Phylogenetic Comparisons of Malware

Virus Bulletin Conference 2007

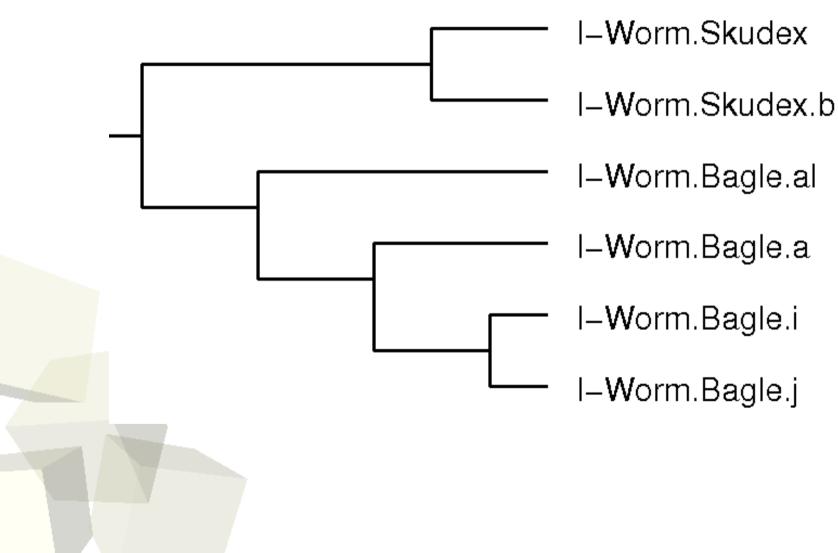
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Understanding Malware Evolution

- Long-lasting malware evolves
 - New exploits, new payloads, detection avoiding, bug fixes, etc.
 - Code is copied between families
 - → Example: Bagle and Agobot
 - → Both released source code: code was used elsewhere
- Q: how to understand / track evolution?
 - How to <u>find</u> relationships between samples?
 - How to <u>explore</u> found relationships?
- One approach: malware phylogenies
 - phylogeny: graph of "species" derivation relationships
 - akin to "tree of life" for biology





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Evaluating Phylogenies

Are phylogeny systems useful in practice?

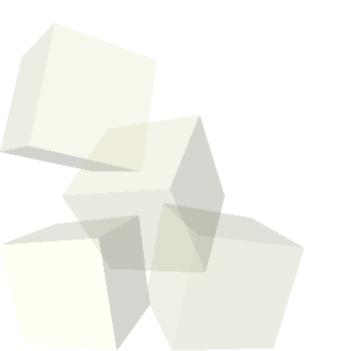
- little published on actually using phylogenies
 some pretty pictures and proof-of-concept
- wanted a kind of case study to find out more
 - → clarification of problems and benefits in practice
 - > be able to report experiences, evaluate phylogeny extraction methods

Target: Agobot malware families

- Agobot source was released widely
 - → was used as basis for many different bots
 - → was available to us, enabling systematic evaluation
- Can expected complicated evolution history
 > easy phylogenies won't expose weaknesses

Outline of Talk

- Recap / introduce malware phylogeny methods
- Agobot study
- Summary of problems and attempted solutions



MALWARE PHYLOGENY TECHNIQUES



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Origins/Parallels in Biology

- Need to <u>reconstruct</u> organism evolution history
 - <u>guess</u> relationships by examining samples
- Similarity method one of two main ways
 - Species A more similar to B than C implies A and B (probably) share a closer ancestor.
 - What is needed to computer-generate models:
 - → 1. Similarity scoring function
 - 2. Graph construction algorithm based on similarity
 - common: hierarchical clustering



Inferring Malware Evolution

Typical in malware phylogenies:

- Similarity-based methods almost exclusively
- Hierarchical clustering is typical
 - > produces strictly binary trees
 - malware evolution known to be non-tree like
 - code sharing, for example --- a gene transfer analogue
- The similarity function often the main difference between techniques
 - different program-to-program comparisons
 - → they choose different aspects of similarity

Similarity Approaches Survey

- Control graph matching ([CE04] [DR05])
 - program similarity = flow similarity
 - (see Liang et al. in this years conference)
- Normalized Compression Distance [W05]
 - program similarity = shared information
 - idea: if to programs are similar their concatenation compresses well
- Feature vector / n-gram based [WKLP05]
 - n-gram: sequence of n characters (bytes, operations,...)
 - program similarity = feature vector similarity

APPLICATION STUDY: AGOBOT RELATED FAMILIES



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Study Design

Data sources:

- ~4000 bot related samples
 - → scanned by BitDefender: selected all "bot" related
 - unpacked & dumped using Norman Sandbox
 - → 1194 distinct samples when unpacked
- 15 bot variants constructed in vitro
 - → used Agobot 3 source code
 - 15 different features turned on/off using #ifdefs
 - 2^15 different combinations possible
 - useful for producing controlled example evolution histories



Exploratory Study

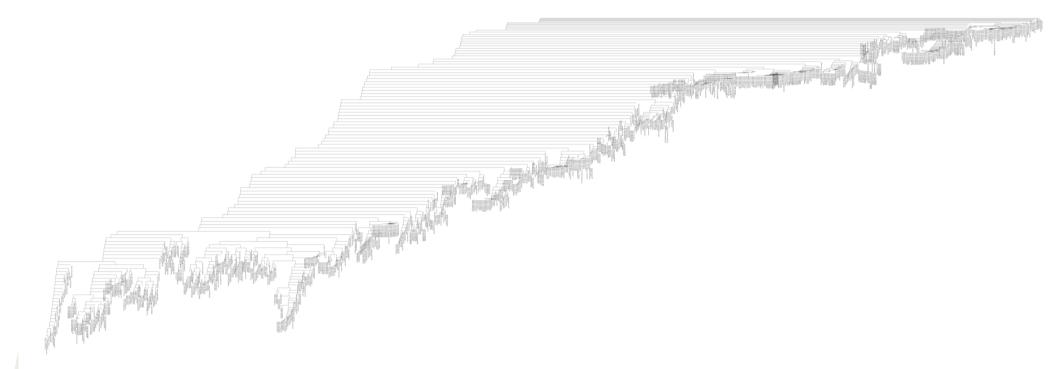
The Plan:

- construct phylogenies using NCD and N-gram based
- understand main evolution features:
 - (1) related families
 - (2) key branch points

The Reality:

- NCD took several days to complete on ~1200 samples
 - (Our N-gram implementation took ~40 mins, including disassembly)
 - → started with N-grams, used NCD for subsets
- Wrestled with results, plenty of ad hoc exploration

It's SO obvious

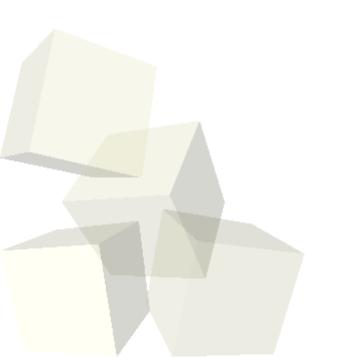


Phylogram (tree) of all 1209 samples



Problem: Tree size

- Tree size was a significant problem
 - was not easily solved by simple zooming and panning





Dealing with Tree size

Tried three approaches to dealing with size:

- Draw trees as unrooted graphs using different layout techniques, instead of "phylograms" (binary trees)
 can help distinguish major groupings visually
- 2. Merge sub-trees with high similarity & common name
 - → 20 closely related SdBots in sub-tree conveys little information about overall evolution
 - family history instead of speciation events
- 3. Split trees to reduce individual tree size
 - → can be explored independently or compared

Splitting Trees

- Idea is to split trees at nodes where sub-trees have "low" similarity
 - for suitable definitions of "low"
 - because "low similarity" \rightarrow "not useful"
 - → if similarity measure working fine:
 - then samples between sub-trees are unrelated
 - → if measure is just not picking up the similarity:
 - trees will be misleading in some way
 - look for other means and indicators (e.g. parallel trees)



Data from Merging & Splitting

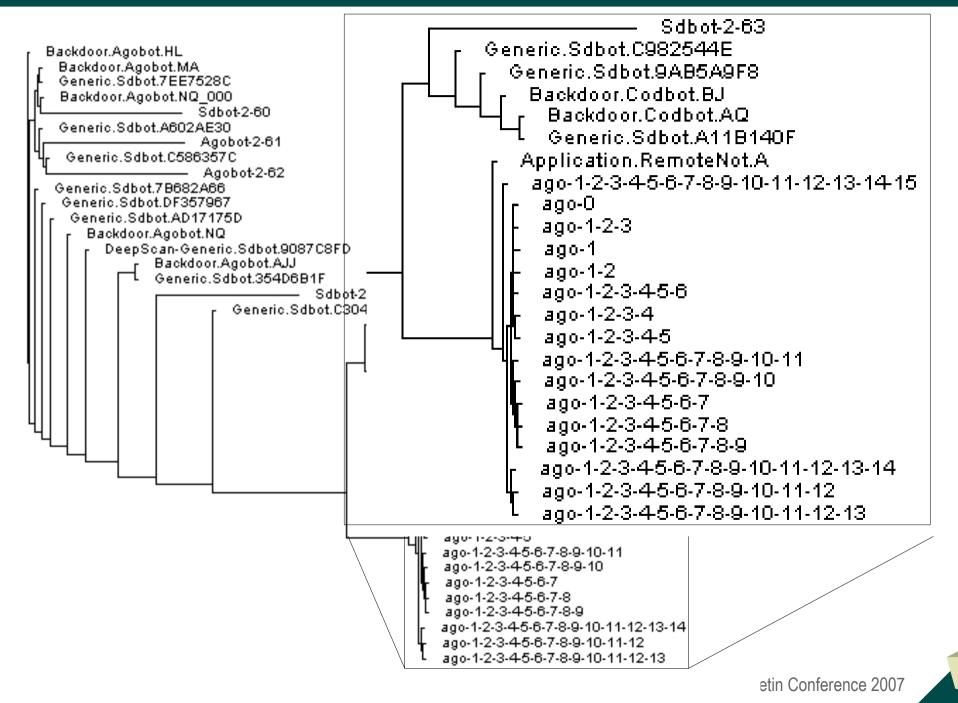
- Merging on common family names
 - 119 samples merged, < 10%
- Splitting on < .4 similarity</p>
 - 356 splits, 308 into single leaf node trees
 - 8 trees with >10 non-leaf nodes, largest was 137 nodes

Family Characteristics

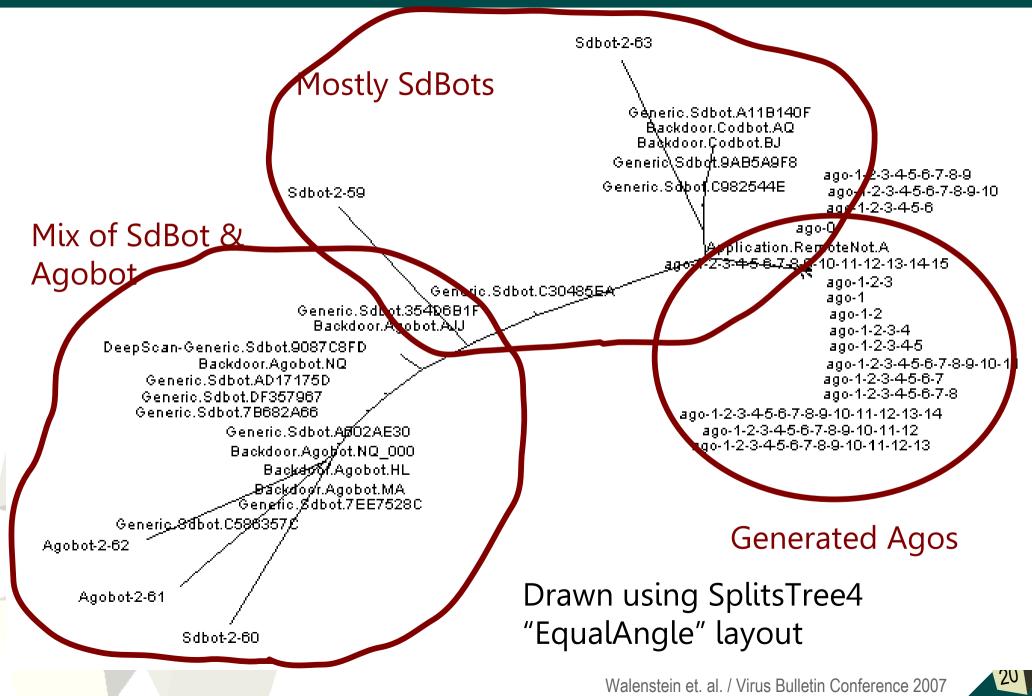
- Most trees had "mixing" of named species
 - Sdbots mixed with Rbots, IRCbots, etc.
 - No clear separation into major lines by any technique we had available
 - data available suggested:
 - highly interleaved development and sharing
 - → bad naming, or
 - → poor phylogenies
 - Order of 10 main branch points with multiple related variants



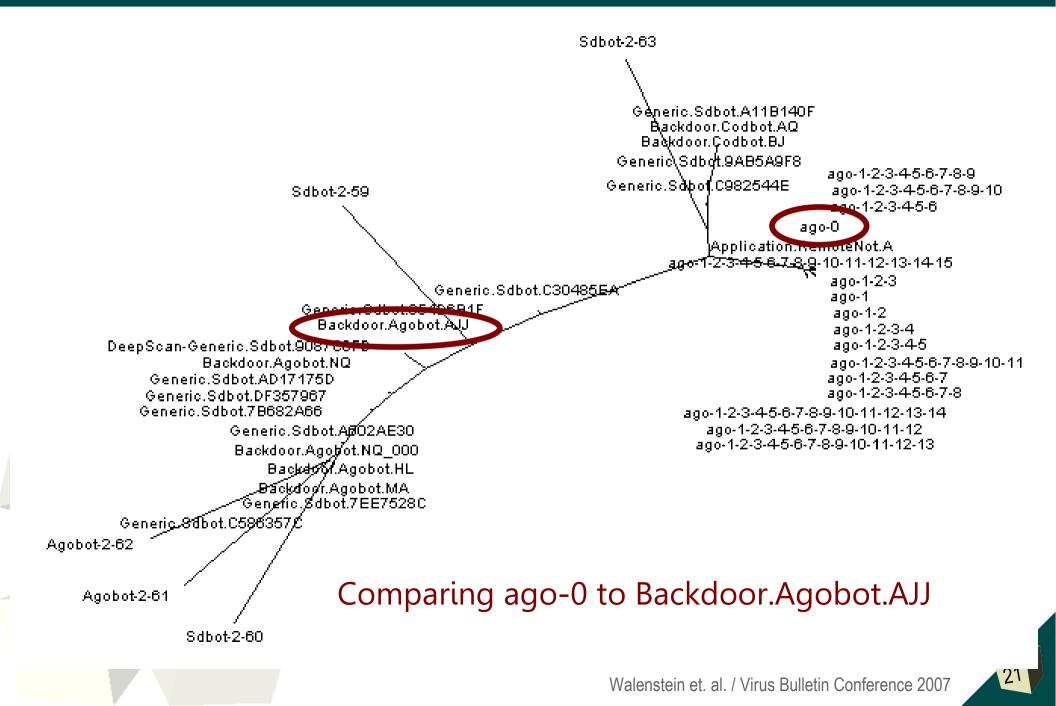
Example Tree (40 Nodes)



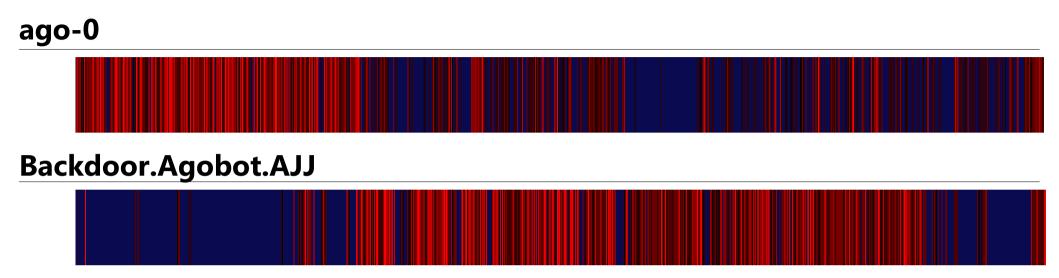
Unrooted Tree Layout



Examining Leaves

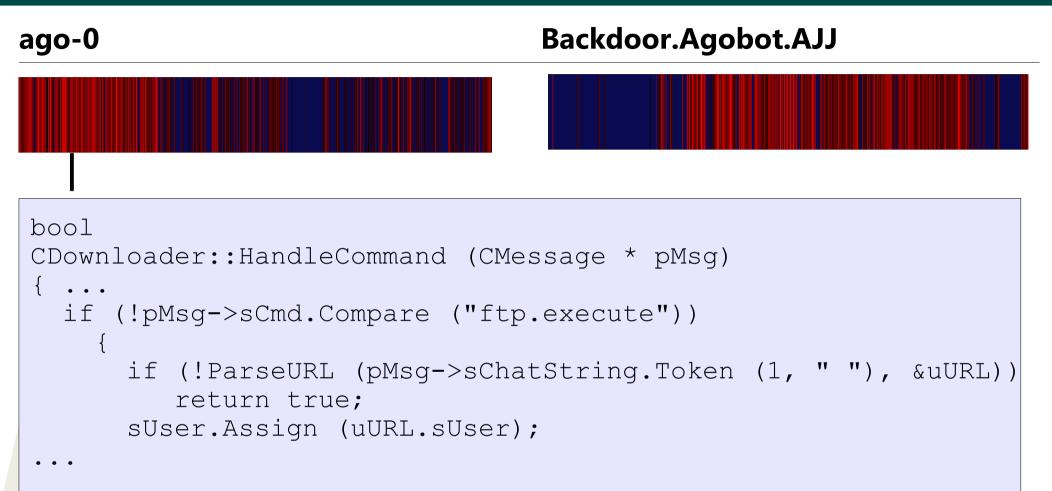


Examining Matches



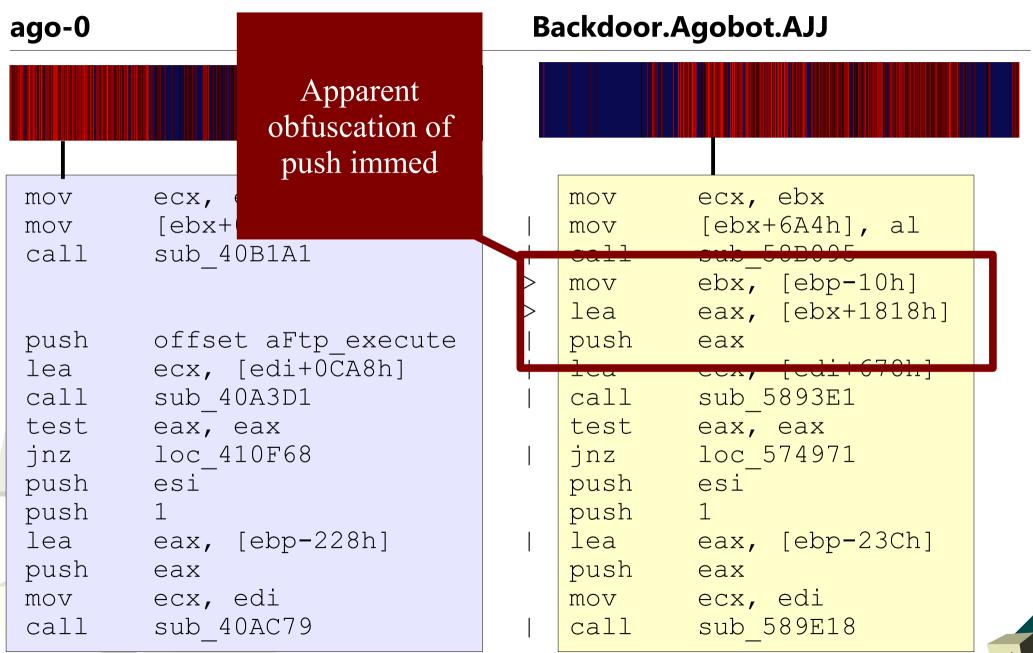
- Visualization of matches between two samples
- Legend:
 - red = match, brighter = more matching n-grams
 - blue = no match

Examining Matches

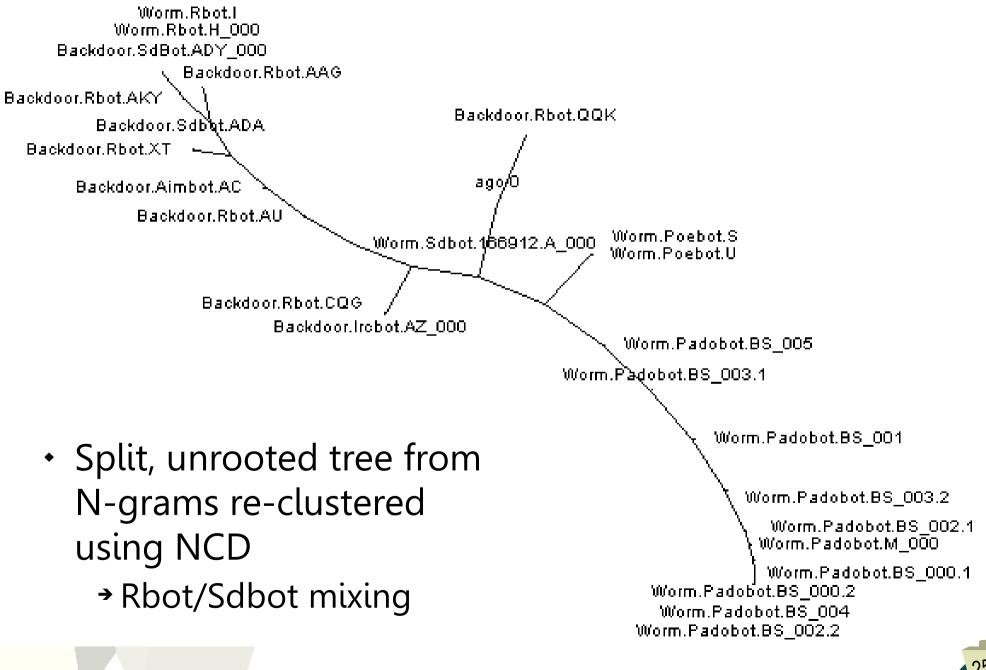


Know this from tracing source to executable

Disassembly Matching



Checks Using NCD



Discussion

- Limitations in exploration of Agobot
 - Sample limitations:
 - collection completeness, unpacking & naming correctness
 - Phylogeny modeling limitations
 Ilimited selection of similarity function, clustering
- Some issues are clearer, regardless
 - Tree size and clustering issues will remain even if the above limitations are met
 - Question raised as to what kinds of insight will be extracted from available data and techniques
 - Tree structures may be poor choice for malware phylogenies

Conclusions / Open Questions

Exposure of knowledge gaps / open question

- How to provide useful analysis support?
 - → our experiences suggest a need to support:
 - splitting, merging, and alternate layouts
 - visualization, comparison, exploration
- Need to explore network-based modeling
 - → current tree extraction may frequently be inappropriate
- How to understand effect of data set / problem
 - → denser / better data set may help
 - wish to investigate Storm

Open Questions & Future Work

- Question of how to sytematically evaluate?
- Have been investigating controlled methods
 - Using artificial evolution trees (from Agobot and others)
 - → A priori known "correct" derivation trees
 - by construction, using automated program mutation
 - Apply phylogeny distance measures to quantitatively compare trees

Links to Online Resources

- CLUTO (glaros.dtc.umn.edu/gkhome/views/cluto)
 - → feature-based and similarity-based clustering
 - → output of graphs, matrices
- SRL's NCD package
 - → NCD between pair files
 - → generates similarity matrix in CLUTO format
 - → www.cacs.louisiana.edu/labs/SRL/projects/NCD
- SplitsTree (www.splitstree.org)
 - calculates tree splits
 - multiple tree layouts

References

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