

Report for Notebook Section_9.ipynb

Path: /mnt/data/Section_9.ipynb

Notebook Summary

- Number of cells: 6
 - Code cells: 6
 - Markdown cells: 0
-

Headings / Structure

- No markdown headings detected in the notebook.
-

Key Imports Found

- No imports detected (static scan)
-

Functions and Classes Detected

- Class: GraphSAGENet
Methods: __init__, forward
-

Outputs Found in Notebook

- stream: 2
 - stream_lines: 24
-

Notable TODO / FIXME Comments

- No TODO/FIXME/XXX comments found.
-

Example Code Snippets (First 5 Code Cells)

Cell 0

```
!pip install torch_geometric
```

```
import torch
```

```
from torch_geometric.data import Data
```

```
from torch_geometric.nn import SAGEConv
import torch.nn.functional as F

# 0,1,1,1]
```

Cell 1

```
#--- Define a small graph with 6 nodes ---
# Node features (2 features per node).
# Here benign users have [1, 0] and malicious have [0, 1] for illustration.
x = torch.tensor([
    [1.0, 0.0], # Node 0 (benign)
    [1.0, 0.0], # Node 1 (benign)
    [1.0, 0.0], # Node 2 (benign)
    [0.0, 1.0], # Node 3 (malicious)
    [0.0, 1.0], # Node 4 (malicious)
    [0.0, 1.0] # Node 5 (malicious)
]),
```

Cell 2

```
# Edge list (undirected). Connect benign users (0-1-2 fully connected)
# and malicious users (3-4-5 fully connected), plus one cross-edge 2-3.
edge_index = (
```

```
torch.tensor([
    [0, 1],
    [1, 0],
    [1, 2],
    [2, 1],
    [0, 2],
    [2, 0],
    [3, 4],
    [4, 3]
]),
```

Cell 3

```
# Labels: 0 = benign, 1 = malicious  
  
# y contains the true labels of the 6 nodes:  
  
# Nodes 0, 1, 2 are benign → label 0  
  
# Nodes 3, 4, 5 are malicious → label  
  
y = torch.tensor([0, 0, 0, 1, 1, 1], dtype=torch.long)
```

```
data = Data(x=x, edge_index=edge_index, y=y)
```

Cell 4

```
# Instantiate model: input dim=2, hidden=4, output dim=2 (benign vs malicious)  
  
model = GraphSAGENet(in_channels=2, hidden_channels=4, out_channels=2)
```

```
# Simple training loop
```

```
optimizer = torch.optim.Adam(model.parameters(), lr=0.01)  
  
model.train()  
  
for epoch in range(50):  
    optimizer.zero_grad()  
  
    out = model(data.x, data.edge_index)  
  
    loss = F.nll_loss(out, data.y)  
  
    loss.backward()  
  
    optimizer.step()
```

Static Analysis — Potential Issues & Recommendations

This is an automated, best-effort static scan. The code was **not executed**.

Recommendations

- **Dependency management:** Add a requirements.txt or environment.yml including all imports.
- **Cell outputs:** Save important results (figures, tables) to external files when needed.
- **Functionization:** Convert repeated logic into reusable functions with docstrings.
- **Error handling:** Use try/except blocks for file I/O or network operations.
- **Type hints & docstrings:** Improve readability and IDE support.
- **Testing:** Add unit tests (e.g., pytest) for important components.

How to Run This Notebook Reproducibly

1. Create a virtual environment (conda/venv).
 2. Install dependencies from requirements.txt.
 3. Open the notebook in JupyterLab/Notebook.
 4. Run cells in order.
 5. For heavy tasks, consider porting logic to Python scripts.
-

Suggestions for Further Improvements

- Split logic into smaller modules/scripts.
 - Add a README explaining purpose, workflow, and expected outputs.
 - Document datasets and add sample data or downloading utilities.
 - Add explanations and captions to visualizations.
-

Summary

- **6** code cells, **0** markdown cells.
 - **0** imports detected (static scan).
 - **1** class (GraphSAGENet).
 - Recommendations provided for clarity, reproducibility, and maintainability.
-