COMP 4334 - Assignment 1

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Assignment Description

This assignment implements the **Grid-Based Close Pairs Problem** using **Spark RDDs** with optimized partitioning, error handling, and efficient distance calculation for spatial data. The goal is to identify pairs of geographic points within a specified distance threshold, leveraging Spark's distributed computation capabilities.

Algorithm Overview

- 1. Data Loading and Cleaning
 - File Paths:
 - geoPoints0.csv
 - geoPoints1.csv
 - Header Removal:
 - Skips the header line starting with "PointID" to avoid parsing errors.
 - Early Partitioning:
 - Repartitioning is applied early to balance the load across workers, reducing data skew.

```
geoPoints_files = [
    "dbfs:/FileStore/shared_uploads/michael.ghattas@du.edu/geoPoints0.csv",
    "dbfs:/FileStore/shared_uploads/michael.ghattas@du.edu/geoPoints1.csv"
]

points_rdd = sc.union([
    sc.textFile(file)
    .filter(lambda line: line.strip() and not line.lower().startswith("pointid"))
    .map(parse_point)
    .filter(lambda point: point is not None)
    .repartition(128) # Early partitioning to balance load
    for file in geoPoints_files
]).cache()
```

2. Point Parsing and Grid Cell Assignment

• Point Parsing:

- Each line is parsed into a **(pointID, x, y)** tuple, with robust error handling for empty or malformed lines.

• Grid Cell Calculation:

- Each point is assigned to a grid cell based on integer division.

```
def parse_point(line):
   try:
        line = line.strip()
        if not line or line.lower().startswith("pointid"):
            return None
       parts = line.split(",")
        if len(parts) != 3:
            return None
       point_id, x, y = parts
        try:
            x = float(x.strip())
            y = float(y.strip())
        except ValueError:
            return None
        point_id = point_id.strip()
        if not point id:
            return None
       return (point_id, x, y)
    except Exception as e:
       return None
```

3. Neighbor Pushing and Grid Mapping

• Efficient Neighbor Generation:

- Points are pushed to nearby cells within the distance threshold to capture boundary cases.

```
lambda p: [
        (cell, p) for cell in generate_relevant_neighbors(*get_grid_cell(p[1], p[2], 0.75))
    ]
).partitionBy(128).cache()
```

4. True Balanced Pair Extraction

- Efficient Pair Checking:
 - The find_close_pairs() function is optimized to avoid redundant distance calculations.
- Consistent Pair Ordering:
 - Pairs are consistently ordered to avoid duplicates.

```
def distance(p1, p2):
   x1, y1 = p1[1], p1[2]
   x2, y2 = p2[1], p2[2]
   return math.sqrt((x1 - x2) ** 2 + (y1 - y2) ** 2)
def find_close_pairs(points_list):
   n = len(points_list)
   pairs = set()
   for i in range(n):
       p1 = points_list[i]
        for j in range(i + 1, n):
           p2 = points_list[j]
            if len(p1) == 3 and len(p2) == 3 and distance(p1, p2) <= 0.75:
                pair = (p1[0], p2[0]) if p1[0] < p2[0] else (p2[0], p1[0])
                pairs.add(pair)
   return pairs
close_pairs_rdd = cells_rdd.partitionBy(128).groupByKey().flatMap(
   lambda cell_points: find_close_pairs(list(cell_points[1]))
).distinct().cache()
```

5. Collecting and Printing Results

```
close_pairs = close_pairs_rdd.collect()
print(f"Dist:0.75")
print(f"{len(close_pairs)} {close_pairs}")

Sample Output:

Dist:0.75
4 [('Pt06', 'Pt09'), ('Pt05', 'Pt07'), ('Pt03', 'Pt10'), ('Pt01', 'Pt15')]
```