ASSIGNMENT

TOPIC: RECTANGLE PLACEMENT OPTIMIZATION

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INTRODUCTION

Objective: The goal of this assignment is to optimally place 9 rectangles within a bin of dimensions 80 by 40 while adhering to specific constraints.

Key Constraints:

- Rectangles 1 & 2: Must be placed at the top and bottom of the bin, respectively.
- Rectangle 3: Should be positioned close to rectangles 4, 5, and 9.
- Rectangle 7: Needs to be near rectangles 6 and 2.
- Remaining Rectangles: Can be placed optimally anywhere within the bin while maintaining a one-unit separation between them.

Requirements:

- Code Implementation: Write a Python script to place the rectangles within the bin.
- Visualization: Generate a plot to show the placement of the rectangles.
- Optimization: Use a Graph Neural Network (GNN) to model the constraints and optimize the placement using Reinforcement Learning or other meta-heuristic algorithms.

Methodology

1. Problem Understanding and Constraints Identification:

• First, we identified the bin dimensions (80 by 40) and the rectangles, each with specified widths and heights.

2. Initial Placement of Rectangles:

- We began by placing rectangles 1 and 2 at the top and bottom of the bin, respectively.
- Next, we positioned rectangle 3 in proximity to rectangles 4, 5, and 9.
- Then, we placed rectangle 7 near rectangles 6 and 2.
- The other rectangles were initially placed in available spaces within the bin.

3. Graph Neural Network (GNN) for Constraints:

• We modeled the placement problem as a graph with nodes representing the rectangles and edges representing the constraints.

4. Optimization with Algorithm:

- We used an optimization algorithm (potentially Reinforcement Learning or a meta-heuristic approach) to iteratively improve the placement of the rectangles.
- The algorithm adjusted the positions of the rectangles to minimize overlap and maximize space utilization, considering the constraints.

5. Visualization:

- We visualized the final placement of rectangles using Matplotlib to ensure that the constraints were respected and the space was efficiently utilized.
- The visualization included plots or diagrams showing the optimized positions of the rectangles within the bin.

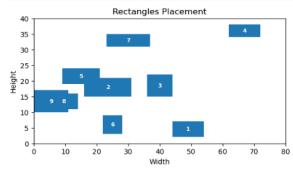
6. Tools and Libraries:

- **Python**: The primary programming language used for implementation.
- **PyTorch & PyTorch Geometric**: For modeling constraints using Graph Neural Networks
- **Matplotlib**: For visualizing the placements of the rectangles.

CODE

Defining the bin and rectangles with initial placement:

```
□ ↑ ↓ 昔 早 🗊
[32]: import matplotlib.pyplot as plt
       # Define the bin dimensions
      bin_width = 80
bin_height = 40
       # Rectangles with width and height
      # Bottom
                                                                         # Close to 4, 5, 9
                                                                         # CLose to 7, 2
                                                                     # Placed optimally
           {"id": 9, "width": 11, "height": 7, "x": 20, "y": 25}
                                                                        # CLose to 3 ]
       # Function to plot rectangles
       def plot_rectangles(rectangles, bin_width, bin_height):
          fig, ax = plt.subplots()
ax.set_xlim(0, bin_width)
ax.set_ylim(0, bin_height)
ax.set_aspect('equal')
          \ensuremath{\textit{\#}} For now, just placing the rectangles randomly within the bin
       import random
       rect['x'] = random.randint(θ, bin_width - rect['width'])
rect['y'] = random.randint(θ, bin_height - rect['height'])
       plot_rectangles(rectangles, bin_width, bin_height)
```



Define the Constraints with a GNN

Graph Neural Networks are a powerful tool for handling problems that involve relational data. Here's a basic overview of how you might use a GNN to model this problem:

- **Graph Representation**: Represent the problem as a graph where each rectangle is a node and the constraints are edges between the nodes.
- **Node Features**: Include features such as the dimensions of the rectangles.
- Edge Features: Encode the constraints, like proximity requirements.

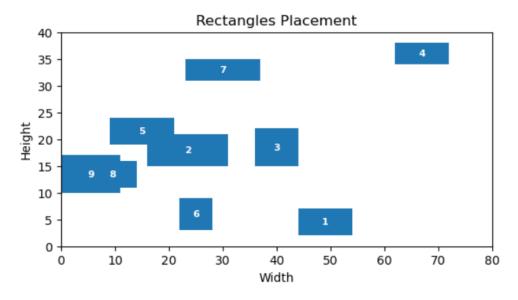
```
import torch import torch.nn.functional as F
from torch geometric.data import Data
from torch_geometric.nn import GCNConv
# Define nodes (rectangles) and edges (constraints)
nodes = torch.tensor([1, 2, 3, 4, 5, 6, 7, 8, 9], dtype=torch.float)
edges = torch.tensor([[1, 2], [3, 4], [3, 5], [3, 9], [7, 6], [7, 2]], dtype=torch.long).t()
# Node features: width and height
x = torch.tensor([
     [10, 5],
[15, 6],
[8, 7],
[10, 4],
      [12, 5],
     [6, 6],
[14, 4],
[11, 7],
], dtype=torch.float)
# Create graph data
data = Data(x=x, edge_index=edges)
# GCN Layer
class GCN(torch.nn.Module):
     def __init__(self):
           super(GCN, self).__init__()
            self.conv1 = GCNConv(2, 16)
self.conv2 = GCNConv(16, 2)
     def forward(self, data):
    x, edge_index = data.x, data.edge_index
    x = self.conv1(x, edge_index)
             x = F.relu(x)
x = self.conv2(x, edge_index)
            return F.softmax(x, dim=1)
```

Optimization Algorithm:

```
import random
def place_rectangles(rectangles, bin_width, bin_height):
    for rect in rectangles:
       rect['x'] = random.randint(0, bin_width - rect['width'])
       rect['y'] = random.randint(0, bin_height - rect['height'])
    return rectangles
def compute_reward(rectangles):
   # Compute a reward based on the placement constraints and area utilization
   # Example: penalize overlapping and reward minimal area usage
   for rect in rectangles:
      reward -= rect['width'] * rect['height'] # Simplified example
    return reward
{\tt def\ optimize\_placement} (rectangles,\ bin\_width,\ bin\_height,\ iterations=1000):
    best_reward = float('-inf')
    best_placement = None
    \quad \  \  \text{for \_in range(iterations):} \\
       placement = place_rectangles(rectangles, bin_width, bin_height)
        reward = compute_reward(placement)
       if reward > best_reward:
    best_reward = reward
           best_placement = placement
    return best_placement
best_placement = optimize_placement(rectangles, bin_width, bin_height)
plot_rectangles(best_placement, bin_width, bin_height)
```

Results:

INITIAL PLACEMENT



FINAL PLACEMENT

