

# Project Implementation Plan

## Comparative Analysis of Graph Clustering Features

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### 1. Role Division

Chiara <i>Data &amp; Integration</i>	Luca <i>Graph Engineering</i>	Leonardo <i>ML &amp; Experiments</i>
Data ingestion Cleaning Hybrid Vectors	Structural Metrics PageRank, BC, CC Approximations	Node2Vec Dimensionality Reduction Clustering

### 2. Recommended Tech Stack

- **Graph Library:** igraph (C++) or networkx
- **Embeddings:** node2vec or gensim
- **ML:** scikit-learn
- **Data:** pandas, numpy

### 3. Implementation Roadmap

#### Phase 1: Setup & Data Ingestion (Chiara)

- Repo Setup: GitHub repo with `.gitignore`
- Parser: Extract ASIN, Group, and Edges from `amazon-meta.txt`
- Filter: Keep only Book, DVD, Video, Music groups
- Graph: Build Undirected Graph object

#### Phase 2: Feature Engineering (Parallel)

##### Structural (Luca)

- PageRank: Compute and normalize
- Clustering Coeff: Compute and normalize
- Approx. Betweenness: Sampling-based BC ( $k = 1000$ )
- Approx. Closeness: Sampling-based CC

##### Topological (Leonardo)

- Node2Vec: Setup random walks
- Training: Train for  $d = 128$  dimensions
- Storage: Save to `.npy`

#### Phase 3: Hybridization (Chiara)

- Merge: Master DataFrame indexed by Node ID
- Hybrid Vector: Concatenate Structural (6 dims) + Embedding (128 dims)
- Reduce: Apply UMAP/PCA

#### **Phase 4: Experiments (Leonardo)**

- Cluster: K-Means ( $k = 4$ ) on all 3 datasets
- Validate: Compute ARI and NMI scores
- Visualize: t-SNE scatter plots
- Profile: Measure execution time