

Interactive Visualization of the Stock Market Graph

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Stock Market Data

- Huge amounts of accessible data on a daily basis
- Consists of a variety of fields such as price, volume, change
- Stock price interactions form a complex system
- Want to understand these interactions of the subsystems



Constructing the Market Graph

1. From a dataset, compute the correlation matrix



- 2. Convert correlation matrix to a graph, where
 - Vertices represent stocks
 - Edges represents a relationships between two stocks

correlation(stock1,stock2) > THRESHOLD



What Are We Visualizing?

- Find clusters/groups of stocks that exhibit certain trading patterns
- Maximum Cliques
 - Highly positively/negatively correlated subsets of stocks
- Independent Sets
 - Completely diversified stocks
- Quasi-Cliques/Independent Sets
 □ Generalizations → allow for near matches
- Clusters ? Cliques/IS interchangeably



Existing Approaches to Visualizing Graph Structures

- Determine target structures (i.e. clusters) a priori and use a standard layout algorithm to show the results
- 2. Use a layout algorithm optimized to visually differentiate target structures
- 3. Our approach: combine the two
 - Find target structures first, but include additional nodes and edges for context
 - Then use force-directed layout algorithm to effectively visualize the results





Motivation: Usage Scenarios

- Portfolio management (static)
- Real-time market analysis (dynamic)

 Exploratory analysis of trading data to gain new insights, spot patterns/trends, etc (static)

Motivation: Visualizing Results from a Real-time Data-mining Pipeline



Viz Client

Viz Client



Visualization Goals

- 1. Visualize different graph structures representing various *patterns* and *trends*
 - (quasi-)cliques and (quasi-)independent sets
 - positively and negatively correlations
- 2. Represent inter-cluster relationships
- 3. Dynamic graph capabilities
- 4. Interaction for efficient data exploration
- 5. Information integration



Force-Directed Graph Layout Model

- Create "summaries" of the graph using the clusters and their induced subgraphs
- Force model: spring-embedded layout
- Spring lengths and tensions parameterized to optimize layout

Highly related clusters should be close

Independent clusters and minimally related clusters should be further apart



F.D. Model Parameterizations

Edge Length

□ Cluster-Cluster edges (CC)

- # intra-cluster edges (shows "connectedness" of clusters)
- □ Cluster-Member edges (CM)
 - Quasi-cliques \rightarrow # intra-cluster edges ("clique contribution")
 - Cliques: cluster sizes (more space to larger clusters)

Tension

- □ CM edges use constant "tight" tension
- CC edge tension proportional to # of inter-cluster links



Differentiating cluster types

- Correlation Metrics: positive negative, independent
 Color encoded
- Cluster types: (quasi-)
 Cliques and (quasi-)
 Independent sets
 - Transparency-encoding for cluster summary
 - Individual members edge length encodes "clique contribution"





Interaction & Information Integration

Interaction Features

- □ Geometric pan/zoom
- Display/hide cluster outlines
- Symbol search for quick navigation
- Overview display for global context
- Node context menus provide stock quotes and news:
 - Stock news from various sources integrated via RSS feeds
 - Online quote details and Google search for provided by opening an external web browser

Dynamic graph capabilities

- Receive remote graph updates via socket connection to a "graph update server"
- Nodes/edges can be added, removed or replaced
- Event-based architecture allows for automatic processing of new updates
- Force-model allows for efficient incremental layouts when new nodes/edges placed "intelligently"



Future Work & Improvements

- Handle overlapping clusters
- Encode other variables
 i.e. node size could encode trade volume
- Ability to view underlying edge weights
- Ability to optionally view complete underlying graph
 - especially the intra-cluster edges



Future Work & Improvements (2)

- Interactively adding/removing nodes and edges
- Semantic zoom
- Focus+Context
- Other clustering methods besides partitioning via (quasi-)cliques and independent sets



Conclusion

Implemented basic Visualization tool for exploring the market graph

- Visualizes different cluster types and their attributes
- User interaction for pan/zoom, on-demand details (quotes, news, web search)
- Dynamic graph capability to support a real-time data processing pipeline



References

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- 3. Frank van Ham and Jarke J. van Wijk. **Interactive visualization of small world graphs**. In *Proceedings of the IEEE Symposium on Information Visualization*, pages 199–206, Washington, DC, USA, 2004. IEEE Computer Society.
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DEMO





THE END!





Construct a Similarity Matrix

Currently, our similarity measure is

$$C_{ij} = \frac{\langle R_i R_j \rangle - \langle R_i \rangle \langle R_j \rangle}{\sqrt{\langle R_i^2 - \langle R_i \rangle^2 \rangle \langle R_j^2 - \langle R_j \rangle^2 \rangle}},$$

where:

$$R_i(t) = \ln P_i(t) / P_i(t-1)$$