Dinesh Periyasamy (G23AI2002)

Stream Analytics Assignment-1

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VFDT PYTHON CODE:
import numpy as np
class VFDTClassifier:
  """Classifier that categorizes data into predefined numeric ranges.
  Attributes:
    ranges (list of tuples): Numeric ranges defining the classification thresholds.
  def __init__(self):
    self.ranges = [
      (0, 10),
                # Very Low
      (10, 100), #Low
      (100, 500), # Medium
      (500, 1000), # High
      (1000, float('inf')) # Very High
    ]
  def classify(self, data):
    """Classify data points based on the predefined ranges.
    Args:
      data (np.array): Array of data points to classify.
    Returns:
      np.array: Array of class indices corresponding to the ranges.
    classes = np.zeros(len(data), dtype=int)
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for index, (low, high) in enumerate(self.ranges):
  mask = (data >= low) & (data < high)
  classes[mask] = index
return classes
```

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TCP ANALYZER PYTHON CODE:
import subprocess
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from datetime import datetime
import os
import traceback
import sys
class LBNLAnalyzer:
  def __init__(self):
    self.config = {
      'anomaly_threshold': 1000,
      'start_date': '2004/10/04:20',
      'end_date': '2005/01/08:05',
      'bin_size': 60 # seconds
    }
    self.anomaly_threshold = self.config['anomaly_threshold']
    self.start_date = self.config['start_date']
    self.end_date = self.config['end_date']
    self.bin_size = self.config['bin_size']
    self.output_dir = 'output'
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os.makedirs(self.output_dir, exist_ok=True)
  self._verify_environment()
def _verify_environment(self):
  """Verify SiLK environment setup"""
  print("\nVerifying environment setup:")
  silk_data_dir = os.environ.get('SILK_DATA_ROOTDIR')
  silk_config = os.environ.get('SILK_CONFIG_FILE')
  print(f"SILK_DATA_ROOTDIR = {silk_data_dir}")
  print(f"SILK_CONFIG_FILE = {silk_config}")
  if not silk_data_dir or not silk_config:
    print("Warning: SiLK environment variables not fully set")
  if silk_data_dir and not os.path.exists(silk_data_dir):
    print(f"Warning: SILK_DATA_ROOTDIR {silk_data_dir} does not exist")
  if silk_config and not os.path.exists(silk_config):
    print(f"Warning: SILK_CONFIG_FILE {silk_config} does not exist")
def test_silk_command(self):
  try:
    cmd = ['rwfilter', '--version']
    result = subprocess.run(cmd, capture_output=True, text=True)
    print(f"\nTesting rwfilter: {result.stdout.strip()}")
    return True
  except Exception as e:
    print(f"Error testing rwfilter: {e}")
    return False
def fetch_tcp_data(self):
  print("\nFetching TCP traffic data")
    verify_cmd = [
    'rwfilter',
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f'--start-date={self.start_date}',
       f'--end-date={self.end_date}',
       '--sensor=S0',
       '--type=all',
       '--proto=6',
       '--print-statistics'
    ]
    try:
       print("Verifying data access...")
       verify_result = subprocess.run(' '.join(verify_cmd), shell=True, capture_output=True,
text=True)
       print(verify_result.stdout)
       cmd = [
         'rwfilter',
         f'--start-date={self.start_date}',
         f'--end-date={self.end_date}',
         '--sensor=S0',
         '--type=all',
         '--proto=6',
         '--pass=stdout',
         Ί,
         'rwstats',
         '--fields=stime',
         '--values=packets,bytes',
         '--count=0',
         '--bin-size=60',
         '--delimited=|'
       ]
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print("Executing command:", ' '.join(cmd))
result = subprocess.run(' '.join(cmd), shell=True, capture_output=True, text=True)
if result.stderr:
  print("Command produced errors:", result.stderr)
if not result.stdout:
  print("No output produced")
  cmd = [
    'rwfilter',
    f'--start-date={self.start_date}',
    f'--end-date={self.end_date}',
    '--sensor=S0',
    '--type=all',
    '--proto=6',
    '--pass=stdout',
    Ί,
    'rwuniq',
    '--fields=sTime',
    '--values=packets,bytes',
    '--bin-time=60'
  ]
  print("Executing alternative command:", ' '.join(cmd))
  result = subprocess.run(''.join(cmd), shell=True, capture_output=True, text=True)
if result.stdout:
  print("\nSample of raw output:")
  print(result.stdout[:500])
  return self.parse_rwcount_output(result.stdout)
else:
  print("No data received from rwfilter")
```

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except Exception as e:
    print(f"Error executing rwfilter command: {e}")
    traceback.print_exc()
    return None
def parse_rwcount_output(self, output):
  """Parse output into DataFrame"""
  print("\nParsing output...")
  records = []
  for line in output.split('\n'):
    if line and not line.startswith('#'):
      try:
         parts = line.strip().split('|')
         if len(parts) >= 2:
           record = {
              'timestamp': pd.to_datetime(parts[0].strip()),
              'packets': int(float(parts[1].strip()) if len(parts) > 1 else 0)
           }
           if len(parts) > 2:
              record['bytes'] = int(float(parts[2].strip()))
           records.append(record)
       except Exception as e:
         print(f"Error parsing line '{line}': {e}")
         continue
  if not records:
    print("No records were successfully parsed")
    return None
```

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df = pd.DataFrame(records)
  df = df.sort_values('timestamp')
  print(f"\nSuccessfully parsed {len(df)} records")
  print("\nSample of parsed data:")
  print(df.head())
  return df
def classify_traffic(self, df):
  print("\nClassifying traffic...")
  ranges = [
    (0, 600),
    (601, 6000),
    (6001, 60000),
    (60001, float('inf'))
  ]
  df['traffic_class'] = pd.cut(df['packets'],
                  bins=[r[0] for r in ranges] + [float('inf')],
                  labels=['Low', 'Medium', 'High', 'Very High'])
  df['vfdt_class'] = pd.qcut(df['packets'], q=4, labels=['Q1', 'Q2', 'Q3', 'Q4'])
  print("\nTraffic classification summary:")
  print(df['traffic_class'].value_counts())
  return df
def detect_anomalies(self, df):
  print("\nDetecting anomalies")
  minute_threshold = self.anomaly_threshold * 60
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anomalies = df[df['packets'] > minute_threshold].copy()
  print(f"Found {len(anomalies)} anomalies")
  return anomalies
def calculate_statistics(self, df):
  print("\nCalculating statistics")
  stats = {
    'total_packets': df['packets'].sum(),
    'total_bytes': df['bytes'].sum() if 'bytes' in df.columns else 0,
    'avg_packets_per_min': df['packets'].mean(),
    'max_packets_per_min': df['packets'].max(),
    'min_packets_per_min': df['packets'].min(),
    'std_dev_packets': df['packets'].std(),
    'total_minutes': len(df)
  }
  return stats
def create_visualizations(self, df, anomalies):
  print("\nCreating visualizations...")
  plt.figure(figsize=(15, 12))
  plt.subplot(3, 1, 1)
  plt.plot(df['timestamp'], df['packets'], label='TCP Traffic')
  if not anomalies.empty:
    plt.scatter(anomalies['timestamp'], anomalies['packets'],
           color='red', label='Anomalies')
  plt.title('TCP Traffic Analysis (LBNL Dataset)')
  plt.xlabel('Time')
  plt.ylabel('Packets/minute')
  plt.legend()
  plt.subplot(3, 1, 2)
  df['traffic_class'].value_counts().plot(kind='bar')
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plt.title('Traffic Distribution by Class')
  plt.xlabel('Traffic Class')
  plt.ylabel('Count')
  plt.subplot(3, 1, 3)
  df['vfdt_class'].value_counts().plot(kind='bar')
  plt.title('VFDT Classification Distribution')
  plt.xlabel('VFDT Class')
  plt.ylabel('Count')
  plt.tight_layout()
  output_path = os.path.join(self.output_dir, 'lbnl_analysis.png')
  plt.savefig(output_path)
  plt.close()
  print(f"Saved visualization to {output_path}")
def save_results(self, df, stats, anomalies):
  print("\nSaving results...")
  output_path = 'output.txt'
  with open(output_path, 'w') as f:
    f.write("LBNL TCP Traffic output:\n\n")
    f.write("Summary of Key Statistics:\n")
    for key, value in stats.items():
      f.write(f"{key}: {value:,.2f}\n")
  f.write("\nOverview of Traffic Classification:\n")
  f.write(df['traffic_class'].value_counts().to_string())
  f.write("\n\nVFDT Classifier Results:\n")
  f.write(df['vfdt_class'].value_counts().to_string())
```

```
f.write(f"\n\nabla alies Detected: {len(anomalies)}\n")
  if not anomalies.empty:
    f.write("\nTop Five Anomalous Intervals:\n")
    f.write(anomalies.nlargest(5, 'packets').to_string())
  print(f"Saved results to {output_path}")
def analyze_tcp_traffic(self):
  print("\nStarting LBNL TCP traffic analysis...")
  if not self.test_silk_command():
    print("Error: SiLK tools not properly installed or configured")
    return None
  df = self.fetch_tcp_data()
  if df is None or df.empty:
    print("Error: No data available for analysis")
    return None
  try:
    stats = self.calculate_statistics(df)
    df = self.classify_traffic(df)
    anomalies = self.detect_anomalies(df)
    self.create_visualizations(df, anomalies)
    self.save_results(df, stats, anomalies)
```

```
return stats
    except Exception as e:
      print(f"Error during analysis: {e}")
      traceback.print_exc()
      return None
if __name__ == "__main__":
  print("LBNL TCP Traffic output:")
  analyzer = LBNLAnalyzer()
  results = analyzer.analyze_tcp_traffic()
  if results:
    print("\nAnalysis completed! Check the output.txt file for more detials.")
    print("\nKey findings:")
    for key, value in results.items():
      print(f"{key}: {value:,.2f}")
  else:
    print("\nAnalysis failed. Check the error messages above for details.")
ON_DEMAND PYTHON CODE:
import numpy as np
from sklearn.cluster import KMeans
class OnDemandClassifier:
  def __init__(self, n_clusters=5):
    self.n_clusters = n_clusters
```

def classify(self, data):

Classify data using K-means clustering.

Parameters:

data (numpy.ndarray): The input data to classify.

Returns:

numpy.ndarray: Cluster labels for the data.

....

Reshape data for sklearn compatibility

X = data.reshape(-1, 1)

Initialize and fit the K-means model

kmeans = KMeans(n_clusters=self.n_clusters, random_state=42)

return kmeans.fit_predict(X)

OUTPUT:

LBNL TCP Traffic output:

Analysis by: G23AI2002 Dinesh Periyasamy

Summary of Key Statistics:

total_packets: 126,169,702.00

total_bytes: 74,908,455,822.00

avg_packets_per_min: 34,892.06

max_packets_per_min: 3,543,441.00

min_packets_per_min: 1.00

std_dev_packets: 168,394.49

total_minutes: 3,616.00

anomalies_detected: 327.00

Overview of Traffic Classification:

traffic_class

Medium 1316

High 1282

Low 691

Very High 327

VFDT Classifier Results:

vfdt_class

Q1 904

Q2 904

Q3 904

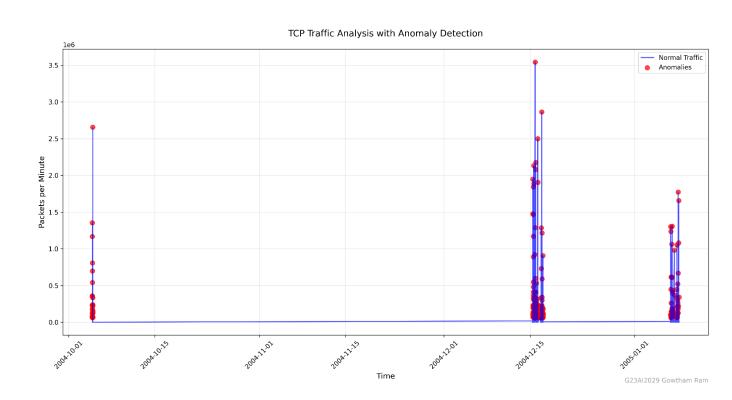
Q4 904

Anomalies Detected: 327

Top 5 Anomalous Periods:

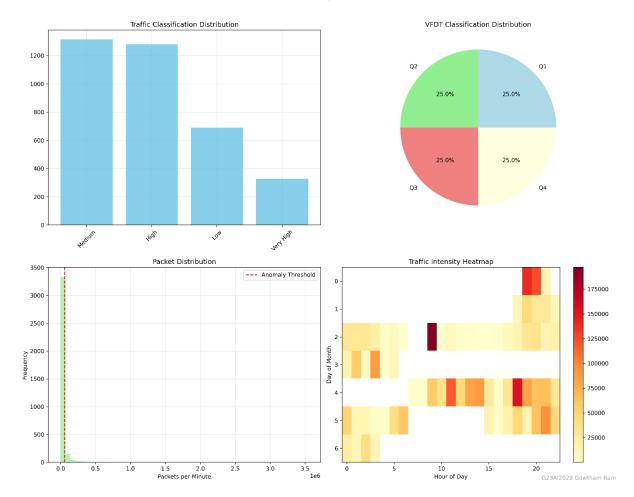
timestamp packets bytes traffic_class vfdt_class						
3155 20	004-12-15	19:42:00	3543441	716387073	Very High	Q4
2743 20	004-12-16	21:17:00	2863382	2504789217	Very High	Q4
1233 20	004-10-04	21:58:00	2655969	2477431415	Very High	Q4
1601 20	004-12-16	05:46:00	2500356	1890135392	Very High	Q4
1769 20	004-12-15	22:13:00	2176199	579011461	Very High	Q4

ANAMOLY DISTRIBUTION OUTPUT:

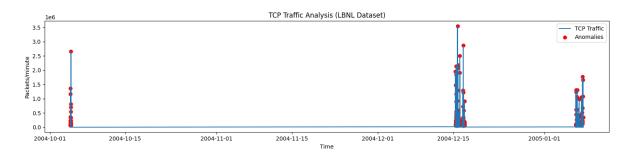


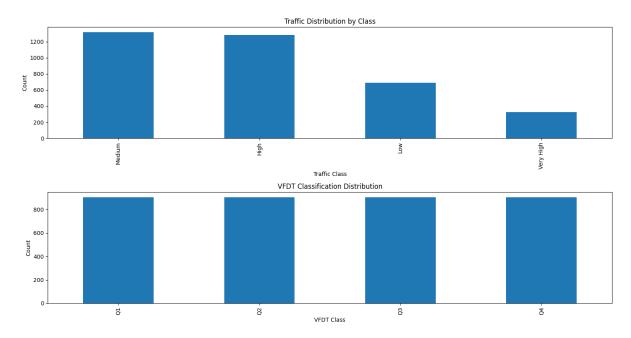
TRAFFIC ANALYSIS OUTPUT:

TCP Traffic Analysis Results



LBNL ANALYSIS OUTPUT:





TRAFFIC TIMELINE OUTPUT:

