISTRATION',
    'CFO': 'POLICY PLANNING FINANCE AND ADMINISTRATION',
    'DEPUTY CITY MGR AND CFO': 'OFFICE OF THE DEPUTY CITY MANAGER',
    'DEPUTY CITY MGR INTERNAL SERVICES': 'OFFICE OF THE DEPUTY CITY MANAGER',
    'CITY MANAGER': 'CITY MANAGERS OFFICE',
    'CORPORATE CONTRACTS': 'CORPORATE REAL ESTATE MANAGEMENT',
    'CORPORATE FINANCE': 'OFFICE OF THE DEPUTY CITY MANAGER',
    'CORPORATE SECURITY': 'CORPORATE REAL ESTATE MANAGEMENT',
    'EMERGENCY MEDICAL SERVICES': 'OFFICE OF EMERGENCY MANAGEMENT',
    'EMPLOYMENT AND SOCIAL SERVICES': 'SOCIAL DEVELOPMENT FINANCE AND  
↪ADMINISTRATION',
    'ENVIRONMENT AND ENERGY': 'ENVIRONMENT AND CLIMATE',
    'ENVIRONMENT AND ENERGY OFFICE': 'ENVIRONMENT AND CLIMATE',
    'FACILITIES MANAGEMENT': 'CORPORATE REAL ESTATE MANAGEMENT',
    'FACILITIES MANAGEMENT DIVISON': 'CORPORATE REAL ESTATE MANAGEMENT',
    'EXECUTIVE MANAGEMENT': 'CITY MANAGERS OFFICE',
    'FINANCE AND ADMINISTRATION': 'POLICY PLANNING FINANCE AND ADMINISTRATION',
    'FINANCIAL PLANNING': 'POLICY PLANNING FINANCE AND ADMINISTRATION',
    'FIRE SERVICES': 'TORONTO FIRE SERVICES',
    'HUMAN RESOURCES': 'PEOPLE & EQUITY',
    'INFORMATION AND TECHNOLOGY': 'TECHNOLOGY SERVICES',
    'LONG TERM CARE HOMES': 'SENIORS SERVICES AND LONGTERM CARE',
```

```

'LONG TERM CARE HOMES AND SERVICES': 'SENIORS SERVICES AND LONGTERM CARE',
'PUBLIC HEALTH': 'TORONTO PUBLIC HEALTH',
'REAL ESTATE SERVICES': 'CORPORATE REAL ESTATE MANAGEMENT',
'SOLID WASTE MANAGEMENT': 'SOLID WASTE MANAGEMENT SERVICES',
'STRATEGIC AND CORPORATE POLICY': 'CITY MANAGERS OFFICE',
'STRATEGIC COMMUNICATIONS': 'CITY MANAGERS OFFICE',
'TREASURER': 'POLICY PLANNING FINANCE AND ADMINISTRATION'
}

# Replace names using the mapping
combined_pcard['division'] = combined_pcard['division'].replace(name_mapping)

combined_pcard['division'].unique()

```

```

[ ]: array(['TORONTO PUBLIC HEALTH', 'ECONOMIC DEVELOPMENT AND CULTURE',
'PARKS FORESTRY AND RECREATION', 'TORONTO FIRE SERVICES',
'TRANSPORTATION SERVICES', 'TORONTO WATER',
'SENIORS SERVICES AND LONGTERM CARE', 'TORONTO PARAMEDIC SERVICES',
'SHELTER SUPPORT AND HOUSING ADMINISTRATION', 'CITY CLERKS OFFICE',
'MUNICIPAL LICENSING AND STANDARDS',
'SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION',
'CORPORATE REAL ESTATE MANAGEMENT',
'POLICY PLANNING FINANCE AND ADMINISTRATION', 'FLEET SERVICES',
'SOLID WASTE MANAGEMENT SERVICES', 'TECHNOLOGY SERVICES',
'CITY PLANNING', 'LEGAL SERVICES', 'CHILDRENS SERVICES',
'ACCOUNTING SERVICES', 'ENGINEERING AND CONSTRUCTION SERVICES',
'PURCHASING AND MATERIALS MANAGEMENT', 'CITY MANAGERS OFFICE',
'TORONTO BUILDING', 'REVENUE SERVICES', 'ENVIRONMENT AND CLIMATE',
'OFFICE OF THE DEPUTY CITY MANAGER',
'PENSION PAYROLL AND EMPLOYEE BENEFITS',
'OFFICE OF EMERGENCY MANAGEMENT', 'COURT SERVICES',
'PEOPLE & EQUITY', 'CUSTOMER EXPERIENCE (311 TORONTO)',
'INTERNAL AUDIT'], dtype=object)

```

```

[ ]: combined_pcard.info()

```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 183736 entries, 0 to 214775
Data columns (total 16 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   division                             183736 non-null object
1   batch_transaction_id                 183736 non-null object
2   transaction_date                     183736 non-null datetime64[ns]
3   card_posting_dt                     183736 non-null datetime64[ns]
4   merchant_name                       183736 non-null object
5   transaction_amt                     183736 non-null float64

```

```

6    trx_currency                    183736 non-null object
7    original_amount                 183736 non-null float64
8    original_currency               183736 non-null object
9    gl_account                      183736 non-null object
10   gl_account_description           183736 non-null object
11   cost_centre_wbselement_orderno  183104 non-null object
12   cost_centre_wbselement_ordernodesdescription 183495 non-null object
13   merchant_type                   183716 non-null float64
14   merchant_type_description       183736 non-null object
15   purpose                         182350 non-null object
dtypes: datetime64[ns](2), float64(3), object(11)
memory usage: 23.8+ MB

```

1.3.2 2. Division Profiling

- 2.1 Analyze expenditure amounts by division and Generate essential statistics for each Division's expenses such as measures of centrality, dispersion, and frequency.
- 2.2 Examine other potential characteristics including the diversity of the Cost Centres within each Division, usage of GL Accounts, choice of currencies, among others.
- 2.3 Aggregate the data entirely or present views by month or quarter.

Note: For divisions with minimal data or significance, group them together for clarity.

2.1 Characterize expenditure amounts by division and generate essential statistics for each Division's expenses such as measures of centrality, dispersion, and frequency.

2.1.1 Centrality, Dispersion, and Frequency Metrics for each Division's expenses:

```

[ ]: # Grouping by division and aggregating transaction amounts
division_stats = combined_pcard.groupby('division')['transaction_amt'].
    .agg(['mean', 'median', 'std', 'count', 'min', 'max']).reset_index()
division_stats['range'] = division_stats['max'] - division_stats['min']

# Rounding to 2 decimal places
cols_to_round = ['mean', 'median', 'std', 'min', 'max', 'range'] # List of
    columns to round
for col in cols_to_round:
    division_stats[col] = division_stats[col].round(2)

division_stats.sort_values(by='count')

```

```

[ ]:

```

	division	mean	median	std	\
12	INTERNAL AUDIT	293.79	293.79	NaN	
6	COURT SERVICES	842.28	263.04	1102.37	
7	CUSTOMER EXPERIENCE (311 TORONTO)	191.32	169.34	363.13	
18	PENSION PAYROLL AND EMPLOYEE BENEFITS	447.04	450.87	420.37	
21	PURCHASING AND MATERIALS MANAGEMENT	448.65	286.17	502.46	
19	PEOPLE & EQUITY	201.52	65.54	386.83	

16	OFFICE OF THE DEPUTY CITY MANAGER	328.11	143.70	570.55
15	OFFICE OF EMERGENCY MANAGEMENT	372.25	92.94	835.95
28	TORONTO BUILDING	613.27	247.69	857.18
0	ACCOUNTING SERVICES	732.83	804.19	730.37
22	REVENUE SERVICES	502.82	131.06	2593.37
27	TECHNOLOGY SERVICES	312.73	136.70	601.50
20	POLICY PLANNING FINANCE AND ADMINISTRATION	500.14	210.87	757.78
1	CHILDRENS SERVICES	651.46	367.82	852.27
4	CITY PLANNING	508.32	129.94	875.40
3	CITY MANAGERS OFFICE	175.43	76.32	292.87
2	CITY CLERKS OFFICE	456.69	187.45	769.35
10	ENVIRONMENT AND CLIMATE	442.94	169.50	875.30
14	MUNICIPAL LICENSING AND STANDARDS	193.70	50.00	480.15
33	TRANSPORTATION SERVICES	214.00	98.50	341.93
9	ENGINEERING AND CONSTRUCTION SERVICES	55.53	30.55	217.06
11	FLEET SERVICES	2711.51	491.00	3693.09
30	TORONTO PARAMEDIC SERVICES	295.32	62.28	1371.50
24	SHELTER SUPPORT AND HOUSING ADMINISTRATION	805.01	265.21	2090.86
26	SOLID WASTE MANAGEMENT SERVICES	205.59	90.39	355.12
29	TORONTO FIRE SERVICES	293.61	77.20	613.57
31	TORONTO PUBLIC HEALTH	387.27	145.00	733.21
23	SENIORS SERVICES AND LONGTERM CARE	396.27	156.50	654.44
25	SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	235.06	50.84	620.60
13	LEGAL SERVICES	93.71	17.00	251.78
32	TORONTO WATER	203.36	104.37	334.34
5	CORPORATE REAL ESTATE MANAGEMENT	199.47	83.87	357.30
8	ECONOMIC DEVELOPMENT AND CULTURE	363.97	79.63	1164.00
17	PARKS FORESTRY AND RECREATION	199.38	81.81	341.98

	count	min	max	range
12	1	293.79	293.79	0.00
6	18	19.66	3384.35	3364.69
7	20	-734.50	827.14	1561.64
18	28	-89.27	1695.00	1784.27
21	110	-318.66	2687.12	3005.78
19	112	-399.80	1199.40	1599.20
16	114	-1400.00	3467.84	4867.84
15	151	-446.80	7289.63	7736.43
28	188	-2970.67	2998.00	5968.67
0	218	-3744.30	3744.30	7488.60
22	240	-777.02	39493.50	40270.52
27	241	-62.09	3164.00	3226.09
20	336	-1017.00	5210.00	6227.00
1	345	-3842.00	4294.00	8136.00
4	350	-371.52	6107.56	6479.08
3	455	-899.54	2541.79	3441.33
2	587	-1000.00	6942.70	7942.70

10	701	-669.53	6667.00	7336.53
14	904	-599.88	8859.20	9459.08
33	1397	-455.82	2841.09	3296.91
9	1426	-3334.26	3334.26	6668.52
11	2005	-4249.25	10000.00	14249.25
30	2075	-2500.00	41205.75	43705.75
24	3085	-16150.00	25000.00	41150.00
26	3190	-1050.90	4514.35	5565.25
29	4319	-2272.69	18958.50	21231.19
31	4936	-3000.00	9945.70	12945.70
23	5298	-5022.49	7079.45	12101.94
25	5458	-4746.00	6780.00	11526.00
13	5821	-785.35	6068.52	6853.87
32	11709	-2807.00	3000.00	5807.00
5	14715	-2527.71	15010.37	17538.08
8	14846	-4234.54	61842.37	66076.91
17	98337	-3000.00	6162.12	9162.12

2.2 Examine other potential characteristics including the diversity of the Cost Centres within each Division, usage of GL Accounts, choice of currencies, among others. 2.2.1 Diversity/Concentration Metrics:

```
[ ]: # Calculate total unique values for each category across the entire dataset
total_unique_cost_centres = combined_pcard['cost_centre_wbselement_orderno'].
    ↪nunique()
total_unique_gl_accounts = combined_pcard['gl_account'].nunique()
total_unique_original_currency = combined_pcard['original_currency'].nunique()
total_unique_merchant_type = combined_pcard['merchant_type'].nunique()
total_unique_purpose = combined_pcard['purpose'].nunique()

# Diversity/Concentration Metrics by Division
def get_diversity_df(groupby_column, total_unique):
    counts = combined_pcard.groupby('division')[groupby_column].nunique()
    percentages = ((counts / total_unique) * 100).round(1)
    df = pd.DataFrame({
        'Count': counts,
        'Percentage of Total (%)': percentages
    }).sort_values(by='Count', ascending=False)
    return df

div_cost_centre_df = get_diversity_df('cost_centre_wbselement_orderno',
    ↪total_unique_cost_centres)
div_gl_account_df = get_diversity_df('gl_account', total_unique_gl_accounts)
div_unique_original_currency = get_diversity_df('original_currency',
    ↪total_unique_original_currency)
```

```

div_unique_merchant_type = get_diversity_df('merchant_type',
↳total_unique_merchant_type)
div_purpose_df = get_diversity_df('purpose', total_unique_purpose)

print("Diversity/Concentration Metrics for each Division:")
print("\nNumber of unique Cost Centres:")
display(div_cost_centre_df)
print(" ")
print("\nNumber of unique GL Accounts:")
display(div_gl_account_df)
print(" ")
print("\nNumber of unique Currencies:")
display(div_unique_original_currency)
print(" ")
print("\nNumber of unique Merchant Types:")
display(div_unique_merchant_type)
print(" ")
print("\nNumber of unique Purposes:")
display(div_purpose_df)

```

Diversity/Concentration Metrics for each Division:

Number of unique Cost Centres:

	Count	Percentage of Total (%)
division		
CORPORATE REAL ESTATE MANAGEMENT	7010	72.9
PARKS FORESTRY AND RECREATION	1532	15.9
ECONOMIC DEVELOPMENT AND CULTURE	172	1.8
TORONTO PUBLIC HEALTH	125	1.3
TORONTO WATER	105	1.1
SENIORS SERVICES AND LONGTERM CARE	98	1.0
SOLID WASTE MANAGEMENT SERVICES	84	0.9
LEGAL SERVICES	76	0.8
SHELTER SUPPORT AND HOUSING ADMINISTRATION	74	0.8
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	65	0.7
TORONTO PARAMEDIC SERVICES	59	0.6
TRANSPORTATION SERVICES	35	0.4
POLICY PLANNING FINANCE AND ADMINISTRATION	32	0.3
MUNICIPAL LICENSING AND STANDARDS	27	0.3
CITY CLERKS OFFICE	25	0.3
TORONTO FIRE SERVICES	25	0.3
OFFICE OF THE DEPUTY CITY MANAGER	20	0.2
OFFICE OF EMERGENCY MANAGEMENT	19	0.2
ENVIRONMENT AND CLIMATE	16	0.2
ACCOUNTING SERVICES	16	0.2
PURCHASING AND MATERIALS MANAGEMENT	11	0.1

REVENUE SERVICES	10	0.1
CITY PLANNING	9	0.1
CITY MANAGERS OFFICE	9	0.1
ENGINEERING AND CONSTRUCTION SERVICES	8	0.1
FLEET SERVICES	7	0.1
COURT SERVICES	6	0.1
PENSION PAYROLL AND EMPLOYEE BENEFITS	6	0.1
TECHNOLOGY SERVICES	5	0.1
TORONTO BUILDING	5	0.1
CHILDRENS SERVICES	3	0.0
PEOPLE & EQUITY	2	0.0
CUSTOMER EXPERIENCE (311 TORONTO)	2	0.0
INTERNAL AUDIT	1	0.0

Number of unique GL Accounts:

	Count	Percentage of Total (%)
division		
PARKS FORESTRY AND RECREATION	219	72.3
ECONOMIC DEVELOPMENT AND CULTURE	147	48.5
TORONTO WATER	138	45.5
SHELTER SUPPORT AND HOUSING ADMINISTRATION	131	43.2
SOLID WASTE MANAGEMENT SERVICES	125	41.3
TORONTO PUBLIC HEALTH	106	35.0
TRANSPORTATION SERVICES	101	33.3
CORPORATE REAL ESTATE MANAGEMENT	99	32.7
TORONTO FIRE SERVICES	95	31.4
SENIORS SERVICES AND LONGTERM CARE	83	27.4
TORONTO PARAMEDIC SERVICES	77	25.4
MUNICIPAL LICENSING AND STANDARDS	77	25.4
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	64	21.1
CITY CLERKS OFFICE	49	16.2
ENVIRONMENT AND CLIMATE	45	14.9
OFFICE OF EMERGENCY MANAGEMENT	41	13.5
REVENUE SERVICES	40	13.2
FLEET SERVICES	37	12.2
OFFICE OF THE DEPUTY CITY MANAGER	28	9.2
TORONTO BUILDING	28	9.2
POLICY PLANNING FINANCE AND ADMINISTRATION	24	7.9
PURCHASING AND MATERIALS MANAGEMENT	23	7.6
ACCOUNTING SERVICES	22	7.3
CITY MANAGERS OFFICE	21	6.9
TECHNOLOGY SERVICES	18	5.9
LEGAL SERVICES	18	5.9
CITY PLANNING	17	5.6
CHILDRENS SERVICES	13	4.3
COURT SERVICES	10	3.3

ENGINEERING AND CONSTRUCTION SERVICES	9	3.0
PENSION PAYROLL AND EMPLOYEE BENEFITS	7	2.3
CUSTOMER EXPERIENCE (311 TORONTO)	6	2.0
PEOPLE & EQUITY	4	1.3
INTERNAL AUDIT	1	0.3

Number of unique Currencies:

	Count	Percentage of Total (%)
division		
ECONOMIC DEVELOPMENT AND CULTURE	20	90.9
TORONTO PUBLIC HEALTH	7	31.8
PARKS FORESTRY AND RECREATION	6	27.3
CITY PLANNING	4	18.2
TORONTO WATER	4	18.2
FLEET SERVICES	3	13.6
SENIORS SERVICES AND LONGTERM CARE	3	13.6
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	3	13.6
SOLID WASTE MANAGEMENT SERVICES	3	13.6
TORONTO FIRE SERVICES	3	13.6
TORONTO PARAMEDIC SERVICES	3	13.6
CITY MANAGERS OFFICE	3	13.6
CITY CLERKS OFFICE	3	13.6
TORONTO BUILDING	2	9.1
POLICY PLANNING FINANCE AND ADMINISTRATION	2	9.1
TECHNOLOGY SERVICES	2	9.1
SHELTER SUPPORT AND HOUSING ADMINISTRATION	2	9.1
REVENUE SERVICES	2	9.1
PURCHASING AND MATERIALS MANAGEMENT	2	9.1
ACCOUNTING SERVICES	2	9.1
PEOPLE & EQUITY	2	9.1
CHILDRENS SERVICES	2	9.1
OFFICE OF THE DEPUTY CITY MANAGER	2	9.1
OFFICE OF EMERGENCY MANAGEMENT	2	9.1
MUNICIPAL LICENSING AND STANDARDS	2	9.1
LEGAL SERVICES	2	9.1
ENVIRONMENT AND CLIMATE	2	9.1
ENGINEERING AND CONSTRUCTION SERVICES	2	9.1
CORPORATE REAL ESTATE MANAGEMENT	2	9.1
TRANSPORTATION SERVICES	2	9.1
PENSION PAYROLL AND EMPLOYEE BENEFITS	1	4.5
INTERNAL AUDIT	1	4.5
CUSTOMER EXPERIENCE (311 TORONTO)	1	4.5
COURT SERVICES	1	4.5

Number of unique Merchant Types:

	Count	Percentage of Total (%)
division		
ECONOMIC DEVELOPMENT AND CULTURE	236	73.5
PARKS FORESTRY AND RECREATION	224	69.8
TORONTO FIRE SERVICES	160	49.8
TORONTO WATER	159	49.5
SHELTER SUPPORT AND HOUSING ADMINISTRATION	159	49.5
TORONTO PUBLIC HEALTH	152	47.4
SENIORS SERVICES AND LONGTERM CARE	143	44.5
CORPORATE REAL ESTATE MANAGEMENT	142	44.2
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	117	36.4
SOLID WASTE MANAGEMENT SERVICES	111	34.6
TORONTO PARAMEDIC SERVICES	102	31.8
CITY CLERKS OFFICE	92	28.7
TRANSPORTATION SERVICES	90	28.0
ENVIRONMENT AND CLIMATE	84	26.2
MUNICIPAL LICENSING AND STANDARDS	79	24.6
OFFICE OF EMERGENCY MANAGEMENT	49	15.3
REVENUE SERVICES	47	14.6
POLICY PLANNING FINANCE AND ADMINISTRATION	43	13.4
OFFICE OF THE DEPUTY CITY MANAGER	42	13.1
FLEET SERVICES	40	12.5
LEGAL SERVICES	39	12.1
TORONTO BUILDING	37	11.5
CITY MANAGERS OFFICE	36	11.2
TECHNOLOGY SERVICES	35	10.9
CHILDRENS SERVICES	35	10.9
PURCHASING AND MATERIALS MANAGEMENT	35	10.9
ACCOUNTING SERVICES	33	10.3
CITY PLANNING	29	9.0
ENGINEERING AND CONSTRUCTION SERVICES	18	5.6
COURT SERVICES	11	3.4
CUSTOMER EXPERIENCE (311 TORONTO)	10	3.1
PEOPLE & EQUITY	9	2.8
PENSION PAYROLL AND EMPLOYEE BENEFITS	7	2.2
INTERNAL AUDIT	1	0.3

Number of unique Purposes:

	Count	Percentage of Total (%)
division		
PARKS FORESTRY AND RECREATION	55494	59.0
ECONOMIC DEVELOPMENT AND CULTURE	8704	9.3
TORONTO WATER	7046	7.5
CORPORATE REAL ESTATE MANAGEMENT	6702	7.1

SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	4206	4.5
SENIORS SERVICES AND LONGTERM CARE	3677	3.9
SHELTER SUPPORT AND HOUSING ADMINISTRATION	2546	2.7
SOLID WASTE MANAGEMENT SERVICES	2321	2.5
TORONTO FIRE SERVICES	2280	2.4
TORONTO PARAMEDIC SERVICES	1032	1.1
TORONTO PUBLIC HEALTH	975	1.0
TRANSPORTATION SERVICES	910	1.0
ENGINEERING AND CONSTRUCTION SERVICES	604	0.6
CITY CLERKS OFFICE	475	0.5
ENVIRONMENT AND CLIMATE	429	0.5
MUNICIPAL LICENSING AND STANDARDS	403	0.4
FLEET SERVICES	388	0.4
CITY PLANNING	278	0.3
CITY MANAGERS OFFICE	250	0.3
REVENUE SERVICES	213	0.2
LEGAL SERVICES	210	0.2
CHILDRENS SERVICES	180	0.2
TECHNOLOGY SERVICES	172	0.2
TORONTO BUILDING	155	0.2
POLICY PLANNING FINANCE AND ADMINISTRATION	144	0.2
ACCOUNTING SERVICES	126	0.1
OFFICE OF EMERGENCY MANAGEMENT	111	0.1
OFFICE OF THE DEPUTY CITY MANAGER	106	0.1
PURCHASING AND MATERIALS MANAGEMENT	105	0.1
CUSTOMER EXPERIENCE (311 TORONTO)	20	0.0
PENSION PAYROLL AND EMPLOYEE BENEFITS	20	0.0
COURT SERVICES	17	0.0
PEOPLE & EQUITY	13	0.0
INTERNAL AUDIT	1	0.0

2.3 Aggregate across the entire dataset or a view aggregated by month or quarter.

2.3.1 Average Monthly/Quarterly Expenditure Amount:

```
[ ]: # Extracting month and quarter from the transaction_date
combined_pcard['transaction_month'] = combined_pcard['transaction_date'].dt.
    ↳ month
combined_pcard['transaction_quarter'] = combined_pcard['transaction_date'].dt.
    ↳ quarter
combined_pcard['transaction_year'] = combined_pcard['transaction_date'].dt.year

# Monthly aggregation
monthly_expenses = combined_pcard.groupby(['division',
    ↳ 'transaction_month'])['transaction_amt'].mean().round(2).reset_index()

# Quarterly aggregation
```

```

quarterly_expenses = combined_pcard.groupby(['division',
↳ 'transaction_quarter'])['transaction_amt'].mean().round(2).reset_index()

# Yearly aggregation
yearly_expenses = combined_pcard.groupby(['division',
↳ 'transaction_year'])['transaction_amt'].mean().round(2).reset_index()

# Pivot tables with NaN values replaced by empty strings
monthly_pivot = monthly_expenses.pivot_table(index='division',
↳ columns='transaction_month', values='transaction_amt', aggfunc='mean').
↳ fillna("")
print("Monthly Average by Division:")
display(monthly_pivot)
print("")

```

Monthly Average by Division:

transaction_month	1	2	3	\
division				
ACCOUNTING SERVICES	534.7	551.99	465.93	
CHILDRENS SERVICES	601.7	778.18	668.27	
CITY CLERKS OFFICE	377.63	405.27	346.46	
CITY MANAGERS OFFICE	139.88	178.22	322.46	
CITY PLANNING	449.77	417.41	584.96	
CORPORATE REAL ESTATE MANAGEMENT	197.26	181.98	199.86	
COURT SERVICES	1862.57			
CUSTOMER EXPERIENCE (311 TORONTO)		142.13		
ECONOMIC DEVELOPMENT AND CULTURE	384.31	342.07	347.66	
ENGINEERING AND CONSTRUCTION SERVICES	50.2	44.46	54.64	
ENVIRONMENT AND CLIMATE	512.34	340.69	442.57	
FLEET SERVICES	2268.69	3026.92	2030.62	
INTERNAL AUDIT				
LEGAL SERVICES	87.55	74.83	107.28	
MUNICIPAL LICENSING AND STANDARDS	311.07	290.0	100.84	
OFFICE OF EMERGENCY MANAGEMENT	188.36	142.35	1115.32	
OFFICE OF THE DEPUTY CITY MANAGER	315.74	288.74	78.01	
PARKS FORESTRY AND RECREATION	175.55	172.03	191.21	
PENSION PAYROLL AND EMPLOYEE BENEFITS		931.97	940.72	
PEOPLE & EQUITY	1199.4	78.59	144.56	
POLICY PLANNING FINANCE AND ADMINISTRATION	328.44	245.09	667.12	
PURCHASING AND MATERIALS MANAGEMENT	337.66	451.06	337.91	
REVENUE SERVICES	248.42	323.46	300.99	
SENIORS SERVICES AND LONGTERM CARE	483.39	453.67	376.53	
SHELTER SUPPORT AND HOUSING ADMINISTRATION	964.63	1145.79	894.02	
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	225.54	276.91	270.24	
SOLID WASTE MANAGEMENT SERVICES	221.51	217.79	219.4	
TECHNOLOGY SERVICES	475.79	227.33	146.63	
TORONTO BUILDING	169.97	611.58	807.34	

TORONTO FIRE SERVICES	269.64	251.24	281.04
TORONTO PARAMEDIC SERVICES	930.86	440.64	225.52
TORONTO PUBLIC HEALTH	439.11	337.5	355.65
TORONTO WATER	206.37	205.59	206.55
TRANSPORTATION SERVICES	153.77	178.55	200.14

transaction_month	4	5	6 \
division			
ACCOUNTING SERVICES	956.2	838.83	693.98
CHILDRENS SERVICES	345.7	608.53	566.68
CITY CLERKS OFFICE	358.99	485.94	660.16
CITY MANAGERS OFFICE	171.53	114.26	66.87
CITY PLANNING	356.59	389.05	601.41
CORPORATE REAL ESTATE MANAGEMENT	205.33	211.39	204.45
COURT SERVICES	1107.4	502.05	93.77
CUSTOMER EXPERIENCE (311 TORONTO)	130.76		36.58
ECONOMIC DEVELOPMENT AND CULTURE	381.61	246.72	366.09
ENGINEERING AND CONSTRUCTION SERVICES	71.31	58.79	55.35
ENVIRONMENT AND CLIMATE	394.65	359.49	321.56
FLEET SERVICES	2435.57	726.57	592.31
INTERNAL AUDIT			
LEGAL SERVICES	76.12	92.7	105.98
MUNICIPAL LICENSING AND STANDARDS	228.59	244.27	252.58
OFFICE OF EMERGENCY MANAGEMENT	202.22	154.72	990.85
OFFICE OF THE DEPUTY CITY MANAGER	232.89	286.04	483.5
PARKS FORESTRY AND RECREATION	188.54	192.54	212.11
PENSION PAYROLL AND EMPLOYEE BENEFITS	488.63	492.12	388.68
PEOPLE & EQUITY	101.04	988.75	1124.35
POLICY PLANNING FINANCE AND ADMINISTRATION	593.47	979.64	291.44
PURCHASING AND MATERIALS MANAGEMENT	304.24	771.64	146.73
REVENUE SERVICES	1897.68	482.08	330.07
SENIORS SERVICES AND LONGTERM CARE	343.94	347.95	351.07
SHELTER SUPPORT AND HOUSING ADMINISTRATION	1011.44	732.44	706.13
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	254.16	195.16	205.27
SOLID WASTE MANAGEMENT SERVICES	219.57	236.42	225.87
TECHNOLOGY SERVICES	349.37	272.17	577.36
TORONTO BUILDING	326.2	573.5	549.38
TORONTO FIRE SERVICES	270.98	269.49	307.9
TORONTO PARAMEDIC SERVICES	290.33	214.44	179.34
TORONTO PUBLIC HEALTH	382.82	426.95	344.85
TORONTO WATER	198.97	201.56	202.7
TRANSPORTATION SERVICES	262.53	250.62	247.02

transaction_month	7	8	9 \
division			
ACCOUNTING SERVICES	821.57	320.39	554.25
CHILDRENS SERVICES	761.47	980.22	655.01
CITY CLERKS OFFICE	761.88	268.94	438.38

CITY MANAGERS OFFICE	311.76	214.11	107.18
CITY PLANNING	384.99	598.61	166.45
CORPORATE REAL ESTATE MANAGEMENT	187.19	195.75	210.15
COURT SERVICES			
CUSTOMER EXPERIENCE (311 TORONTO)	465.53		195.28
ECONOMIC DEVELOPMENT AND CULTURE	361.73	319.74	472.17
ENGINEERING AND CONSTRUCTION SERVICES	52.77	40.3	61.78
ENVIRONMENT AND CLIMATE	458.46	477.35	575.23
FLEET SERVICES	457.54	5642.1	5211.09
INTERNAL AUDIT			293.79
LEGAL SERVICES	90.6	94.39	95.9
MUNICIPAL LICENSING AND STANDARDS	59.23	135.12	338.35
OFFICE OF EMERGENCY MANAGEMENT	398.58	450.87	364.0
OFFICE OF THE DEPUTY CITY MANAGER	819.25	1000.0	191.5
PARKS FORESTRY AND RECREATION	215.66	210.85	235.1
PENSION PAYROLL AND EMPLOYEE BENEFITS	149.27	480.63	-89.27
PEOPLE & EQUITY		-374.81	
POLICY PLANNING FINANCE AND ADMINISTRATION	408.25	450.09	127.16
PURCHASING AND MATERIALS MANAGEMENT	865.12	451.46	604.62
REVENUE SERVICES	303.6	343.99	328.55
SENIORS SERVICES AND LONGTERM CARE	423.22	356.97	341.21
SHELTER SUPPORT AND HOUSING ADMINISTRATION	458.7	456.57	574.09
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	298.41	384.01	142.04
SOLID WASTE MANAGEMENT SERVICES	201.02	180.21	134.42
TECHNOLOGY SERVICES	105.71	106.91	151.06
TORONTO BUILDING	513.05	1047.37	435.08
TORONTO FIRE SERVICES	288.56	292.33	238.63
TORONTO PARAMEDIC SERVICES	206.72	234.1	254.55
TORONTO PUBLIC HEALTH	341.57	497.49	394.87
TORONTO WATER	207.91	193.39	214.83
TRANSPORTATION SERVICES	208.93	151.27	270.63
transaction_month	10	11	12
division			
ACCOUNTING SERVICES	799.84	363.92	537.31
CHILDRENS SERVICES	1396.4	591.9	511.68
CITY CLERKS OFFICE	401.86	490.64	503.77
CITY MANAGERS OFFICE	580.75	185.45	305.38
CITY PLANNING	654.42	840.14	948.7
CORPORATE REAL ESTATE MANAGEMENT	206.39	214.36	179.83
COURT SERVICES		894.88	19.66
CUSTOMER EXPERIENCE (311 TORONTO)	275.08	1.71	272.16
ECONOMIC DEVELOPMENT AND CULTURE	385.91	341.19	454.0
ENGINEERING AND CONSTRUCTION SERVICES	41.45	91.62	105.87
ENVIRONMENT AND CLIMATE	822.55	306.0	578.14
FLEET SERVICES	4230.27	1894.93	1879.97
INTERNAL AUDIT			
LEGAL SERVICES	84.72	115.06	126.82

MUNICIPAL LICENSING AND STANDARDS	227.81	178.69	135.87
OFFICE OF EMERGENCY MANAGEMENT	1234.04	684.64	932.85
OFFICE OF THE DEPUTY CITY MANAGER	859.12	111.82	336.32
PARKS FORESTRY AND RECREATION	188.94	188.43	232.12
PENSION PAYROLL AND EMPLOYEE BENEFITS			
PEOPLE & EQUITY	78.05		1199.4
POLICY PLANNING FINANCE AND ADMINISTRATION	296.76	353.51	140.86
PURCHASING AND MATERIALS MANAGEMENT	515.21	347.27	367.02
REVENUE SERVICES	432.0	354.85	274.83
SENIORS SERVICES AND LONGTERM CARE	391.2	475.11	414.44
SHELTER SUPPORT AND HOUSING ADMINISTRATION	559.52	713.12	1081.76
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	241.16	244.13	200.24
SOLID WASTE MANAGEMENT SERVICES	194.14	177.8	176.02
TECHNOLOGY SERVICES	200.61	180.65	415.26
TORONTO BUILDING	849.65	896.75	811.93
TORONTO FIRE SERVICES	253.5	404.16	438.78
TORONTO PARAMEDIC SERVICES	301.18	134.98	178.94
TORONTO PUBLIC HEALTH	348.17	360.59	469.13
TORONTO WATER	197.55	200.95	203.23
TRANSPORTATION SERVICES	198.41	200.45	210.9

```
[ ]: quarterly_pivot = quarterly_expenses.pivot_table(index='division',
↳columns='transaction_quarter', values='transaction_amt', aggfunc='mean').
↳fillna("")
print("\nQuarterly Average by Division:")
display(quarterly_pivot)
print("")
```

Quarterly Average by Division:

transaction_quarter	1	2	3 \
division			
ACCOUNTING SERVICES	520.19	836.7	603.19
CHILDRENS SERVICES	685.86	498.79	743.26
CITY CLERKS OFFICE	376.32	511.16	481.95
CITY MANAGERS OFFICE	225.1	108.72	203.04
CITY PLANNING	492.23	461.08	329.75
CORPORATE REAL ESTATE MANAGEMENT	192.8	207.02	197.6
COURT SERVICES	1862.57	500.64	
CUSTOMER EXPERIENCE (311 TORONTO)	142.13	99.37	285.37
ECONOMIC DEVELOPMENT AND CULTURE	357.06	329.74	393.4
ENGINEERING AND CONSTRUCTION SERVICES	49.67	62.12	51.24
ENVIRONMENT AND CLIMATE	434.72	356.91	492.11
FLEET SERVICES	2446.58	1318.36	4517.54
INTERNAL AUDIT			293.79
LEGAL SERVICES	91.15	89.44	93.77

MUNICIPAL LICENSING AND STANDARDS	212.95	240.2	122.89
OFFICE OF EMERGENCY MANAGEMENT	223.77	268.73	384.24
OFFICE OF THE DEPUTY CITY MANAGER	197.02	345.34	604.3
PARKS FORESTRY AND RECREATION	179.99	198.48	218.19
PENSION PAYROLL AND EMPLOYEE BENEFITS	934.89	467.97	207.39
PEOPLE & EQUITY	115.77	578.79	-374.81
POLICY PLANNING FINANCE AND ADMINISTRATION	411.87	714.78	325.45
PURCHASING AND MATERIALS MANAGEMENT	377.99	429.8	566.35
REVENUE SERVICES	295.76	790.19	326.42
SENIORS SERVICES AND LONGTERM CARE	433.82	347.85	372.53
SHELTER SUPPORT AND HOUSING ADMINISTRATION	1003.47	808.83	506.51
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	259.29	216.34	242.86
SOLID WASTE MANAGEMENT SERVICES	219.55	227.34	168.78
TECHNOLOGY SERVICES	306.46	417.22	125.18
TORONTO BUILDING	541.41	492.47	640.39
TORONTO FIRE SERVICES	268.31	281.89	271.69
TORONTO PARAMEDIC SERVICES	485.18	225.67	235.84
TORONTO PUBLIC HEALTH	375.75	385.41	407.51
TORONTO WATER	206.18	201.09	205.18
TRANSPORTATION SERVICES	177.0	253.37	217.84

transaction_quarter 4

division

ACCOUNTING SERVICES	617.94
CHILDRENS SERVICES	1126.92
CITY CLERKS OFFICE	459.62
CITY MANAGERS OFFICE	295.07
CITY PLANNING	832.13
CORPORATE REAL ESTATE MANAGEMENT	201.04
COURT SERVICES	676.08
CUSTOMER EXPERIENCE (311 TORONTO)	196.56
ECONOMIC DEVELOPMENT AND CULTURE	390.33
ENGINEERING AND CONSTRUCTION SERVICES	63.12
ENVIRONMENT AND CLIMATE	526.2
FLEET SERVICES	2962.45
INTERNAL AUDIT	
LEGAL SERVICES	104.63
MUNICIPAL LICENSING AND STANDARDS	180.87
OFFICE OF EMERGENCY MANAGEMENT	1033.47
OFFICE OF THE DEPUTY CITY MANAGER	485.64
PARKS FORESTRY AND RECREATION	199.83
PENSION PAYROLL AND EMPLOYEE BENEFITS	
PEOPLE & EQUITY	1087.27
POLICY PLANNING FINANCE AND ADMINISTRATION	287.76
PURCHASING AND MATERIALS MANAGEMENT	389.66
REVENUE SERVICES	356.31
SENIORS SERVICES AND LONGTERM CARE	428.08
SHELTER SUPPORT AND HOUSING ADMINISTRATION	786.02

SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	232.22
SOLID WASTE MANAGEMENT SERVICES	182.42
TECHNOLOGY SERVICES	255.31
TORONTO BUILDING	851.29
TORONTO FIRE SERVICES	359.88
TORONTO PARAMEDIC SERVICES	209.83
TORONTO PUBLIC HEALTH	388.85
TORONTO WATER	200.55
TRANSPORTATION SERVICES	202.3

```
[ ]: yearly_pivot = yearly_expenses.pivot_table(index='division',
    ↳columns='transaction_year', values='transaction_amt', aggfunc='mean').
    ↳fillna("")
print("\nYearly Average by Division:")
display(yearly_pivot)
```

Yearly Average by Division:

transaction_year	2014	2015	2016 \
division			
ACCOUNTING SERVICES		721.84	608.14
CHILDRENS SERVICES		645.63	903.74
CITY CLERKS OFFICE		407.77	439.22
CITY MANAGERS OFFICE		227.75	226.44
CITY PLANNING		645.06	376.63
CORPORATE REAL ESTATE MANAGEMENT	92.01	183.63	200.87
COURT SERVICES		473.39	
CUSTOMER EXPERIENCE (311 TORONTO)			
ECONOMIC DEVELOPMENT AND CULTURE	88.91	345.61	334.61
ENGINEERING AND CONSTRUCTION SERVICES		39.08	74.49
ENVIRONMENT AND CLIMATE		418.93	416.12
FLEET SERVICES		2699.61	2756.53
INTERNAL AUDIT			293.79
LEGAL SERVICES		40.54	139.44
MUNICIPAL LICENSING AND STANDARDS		122.88	138.93
OFFICE OF EMERGENCY MANAGEMENT	186.53	163.89	1246.64
OFFICE OF THE DEPUTY CITY MANAGER		372.99	218.84
PARKS FORESTRY AND RECREATION	173.07	190.88	207.44
PENSION PAYROLL AND EMPLOYEE BENEFITS		601.91	540.01
PEOPLE & EQUITY		1199.4	93.02
POLICY PLANNING FINANCE AND ADMINISTRATION		245.67	311.58
PURCHASING AND MATERIALS MANAGEMENT		401.79	586.97
REVENUE SERVICES	-112.86	384.07	338.28
SENIORS SERVICES AND LONGTERM CARE	656.21	450.5	428.88
SHELTER SUPPORT AND HOUSING ADMINISTRATION	499.44	536.08	791.25
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION		223.44	217.65

SOLID WASTE MANAGEMENT SERVICES	70.81	183.37	172.89
TECHNOLOGY SERVICES		317.73	202.39
TORONTO BUILDING		619.11	480.65
TORONTO FIRE SERVICES	386.64	274.9	287.23
TORONTO PARAMEDIC SERVICES		187.66	208.32
TORONTO PUBLIC HEALTH	150.34	370.35	363.57
TORONTO WATER	203.68	199.05	198.68
TRANSPORTATION SERVICES		237.22	221.95

transaction_year	2017	2018
division		
ACCOUNTING SERVICES	798.98	1035.89
CHILDRENS SERVICES	811.99	332.71
CITY CLERKS OFFICE	421.25	606.65
CITY MANAGERS OFFICE	259.83	94.86
CITY PLANNING	447.96	805.94
CORPORATE REAL ESTATE MANAGEMENT	195.49	224.92
COURT SERVICES	2290.34	383.7
CUSTOMER EXPERIENCE (311 TORONTO)	237.55	105.48
ECONOMIC DEVELOPMENT AND CULTURE	418.13	331.4
ENGINEERING AND CONSTRUCTION SERVICES	69.07	
ENVIRONMENT AND CLIMATE	507.7	410.85
FLEET SERVICES	2796.81	2414.68
INTERNAL AUDIT		
LEGAL SERVICES	237.8	270.57
MUNICIPAL LICENSING AND STANDARDS	226.91	328.81
OFFICE OF EMERGENCY MANAGEMENT	470.65	419.51
OFFICE OF THE DEPUTY CITY MANAGER	596.74	213.5
PARKS FORESTRY AND RECREATION	203.06	192.45
PENSION PAYROLL AND EMPLOYEE BENEFITS	234.3	759.92
PEOPLE & EQUITY	423.21	
POLICY PLANNING FINANCE AND ADMINISTRATION	577.68	787.65
PURCHASING AND MATERIALS MANAGEMENT	370.23	492.57
REVENUE SERVICES	323.42	820.49
SENIORS SERVICES AND LONGTERM CARE	317.07	365.83
SHELTER SUPPORT AND HOUSING ADMINISTRATION	859.76	1030.41
SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	264.84	221.07
SOLID WASTE MANAGEMENT SERVICES	217.46	244.72
TECHNOLOGY SERVICES	370.84	312.94
TORONTO BUILDING	786.83	644.51
TORONTO FIRE SERVICES	301.71	322.66
TORONTO PARAMEDIC SERVICES	360.05	559.08
TORONTO PUBLIC HEALTH	413.71	387.69
TORONTO WATER	208.31	209.15
TRANSPORTATION SERVICES	199.54	201.65

The dataset output above shows the yearly average employee expenditures for various divisions within the Toronto City government. Here's an interpretation of sample divisions:

- **Accounting Services:**
 - Expenditures have shown an increasing trend from 2014 to 2017, peaking in 2017 (\$1035.89).
- **Children’s Services:**
 - Expenditures vary with the highest in 2015 (\$903.74) and a significant drop in 2018 (\$332.71).
- **Economic Development and Culture:**
 - Expenditures vary, peaking in 2016 (\$418.13) and showing fluctuations in other years.
- **Engineering and Construction Services:**
 - Expenditures vary, with noticeable amounts in 2015 (\$74.49) and 2016 (\$69.07).
- **Environment and Climate:**
 - Expenditures have fluctuations, with the highest in 2017 (\$507.70) and variations in other years.
- **Fleet Services:**
 - Expenditures show fluctuations, with relatively stable amounts in 2015 to 2017 and a slight decrease in 2018 (\$2414.68).
- **Internal Audit:**
 - Expenditures are present only in 2015 (\$293.79).

And so on, Patterns of growth, decline, and fluctuations in expenditure can be observed across divisions and years.

1.3.3 3. Visualization

- 3.1 Craft a visual representation showcasing the temporal evolution of expenditures by Division.
- 3.2 Focus primarily on transaction amounts but also consider integrating other insightful metrics.

Note: - Combine smaller Divisions for clarity. - Showcase multiple visualizations, if necessary, to portray the data most effectively and reveal significant insights.

3.1 Box Plot Analysis showcasing the temporal evolution of expenditures by Division Y-o-Y.

```
[ ]: import textwrap

# Get the unique transaction years from the 'transaction_year' column
unique_years = combined_pcard['transaction_year'].unique()
unique_years.sort()

# Get an ordered list of divisions and wrap the text
ordered_divisions = combined_pcard['division'].sort_values().unique()
wrapped_divisions = ['\n'.join(textwrap.wrap(div, width=15)) for div in
↳ ordered_divisions]

# Create a figure and axes objects for subplots
```

```

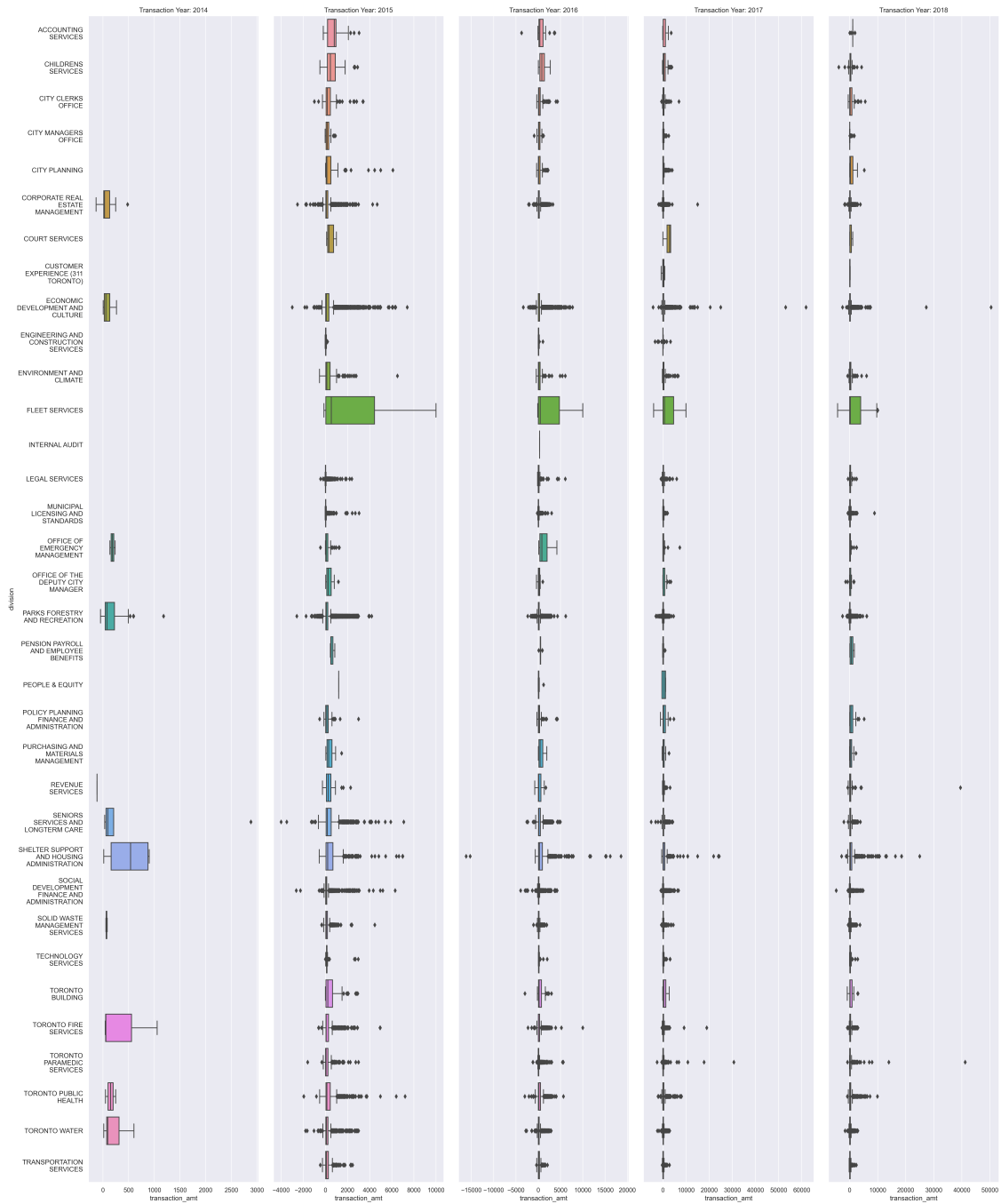
fig, axes = plt.subplots(nrows=1, ncols=len(unique_years),
    figsize=(5*len(unique_years), 30))

# For each year, create a boxplot
for idx, year in enumerate(unique_years):
    yearly_data = combined_pcard[combined_pcard['transaction_year'] == year]
    sns.boxplot(x="transaction_amt", y="division", data=yearly_data,
    ax=axes[idx], order=ordered_divisions)
    axes[idx].set_title(f'Transaction Year: {year}')

    # Remove y-axis labels for plots after the first one and adjust font size
    if idx > 0:
        axes[idx].set_yticklabels([])
        axes[idx].set_ylabel('')
    else:
        axes[idx].set_yticklabels(wrapped_divisions, fontsize=12)

plt.tight_layout()
plt.show()

```



```
[ ]: #3.1.1
#Extract box-plot statistics for further analysis
# Adjust display setting
pd.set_option('display.max_rows', None)
pd.set_option('display.expand_frame_repr', False)
```

```

# List to store the results
results = []

# Unique years and divisions
unique_years = combined_pcard['transaction_year'].unique()
unique_years.sort()
ordered_divisions = combined_pcard['division'].sort_values().unique()

# Extract box plot data for each year and division
for year in unique_years:
    for division in ordered_divisions:
        subset = combined_pcard[(combined_pcard['transaction_year'] == year) &
        ↪(combined_pcard['division'] == division)]

        # Compute the required statistics
        min_val = subset['transaction_amt'].min()
        q1 = subset['transaction_amt'].quantile(0.25)
        median = subset['transaction_amt'].median()
        q3 = subset['transaction_amt'].quantile(0.75)
        max_val = subset['transaction_amt'].max()

        results.append([year, division, min_val, q1, median, q3, max_val])

# Convert to a DataFrame
df_boxplot_data = pd.DataFrame(results, columns=['Year', 'Division', 'Minimum_
↪Expenditures by Division', 'Q1', 'Median Expenditures by Division',
                                                'Q3', 'Maximum Expenditures_
↪by Division'])

# # Pivot the DataFrame with multi-level columns
pivoted = df_boxplot_data.set_index(['Year', 'Division']).unstack('Division')

# Formatting function for currency
def format_currency(val):
    if pd.notna(val):
        try:
            numeric_val = float(val)
            return "{:,.2f}".format(numeric_val)
        except ValueError:
            return str(val) # If not numeric, return the original value as a
↪string
    else:
        return ""

# Apply the formatting function across the DataFrame
formatted_pivoted = pivoted.applymap(format_currency)

```

formatted_pivoted.T

[]: Year					
2014	2015	2016	2017	2018	
Division					
Minimum Expenditures by Division ACCOUNTING SERVICES					
-199.86	-3,744.30	-35.00	90.46		
CHILDRENS SERVICES					
-499.00	21.30	-165.00	-3,842.00		
CITY CLERKS OFFICE					
-1,000.00	-317.76	-584.25	-475.00		
CITY MANAGERS OFFICE					
-22.60	-899.54	-22.60	-22.60		
CITY PLANNING					
8.00	-371.52	-14.53	8.00		
CORPORATE REAL ESTATE MANAGEMENT					
-131.02	-2,527.71	-2,163.00	-1,836.25	-1,798.27	
COURT SERVICES					
103.88		19.66	31.63		
CUSTOMER EXPERIENCE (311 TORONTO)					
-734.50	18.59				
ECONOMIC DEVELOPMENT AND CULTURE					
6.10	-3,000.00	-3,338.87	-4,234.54	-2,486.00	
ENGINEERING AND CONSTRUCTION SERVICES					
-20.00	1.00	-3,334.26			
ENVIRONMENT AND CLIMATE					
-540.00	-446.35	-231.28	-669.53		
FLEET SERVICES					
-128.00	-130.86	-3,986.00	-4,249.25		
INTERNAL AUDIT					
293.79					
LEGAL SERVICES					
-440.70	-141.25	-785.35	-542.40		
MUNICIPAL LICENSING AND STANDARDS					
-30.00	-209.94	-35.46	-599.88		
OFFICE OF EMERGENCY MANAGEMENT					
135.60	-446.80	112.99	7.33	24.39	
OFFICE OF THE DEPUTY CITY MANAGER					
21.77	-395.55	-11.30	-1,400.00		
PARKS FORESTRY AND RECREATION					
-46.15	-2,587.70	-2,332.32	-3,000.00	-2,484.00	
PENSION PAYROLL AND EMPLOYEE BENEFITS					
450.87	111.87	-89.27	84.75		
PEOPLE & EQUITY					
1,199.40	-12.77	-399.80			
POLICY PLANNING FINANCE AND ADMINISTRATION					

-524.42	-310.98	-1,017.00	13.57	
			PURCHASING AND MATERIALS MANAGEMENT	
33.88	33.88	-318.66	7.62	
			REVENUE SERVICES	
-112.86	-271.06	-777.02	-177.47	-627.77
			SENIORS SERVICES AND LONGTERM CARE	
35.64	-4,000.00	-2,547.19	-5,022.49	-2,009.13
			SHELTER SUPPORT AND HOUSING ADMINISTRATION	
16.03	-560.98	-16,150.00	-511.78	-2,830.00
			SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION	
-2,639.12	-3,949.35	-847.50	-4,746.00	
			SOLID WASTE MANAGEMENT SERVICES	
57.57	-337.87	-1,050.90	-640.71	-503.91
			TECHNOLOGY SERVICES	
2.12	28.74	-62.09	-4.99	
			TORONTO BUILDING	
0.55	-2,970.67	-44.09	-852.60	
			TORONTO FIRE SERVICES	
45.62	-622.68	-2,272.69	-395.50	-768.01
			TORONTO PARAMEDIC SERVICES	
-1,606.46	-1,205.71	-2,500.00	-716.99	
			TORONTO PUBLIC HEALTH	
50.68	-1,950.00	-3,000.00	-2,059.42	-795.00
			TORONTO WATER	
17.76	-1,790.10	-2,807.00	-2,339.10	-1,578.09
			TRANSPORTATION SERVICES	
-455.82	-398.85	-292.25	-168.37	
Q1			ACCOUNTING SERVICES	
194.79	147.13	151.26	1,107.40	
			CHILDRENS SERVICES	
186.21	367.25	132.46	120.00	
			CITY CLERKS OFFICE	
66.32	57.34	56.74	54.34	
			CITY MANAGERS OFFICE	
60.55	49.88	83.66	34.61	
			CITY PLANNING	
55.51	39.60	36.70	84.18	
			CORPORATE REAL ESTATE MANAGEMENT	
16.54	33.90	38.03	33.81	34.58
			COURT SERVICES	
208.41		1,784.66	66.50	
			CUSTOMER EXPERIENCE (311 TORONTO)	
-57.22	29.50			
			ECONOMIC DEVELOPMENT AND CULTURE	
26.12	26.77	22.86	26.32	20.69
			ENGINEERING AND CONSTRUCTION SERVICES	
9.00	30.35	30.55		

			ENVIRONMENT AND CLIMATE
37.55	36.76	51.76	53.58
			FLEET SERVICES
30.00	45.09	57.00	50.00
			INTERNAL AUDIT
293.79			
			LEGAL SERVICES
3.50	11.50	56.50	95.62
			MUNICIPAL LICENSING AND STANDARDS
10.17	11.78	56.98	35.22
			OFFICE OF EMERGENCY MANAGEMENT
161.06	20.05	357.01	24.86 39.30
			OFFICE OF THE DEPUTY CITY MANAGER
120.26	39.02	77.83	24.20
			PARKS FORESTRY AND RECREATION
44.72	33.41	33.90	33.49 31.57
			PENSION PAYROLL AND EMPLOYEE BENEFITS
479.12	450.87	36.05	190.69
			PEOPLE & EQUITY
1,199.40	20.13	-374.81	
			POLICY PLANNING FINANCE AND ADMINISTRATION
22.60	22.60	91.12	86.16
			PURCHASING AND MATERIALS MANAGEMENT
151.22	141.38	57.90	55.28
			REVENUE SERVICES
-112.86	87.27	45.18	39.53 46.39
			SENIORS SERVICES AND LONGTERM CARE
60.00	58.46	56.42	49.78 60.24
			SHELTER SUPPORT AND HOUSING ADMINISTRATION
160.91	50.85	72.19	69.13 92.64
			SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION
19.02	22.41	22.00	22.67
			SOLID WASTE MANAGEMENT SERVICES
64.19	38.94	39.73	49.64 43.98
			TECHNOLOGY SERVICES
90.65	74.09	61.77	67.39
			TORONTO BUILDING
44.99	74.66	152.53	70.09
			TORONTO FIRE SERVICES
52.42	50.31	46.33	54.91 56.50
			TORONTO PARAMEDIC SERVICES
15.00	16.00	15.00	20.41
			TORONTO PUBLIC HEALTH
100.51	45.20	39.25	39.00 36.16
			TORONTO WATER
73.81	41.79	41.97	40.64 41.18
			TRANSPORTATION SERVICES

22.58	24.64	23.04	34.30
Median	Expenditures	by Division	ACCOUNTING SERVICES
804.19	367.22	1,084.80	1,107.40
			CHILDRENS SERVICES
447.68	819.25	471.77	187.57
			CITY CLERKS OFFICE
186.71	188.06	142.98	223.36
			CITY MANAGERS OFFICE
155.12	150.00	101.39	51.78
			CITY PLANNING
139.13	113.00	118.65	337.87
			CORPORATE REAL ESTATE MANAGEMENT
32.32	79.10	90.32	76.39 86.70
			COURT SERVICES
317.04		2,878.68	148.31
			CUSTOMER EXPERIENCE (311 TORONTO)
250.00	45.18		
			ECONOMIC DEVELOPMENT AND CULTURE
57.86	79.10	75.77	87.27 73.43
			ENGINEERING AND CONSTRUCTION SERVICES
17.00	32.45	31.60	
			ENVIRONMENT AND CLIMATE
120.00	158.66	186.61	190.97
			FLEET SERVICES
539.00	457.00	695.09	272.00
			INTERNAL AUDIT
293.79			
			LEGAL SERVICES
9.50	60.00	172.50	175.00
			MUNICIPAL LICENSING AND STANDARDS
14.76	29.97	100.00	95.00
			OFFICE OF EMERGENCY MANAGEMENT
186.53	72.52	819.79	136.49 106.62
			OFFICE OF THE DEPUTY CITY MANAGER
250.30	151.82	173.74	90.85
			PARKS FORESTRY AND RECREATION
79.08	82.50	83.83	82.83 75.11
			PENSION PAYROLL AND EMPLOYEE BENEFITS
507.37	502.88	70.61	629.98
			PEOPLE & EQUITY
1,199.40	51.69	504.38	
			POLICY PLANNING FINANCE AND ADMINISTRATION
117.29	135.60	294.96	470.85
			PURCHASING AND MATERIALS MANAGEMENT
293.82	422.77	188.12	347.52
			REVENUE SERVICES
-112.86	248.60	111.33	132.23 121.38

94.29	174.80	164.30	SENIORS SERVICES AND LONGTERM CARE 135.54 152.94
539.62	203.09	279.11	SHELTER SUPPORT AND HOUSING ADMINISTRATION 255.38 319.76
44.29	52.98	54.84	SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION 52.67
70.81	84.80	80.34	SOLID WASTE MANAGEMENT SERVICES 95.00 100.00
110.25	136.70	143.18	TECHNOLOGY SERVICES 168.57
214.07	199.44	440.00	TORONTO BUILDING 153.69
59.23	73.45	65.19	TORONTO FIRE SERVICES 90.40 97.11
57.82	57.00	65.87	TORONTO PARAMEDIC SERVICES 87.02
150.34	134.61	150.20	TORONTO PUBLIC HEALTH 145.00 150.22
96.14	100.52	108.55	TORONTO WATER 103.87 103.30
101.69	101.70	96.03	TRANSPORTATION SERVICES 95.31
Q3			ACCOUNTING SERVICES
961.90	1,084.80	1,084.80	1,107.40
896.87	1,400.00	1,024.34	CHILDRENS SERVICES 450.94
446.35	452.00	448.91	CITY CLERKS OFFICE 846.55
305.25	378.64	279.75	CITY MANAGERS OFFICE 75.50
499.57	438.44	296.53	CITY PLANNING 1,200.86
132.02	213.59	245.33	CORPORATE REAL ESTATE MANAGEMENT 214.30 243.99
737.98		3,384.35	COURT SERVICES 632.79
495.97	186.19		CUSTOMER EXPERIENCE (311 TORONTO)
134.71	312.20	309.95	ECONOMIC DEVELOPMENT AND CULTURE 363.24 264.38
45.00	119.45	93.22	ENGINEERING AND CONSTRUCTION SERVICES
426.65	426.16	455.99	ENVIRONMENT AND CLIMATE 478.79
4,440.00	4,746.00	4,624.81	FLEET SERVICES 3,936.00
			INTERNAL AUDIT

293.79				
				LEGAL SERVICES
27.00	169.50	325.00	381.38	
				MUNICIPAL LICENSING AND STANDARDS
41.14	100.00	253.31	250.00	
				OFFICE OF EMERGENCY MANAGEMENT
212.00	223.20	1,991.52	341.69 323.86	
				OFFICE OF THE DEPUTY CITY MANAGER
521.11	333.78	819.25	399.10	
				PARKS FORESTRY AND RECREATION
226.00	216.43	225.72	222.82 198.32	
				PENSION PAYROLL AND EMPLOYEE BENEFITS
677.43	551.36	231.58	1,199.21	
				PEOPLE & EQUITY
1,199.40	143.08	1,180.64		
				POLICY PLANNING FINANCE AND ADMINISTRATION
254.16	310.98	1,084.80	1,107.40	
				PURCHASING AND MATERIALS MANAGEMENT
593.99	1,028.40	573.77	739.79	
				REVENUE SERVICES
-112.86	498.72	610.20	354.84 426.25	
				SENIORS SERVICES AND LONGTERM CARE
209.61	523.07	483.95	366.30 391.58	
				SHELTER SUPPORT AND HOUSING ADMINISTRATION
878.14	678.68	922.08	794.67 792.90	
				SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION
123.97	145.94	156.45	150.38	
				SOLID WASTE MANAGEMENT SERVICES
77.43	180.19	179.67	225.99 250.36	
				TECHNOLOGY SERVICES
145.66	187.96	270.52	309.25	
				TORONTO BUILDING
638.02	727.61	1,275.76	904.88	
				TORONTO FIRE SERVICES
557.15	276.85	290.33	283.79 352.91	
				TORONTO PARAMEDIC SERVICES
224.44	130.00	219.76	291.85	
				TORONTO PUBLIC HEALTH
200.17	438.15	502.14	432.40 425.00	
				TORONTO WATER
315.53	226.04	227.47	233.15 248.59	
				TRANSPORTATION SERVICES
268.93	268.05	223.74	225.99	
Maximum Expenditures by Division				ACCOUNTING SERVICES
3,048.46	3,744.30	3,610.35	1,929.99	
				CHILDRENS SERVICES
2,909.75	2,720.00	3,904.15	4,294.00	

			CITY CLERKS OFFICE		
3,424.15	4,334.01	6,942.70	5,554.02		
			CITY MANAGERS OFFICE		
908.42	1,207.39	2,541.79	1,704.66		
			CITY PLANNING		
6,107.56	2,226.00	4,044.65	5,256.79		
			CORPORATE REAL ESTATE MANAGEMENT		
482.40	4,677.30	3,312.60	15,010.37	3,842.00	
			COURT SERVICES		
1,000.00		3,384.35	1,107.40		
			CUSTOMER EXPERIENCE (311 TORONTO)		
827.14	243.18				
			ECONOMIC DEVELOPMENT AND CULTURE		
265.59	7,400.00	7,694.42	61,842.37	50,413.79	
			ENGINEERING AND CONSTRUCTION SERVICES		
198.10	1,054.64	3,334.26			
			ENVIRONMENT AND CLIMATE		
6,515.02	6,075.78	6,667.00	6,073.75		
			FLEET SERVICES		
10,000.00	10,000.00	10,000.00	10,000.00		
			INTERNAL AUDIT		
293.79					
			LEGAL SERVICES		
2,395.60	6,068.52	5,987.18	2,521.30		
			MUNICIPAL LICENSING AND STANDARDS		
3,046.51	2,997.89	2,034.46	8,859.20		
			OFFICE OF EMERGENCY MANAGEMENT		
237.46	1,265.60	4,187.78	7,289.63	2,509.11	
			OFFICE OF THE DEPUTY CITY MANAGER		
1,173.55	1,030.00	3,467.84	1,600.00		
			PARKS FORESTRY AND RECREATION		
1,180.19	4,186.65	6,158.75	4,622.95	6,162.12	
			PENSION PAYROLL AND EMPLOYEE BENEFITS		
847.50	932.25	1,015.87	1,695.00		
			PEOPLE & EQUITY		
1,199.40	1,199.40	1,199.40			
			POLICY PLANNING FINANCE AND ADMINISTRATION		
3,000.00	4,294.00	4,746.90	5,210.00		
			PURCHASING AND MATERIALS MANAGEMENT		
1,467.87	1,919.20	2,687.12	2,284.86		
			REVENUE SERVICES		
-112.86	2,282.60	1,717.23	3,164.00	39,493.50	
			SENIORS SERVICES AND LONGTERM CARE		
2,881.50	7,079.45	4,969.51	4,282.70	3,836.35	
			SHELTER SUPPORT AND HOUSING ADMINISTRATION		
902.48	6,999.28	18,573.46	24,343.75	25,000.00	
			SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION		

6,305.00	4,237.50	6,780.00	4,853.35	
				SOLID WASTE MANAGEMENT SERVICES
84.05	4,469.57	1,937.30	4,514.35	3,692.78
				TECHNOLOGY SERVICES
2,966.16	2,017.05	3,164.00	2,904.89	
				TORONTO BUILDING
2,919.58	2,994.50	2,745.90	2,998.00	
				TORONTO FIRE SERVICES
1,055.08	4,968.00	10,000.00	18,958.50	2,999.44
				TORONTO PARAMEDIC SERVICES
3,000.00	5,619.00	30,650.50	41,205.75	
				TORONTO PUBLIC HEALTH
250.00	7,211.75	5,671.38	8,085.15	9,945.70
				TORONTO WATER
603.88	3,000.00	2,928.09	2,959.72	2,932.35
				TRANSPORTATION SERVICES
2,467.02	2,092.97	2,841.09	2,431.66	

```
[ ]: # combined_pcard[(combined_pcard['division'] == 'SHELTER SUPPORT AND HOUSING_
↳ADMINISTRATION') & (combined_pcard['transaction_year'] == 2016)].
↳sort_values(by='transaction_amt')
```

3.1.2 Interpretation of the Box plot statistics

- The data shows a dynamic financial landscape with various divisions undergoing changes in expenditures Year on Year. Some divisions show consistent trends(E.g CITY MANAGER'S OFFICE), while others fluctuate significantly.(E.g. ACCOUNTING SERVICES)
- **Comparative Expenditure:** Divisions such as TECHNOLOGY SERVICES and TORONTO WATER have similar ranges of expenditures across the years.
- **Missing Data** Some divisions have missing data for certain years, such as CUSTOMER EXPERIENCE (311 TORONTO). It's important to understand if these divisions were non-operational in those years or if data is simply unavailable.
- **Division Variability** Some divisions have relatively stable expenditures over the years, such as FLEET SERVICES. This consistency might indicate a capped expenditure or a budget that's rigorously adhered to. In contrast, other divisions like TORONTO PARAMEDIC SERVICES show significant fluctuation, with the maximum expenditure jumping from \$3,000 to over \$41,000 within the period. Also, ECONOMIC DEVELOPMENT AND CULTURE, show a dramatic increase in their maximum expenditures, from a mere \$265 in one year to a whopping \$61,842 in another.
- **Potential Anomalies** Many divisions have negative expenditures, understanding the exact implications would require more context on the G/L accounts and Purpose of purchase. E.g. The SHELTER SUPPORT AND HOUSING ADMINISTRATION division shows significant negative expenditures in 2016 of -\$16,150 and -\$15,246 as a result of refunds for gifts from clients during LD. These negative values requires further investigation with the government on such transactions. On the other end SHELTER SUPPORT AND HOUSING ADMINISTRATION has been

increasing their maximum expenditures year over year, potentially indicating an expanding scope of services or higher costs among other divisions.

- While some divisions have escalating costs, others are maintaining or reducing their maximum expenditures. The reasons for these could be multifaceted, ranging from inflation, increased operational scope, cost-saving measures, economic events affecting the city, or changes in budget allocations.

A deep dive of the specific causes for these expenditure trends, can reveal the factors at play. For a more in-depth understanding and actionable insights, it would be beneficial to collaborate with a financial analyst familiar with the city's budgeting and finance operations or interrogate the data further for more insights.

1.3.4 4. Cluster Creation

4.1 Group Divisions with similar behaviours based on the profiles developed in the prior step.

- For clustering we will create new features for our baseline model. We will use KMeans clustering algorithm with silhouette score for measuring the optimal number of clusters we should be having for our new dataframe(`division_features`).
- Our new data set will comprise of
 - `division`
 - unique counts of `batch_transaction_id`, `gl_accounts`, `merchant_name` and `purpose`
 - mean, sum, standard deviation, minimum, maximum, median & IQR of the transaction amounts
- This will help us have a better understanding of the data before we can recommend an improved model.

```
[ ]: from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
import matplotlib.pyplot as plt

# Create features for each division
division_features = combined_pcard.groupby(['division']).agg({
    'batch_transaction_id': 'nunique',
    'transaction_amt': ['mean', 'sum', 'std', 'min', 'max', 'median',
                        lambda x: x.quantile(0.25), lambda x: x.quantile(0.75)],
    'gl_account': 'nunique',
    'merchant_name': 'nunique',
    'purpose': 'nunique'
}).reset_index()

# Update the columns names after aggregation
division_features.columns = ["_".join(x) for x in division_features.columns.
                             ravel()]
division_features = division_features.rename(columns={'division_': 'division',
```

```

↳ 'transaction_amt_<lambda_0>': 'transaction_amt_q1',
↳ 'transaction_amt_<lambda_1>': 'transaction_amt_q3'})

# Compute the IQR
division_features['transaction_amt_iqr'] =
↳ division_features['transaction_amt_q3'] -
↳ division_features['transaction_amt_q1']

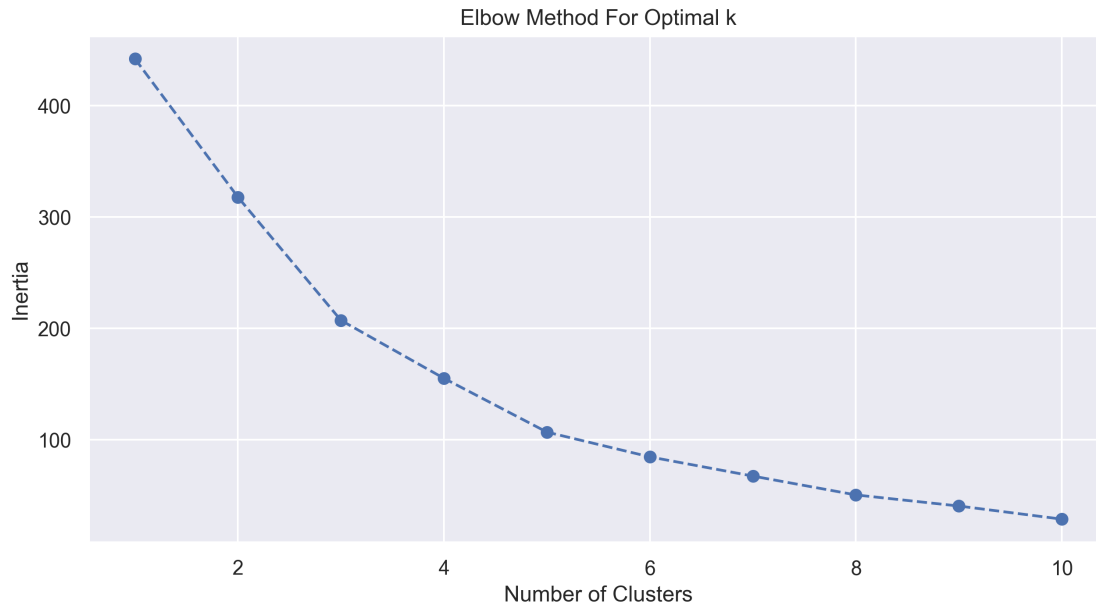
# Handling NaN values
# We are replacing NaN values with the mean of the respective columns
for col in division_features.columns:
    if division_features[col].dtype != 'object': # Avoid filling for
↳ 'division' column
        division_features[col].fillna(division_features[col].mean(),
↳ inplace=True)

# Scale the features (excluding 'division')
scaler = StandardScaler()
scaled_features = scaler.fit_transform(division_features.drop(['division'],
↳ axis=1))

# Elbow method to find optimal number of clusters
inertia = []
for i in range(1, 11): # Testing for up to 10 clusters
    kmeans = KMeans(n_clusters=i, random_state=0)
    kmeans.fit(scaled_features)
    inertia.append(kmeans.inertia_)

# Plotting the elbow curve
plt.figure(figsize=(10, 5))
plt.plot(range(1, 11), inertia, marker='o', linestyle='--')
plt.title('Elbow Method For Optimal k')
plt.xlabel('Number of Clusters')
plt.ylabel('Inertia')
plt.show()

```

```
[ ]: from sklearn.metrics import silhouette_score

# Define the range of clusters
cluster_range = range(4, 15)

# List to store silhouette scores
silhouette_scores = []

# Calculate silhouette scores for each number of clusters
for n_clusters in cluster_range:
    kmeans = KMeans(n_clusters=n_clusters, random_state=42)
    cluster_assignments = kmeans.fit_predict(scaled_features)
    silhouette_avg = silhouette_score(scaled_features, cluster_assignments)
    silhouette_scores.append(silhouette_avg)

# Create a DataFrame to store the results
df_silhouette = pd.DataFrame({
    'Number of Clusters': cluster_range,
    'Silhouette Score': silhouette_scores
})

# Display the DataFrame
df_silhouette

def highlight_max(s):
    """
    Highlight the maximum in a Series yellow.
```

```

"""
is_max = s == s.max()
return ['background-color: yellow' if v else '' for v in is_max]

# Apply the styling function to the DataFrame
styled_df = df_silhouette.style.apply(highlight_max, subset=['Silhouette_
↳Score'])

# Display the styled DataFrame
styled_df

```

```
[ ]: <pandas.io.formats.style.Styler at 0x7f9b029e56c0>
```

```

[ ]: # Based on the above plot results, we can see that the mean silhouette_
↳Coefficient is highest at 5 clusters (0.3630),
# indicating improved model performance and better cluster assignment for each_
↳dataset instance.

kmeans = KMeans(n_clusters=5,random_state = 42)
division_features['cluster'] = kmeans.fit_predict(scaled_features)

division_features.sort_values(by='cluster').applymap(format_currency)

```

```

[ ]:
division batch_transaction_id_nunique
transaction_amt_mean transaction_amt_sum transaction_amt_std transaction_amt_min
transaction_amt_max transaction_amt_median transaction_amt_q1 transaction_amt_q3
gl_account_nunique merchant_name_nunique purpose_nunique transaction_amt_iqr
cluster
30 TORONTO PARAMEDIC SERVICES 2,074.00
295.32 612,783.56 1,371.50 -2,500.00
41,205.75 62.28 16.00 210.76
77.00 568.00 1,032.00 194.76 0.00
8 ECONOMIC DEVELOPMENT AND CULTURE 14,846.00
363.97 5,403,507.87 1,164.00 -4,234.54
61,842.37 79.63 24.85 316.40
147.00 4,174.00 8,704.00 291.55 0.00
24 SHELTER SUPPORT AND HOUSING ADMINISTRATION 3,085.00
805.01 2,483,463.91 2,090.86 -16,150.00
25,000.00 265.21 66.67 745.39
131.00 749.00 2,546.00 678.72 0.00
22 REVENUE SERVICES 240.00
502.82 120,676.44 2,593.37 -777.02
39,493.50 131.06 48.79 408.30
40.00 98.00 213.00 359.51 0.00
17 PARKS FORESTRY AND RECREATION 98,337.00
199.38 19,606,012.25 341.98 -3,000.00
6,162.12 81.81 33.48 218.47

```

219.00	5,743.00	55,494.00	184.99	1.00
16	OFFICE OF THE DEPUTY CITY MANAGER			114.00
328.11	37,404.59	570.55	-1,400.00	
3,467.84	143.69	29.06	446.78	
28.00	76.00	106.00	417.72	2.00
31	TORONTO PUBLIC HEALTH			4,936.00
387.27	1,911,541.00	733.21	-3,000.00	
9,945.70	145.00	39.50	447.48	
106.00	1,619.00	975.00	407.98	2.00
29	TORONTO FIRE SERVICES			4,319.00
293.61	1,268,108.36	613.57	-2,272.69	
18,958.50	77.20	50.59	282.79	
95.00	959.00	2,280.00	232.20	2.00
28	TORONTO BUILDING			188.00
613.27	115,294.34	857.18	-2,970.67	
2,998.00	247.69	68.47	852.60	
28.00	66.00	155.00	784.13	2.00
27	TECHNOLOGY SERVICES			241.00
312.73	75,367.47	601.50	-62.09	
3,164.00	136.70	74.09	266.58	
18.00	86.00	172.00	192.49	2.00
26	SOLID WASTE MANAGEMENT SERVICES			3,190.00
205.59	655,826.73	355.12	-1,050.90	
4,514.35	90.39	42.94	211.87	
125.00	523.00	2,321.00	168.93	2.00
25	SOCIAL DEVELOPMENT FINANCE AND ADMINISTRATION			5,458.00
235.06	1,282,970.63	620.60	-4,746.00	
6,780.00	50.84	21.45	144.99	
64.00	1,153.00	4,206.00	123.54	2.00
23	SENIORS SERVICES AND LONGTERM CARE			5,298.00
396.27	2,099,412.19	654.44	-5,022.49	
7,079.45	156.50	54.56	449.65	
83.00	1,034.00	3,677.00	395.09	2.00
21	PURCHASING AND MATERIALS MANAGEMENT			110.00
448.65	49,351.13	502.46	-318.66	
2,687.12	286.17	71.67	727.17	
23.00	63.00	105.00	655.50	2.00
20	POLICY PLANNING FINANCE AND ADMINISTRATION			336.00
500.14	168,046.99	757.78	-1,017.00	
5,210.00	210.87	47.04	820.31	
24.00	116.00	144.00	773.27	2.00
19	PEOPLE & EQUITY			112.00
201.52	22,570.71	386.83	-399.80	
1,199.40	65.55	20.04	179.14	
4.00	60.00	13.00	159.09	2.00
33	TRANSPORTATION SERVICES			1,397.00
214.00	298,957.10	341.93	-455.82	

2,841.09	98.50	25.09	242.90
101.00	330.00	910.00	217.81 2.00
15	OFFICE OF EMERGENCY MANAGEMENT		151.00
372.25	56,209.00	835.95	-446.80
7,289.63	92.94	30.00	337.85
41.00	94.00	111.00	307.85 2.00
2	CITY CLERKS OFFICE		587.00
456.69	268,076.01	769.35	-1,000.00
6,942.70	187.45	59.76	538.48
49.00	294.00	475.00	478.72 2.00
3	CITY MANAGERS OFFICE		455.00
175.43	79,819.24	292.87	-899.54
2,541.79	76.32	39.00	198.18
21.00	331.00	250.00	159.18 2.00
4	CITY PLANNING		350.00
508.32	177,912.29	875.40	-371.52
6,107.56	129.94	46.45	552.62
17.00	97.00	278.00	506.18 2.00
5	CORPORATE REAL ESTATE MANAGEMENT		14,700.00
199.47	2,935,159.62	357.30	-2,527.71
15,010.37	83.87	35.00	227.99
99.00	1,179.00	6,702.00	192.99 2.00
32	TORONTO WATER		11,709.00
203.36	2,381,143.08	334.34	-2,807.00
3,000.00	104.37	41.32	230.57
138.00	1,334.00	7,046.00	189.25 2.00
9	ENGINEERING AND CONSTRUCTION SERVICES		1,426.00
55.53	79,184.69	217.06	-3,334.26
3,334.26	30.55	10.00	63.90
9.00	19.00	604.00	53.90 2.00
7	CUSTOMER EXPERIENCE (311 TORONTO)		20.00
191.32	3,826.44	363.13	-734.50
827.14	169.34	25.63	313.75
6.00	13.00	20.00	288.11 2.00
13	LEGAL SERVICES		5,821.00
93.71	545,494.58	251.78	-785.35
6,068.52	17.00	6.50	100.00
18.00	116.00	210.00	93.50 2.00
14	MUNICIPAL LICENSING AND STANDARDS		904.00
193.70	175,101.84	480.15	-599.88
8,859.20	50.00	13.87	167.14
77.00	270.00	403.00	153.27 2.00
10	ENVIRONMENT AND CLIMATE		701.00
442.94	310,498.49	875.30	-669.53
6,667.00	169.50	44.35	454.88
45.00	335.00	429.00	410.53 2.00
11	FLEET SERVICES		2,005.00

2,711.51	5,436,574.33	3,693.09	-4,249.25	
10,000.00	491.00	55.00	4,543.25	
37.00	97.00	388.00	4,488.25	3.00
6		COURT SERVICES		18.00
842.28	15,160.97	1,102.37	19.66	
3,384.35	263.05	114.99	1,080.55	
10.00	14.00	17.00	965.56	4.00
12		INTERNAL AUDIT		1.00
293.79	293.79	818.42	293.79	
293.79	293.79	293.79	293.79	
1.00	1.00	1.00	0.00	4.00
18	PENSION PAYROLL AND EMPLOYEE BENEFITS			28.00
447.04	12,517.14	420.37	-89.27	
1,695.00	450.87	88.14	625.39	
7.00	9.00	20.00	537.25	4.00
1		CHILDRENS SERVICES		345.00
651.46	224,753.04	852.27	-3,842.00	
4,294.00	367.82	140.00	853.15	
13.00	130.00	180.00	713.15	4.00
0		ACCOUNTING SERVICES		218.00
732.83	159,757.66	730.37	-3,744.30	
3,744.30	804.19	194.79	1,084.80	
22.00	59.00	126.00	890.01	4.00

4.2 Describe each cluster and elucidate its unique attributes.

```
[ ]: clusters = division_features['cluster'].nunique()
cluster_descriptions = []

for cluster in range(clusters):
    cluster_data = division_features[division_features['cluster'] == cluster]

    # Fetching basic stats
    mean_transaction_amt = cluster_data['transaction_amt_mean'].mean()
    median_transaction_amt = cluster_data['transaction_amt_median'].mean()
    max_transaction_amt = cluster_data['transaction_amt_max'].max()
    min_transaction_amt = cluster_data['transaction_amt_min'].min()
    num_divisions = cluster_data['division'].nunique()

    description = {
        "Cluster": cluster,
        "Num_Divisions": num_divisions,
        "Avg_Mean_Transaction": mean_transaction_amt,
        "Avg_Median_Transaction": median_transaction_amt,
        "Max_Transaction": max_transaction_amt,
        "Min_Transaction": min_transaction_amt
    }
```

```

cluster_descriptions.append(description)

# Convert list of dictionaries to DataFrame
description_df = pd.DataFrame(cluster_descriptions)

description_df.applymap(format_currency)

```

```

[ ]:  Cluster Num_Divisions Avg_Mean_Transaction Avg_Median_Transaction
Max_Transaction Min_Transaction
0      0.00          4.00          491.78          134.54
61,842.37     -16,150.00
1      1.00          1.00          199.38          81.81
6,162.12     -3,000.00
2      2.00         23.00          305.60          122.63
18,958.50     -5,022.49
3      3.00          1.00         2,711.51          491.00
10,000.00     -4,249.25
4      4.00          5.00          593.48          435.94
4,294.00     -3,842.00

```

Cluster 0 (TORONTO PARAMEDIC SERVICES,ECONOMIC DEVELOPMENT AND CULTURE,SHELTER SUPPORT AND HOUSING ADMINISTRATION,REVENUE SERVICES): - **Divisions:** This cluster comprises 4 divisions. - **Transaction Attributes:** Divisions in this cluster have an average mean transaction of approximately \$491.78 and a median transaction of about \$134.54. The range of transaction amounts in this cluster is quite wide, spanning from -\$16,150.00 to a substantial \$61,842.37. - **Unique Attribute:** Cluster 0 represents divisions with diverse transaction behaviors. These divisions handle a wide range of transaction values, possibly indicating occasional high-value transactions.

Cluster 1(PARKS FORESTRY AND RECREATION): - **Divisions:** This cluster is unique as it contains only 1 division. - **Transaction Attributes:** The mean transaction for this division is around \$199.38, with a median transaction of \$81.81. Transactions in this division range from -\$3,000.00 to \$6,162.12. - **Unique Attribute:** Despite having only one division, this cluster stands out due to the division's modest transaction values. However, the narrower range of transactions compared to other clusters is notable.

Cluster 2 (OTHERS): - **Divisions:** The largest cluster, containing 23 divisions. - **Transaction Attributes:** On average, divisions in this cluster have a mean transaction amount of about \$305.60, with a median transaction of \$122.63. Transaction values within this cluster range from -\$5,022.49 to \$18,958.50. - **Unique Attribute:** Cluster 2, despite having the most divisions, deals with a wide variety of transactions but doesn't have the broadest range of transaction values among the clusters.

Cluster 3(FLEET SERVICES): - **Divisions:** Similar to Cluster 1, Cluster 3 has only 1 division. - **Transaction Attributes:** The mean transaction for this division is notably high at approximately \$2,711.51, with a median transaction of \$491.00. Transactions here span from -\$4,249.25 to \$10,000.00. - **Unique Attribute:** With a solitary division, Cluster 3's standout characteristic is its high mean transaction value, suggesting that this specific division handles larger monetary

transactions.

Cluster 4(COURT SERVICES,INTERNAL AUDIT,PENSION PAYROLL AND EMPLOYEE BENEFITS,CHILDRENS SERVICES,ACCOUNTING SERVICES): - **Divisions:** This cluster comprises 5 divisions. - **Transaction Attributes:** Divisions in Cluster 4 have an average mean transaction amount of around \$593.48. The median transaction for these divisions is approximately \$435.94. Transactions within this cluster range from a minimum of -\$3,842.00 to a maximum of \$4,294.00. - **Unique Attribute:** Cluster 4 showcases divisions with moderate transaction values, representing typical and routine transactions, given the relatively narrow range between the maximum and minimum transaction amounts.

In summary, the analysis of the updated dataset reveals varying transaction behaviors across the different clusters. Clusters 0 and 4, each consisting of multiple divisions, represent more common transaction patterns. Cluster 1 exhibits a broader range of transaction values, while Clusters 2 and 3 highlight unique transaction behaviors of individual divisions.

1.3.5 5. Anomaly Detection

- 5.1 Utilizing the profiles, pinpoint Divisions that manifest anomalous behaviours.
- 5.2 Further subdivide by clusters to highlight any irregularities if beneficial.

```
[ ]: # To identify anomalous behavior in divisions based on the provided data, we'll
    ↪ follow a step-by-step approach:

# 1. Data Standardization: Since different features have different scales,
    ↪ it's essential to standardize the data to ensure equal weightage to all
    ↪ features.

# 2. PCA for Visualization: Principal Component Analysis (PCA) can be used
    ↪ to reduce the dimensionality of the data and visualize it in 2D or 3D space
    ↪ to spot any potential outliers or anomalous behavior.

# 3. IQR Method: We can use the Interquartile Range (IQR) method to
    ↪ identify outliers in the principal components. The IQR is a measure of
    ↪ statistical spread between the 25th percentile (Q1) and the 75th percentile
    ↪ (Q3). Anything outside the range [Q1 - 1.5*IQR, Q3 + 1.5*IQR] can be
    ↪ considered as an anomaly or outlier.

# 4. Analysis and Interpretation: Based on the results, we can then
    ↪ identify divisions that show anomalous behavior.

# Let's walk through the process:

from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
import matplotlib.pyplot as plt
import seaborn as sns
```

```

# Step 1: Data Standardization
features = division_features.drop(['division', 'cluster'], axis=1)
scaled_features = StandardScaler().fit_transform(features)

# Step 2: PCA for Visualization
pca = PCA(n_components=2)
principal_components = pca.fit_transform(scaled_features)
principal_df = pd.DataFrame(data=principal_components, columns=['PC1', 'PC2'])
principal_df['division'] = division_features['division']

# Step 3: IQR Method for anomaly detection
Q1 = principal_df[['PC1', 'PC2']].quantile(0.25)
Q3 = principal_df[['PC1', 'PC2']].quantile(0.75)
IQR = Q3 - Q1
filter = (principal_df[['PC1', 'PC2']] < (Q1 - 1.5 * IQR)) |
        (principal_df[['PC1', 'PC2']] > (Q3 + 1.5 * IQR))

anomalous_divisions_df = principal_df[filter.any(axis=1)]

# Display anomalous divisions
print(anomalous_divisions_df)

# Attach cluster labels from KMeans
principal_df['cluster'] = division_features['cluster']

# Plotting the 2D PCA
plt.figure(figsize=(10, 8))
sns.scatterplot(x='PC1', y='PC2', hue='cluster', data=principal_df,
               palette='Set2', s=100)

# Annotate each anomalous division with its name
for _, row in anomalous_divisions_df.iterrows():
    plt.annotate(row['division'], (row['PC1'], row['PC2']), fontsize=9, alpha=0.
    7, color='black')

plt.title('2D PCA of Clusters')
plt.show()

```

	PC1	PC2	division
8	3.472690	1.950503	ECONOMIC DEVELOPMENT AND CULTURE
11	-5.195409	8.652914	FLEET SERVICES
17	9.493510	4.338936	PARKS FORESTRY AND RECREATION
24	0.135498	3.023488	SHELTER SUPPORT AND HOUSING ADMINISTRATION



5.3 Rank Divisions by their “abnormality”, opting for criteria like the Top 10 Divisions or the Top 10% based on specific metrics. We considered Distance from the Median approach due to it’s simplicity of use:

1. **Distance from the Median:** One common method to rank the abnormality of divisions is to calculate the distance of each point from the median (or mean) of the PCA components. The idea is that the farther a division is from the central tendency of the data, the more abnormal it is.

```
[ ]: # Calculate the median for PC1 and PC2
median_PC1 = principal_df['PC1'].median()
median_PC2 = principal_df['PC2'].median()

# Compute the distance of each division from the median of the PCA components
anomalous_divisions_df['distance_from_median'] = np.
    ↪sqrt((anomalous_divisions_df['PC1'] - median_PC1)**2 +
        ↪(anomalous_divisions_df['PC2'] - median_PC2)**2).round(2)

# Rank the divisions based on the distance
anomalous_divisions_df = anomalous_divisions_df.
    ↪sort_values(by='distance_from_median', ascending=False)
```

```
# Reset index, drop old index, and start counting from 1
anomalous_divisions_df = anomalous_divisions_df[['division',
↪ 'distance_from_median']].reset_index(drop=True)
anomalous_divisions_df.index = anomalous_divisions_df.index + 1

# Display the ranked anomalous divisions
anomalous_divisions_df
```

```
[ ]:
```

	division	distance_from_median
1	PARKS FORESTRY AND RECREATION	10.91
2	FLEET SERVICES	10.36
3	ECONOMIC DEVELOPMENT AND CULTURE	4.48
4	SHELTER SUPPORT AND HOUSING ADMINISTRATION	3.53

5.4 Recommend a strategy to differentiate between “normal” and “abnormal” Divisions through sampling. Using sampling to differentiate between normal and abnormal divisions requires a statistical approach that involves examining the distribution of data and determining whether a division’s characteristics fall within expected ranges. Here’s a strategy we can follow after we have done procedure 1 to 5 above:

1. Define Normal and Abnormal Criteria:

- We need to decide on criteria that define what is considered **normal** and **abnormal** for divisions based on the specific attributes we’re analyzing (e.g., transaction amounts).
- This could involve setting thresholds based on historical data.

2. Sampling:

- Select a sample of divisions from our dataset. The sample size will depend on the size of our dataset and the desired level of confidence.
- we can use random sampling or other sampling methods based on our analysis goals.

3. Compare with Criteria:

- Compare the sample statistics with the predefined criteria for **normal** and **abnormal** divisions.
- Divisions that fall within the expected ranges are considered **normal**, while those that significantly deviate from the criteria may be **abnormal**.

4. Statistical Tests:

- Perform statistical tests if necessary to quantify the differences between sample statistics and criteria.
- For example, we can use Z-tests, T-tests, or other relevant tests to compare means or other attributes.

5. Visualization:

- Visualize the sample data and their distributions to see how they align with our **normal** and **abnormal** criteria.

6. Decision Making:

- Based on our analysis, make decisions about whether each division is considered **normal** or **abnormal** according to the predefined criteria.
- Keep in mind that **abnormal** divisions might need further investigation to understand the underlying reasons for deviations.

7. Iterative Process:

- This process may require multiple iterations as we refine our criteria, sample size, or analysis methods.

Note: What's considered **normal** and **abnormal** can depend on the context and business requirements. Using statistical techniques and sampling can help us identify divisions that exhibit unusual behaviors or characteristics, leading to a more data-driven decision-making process.