



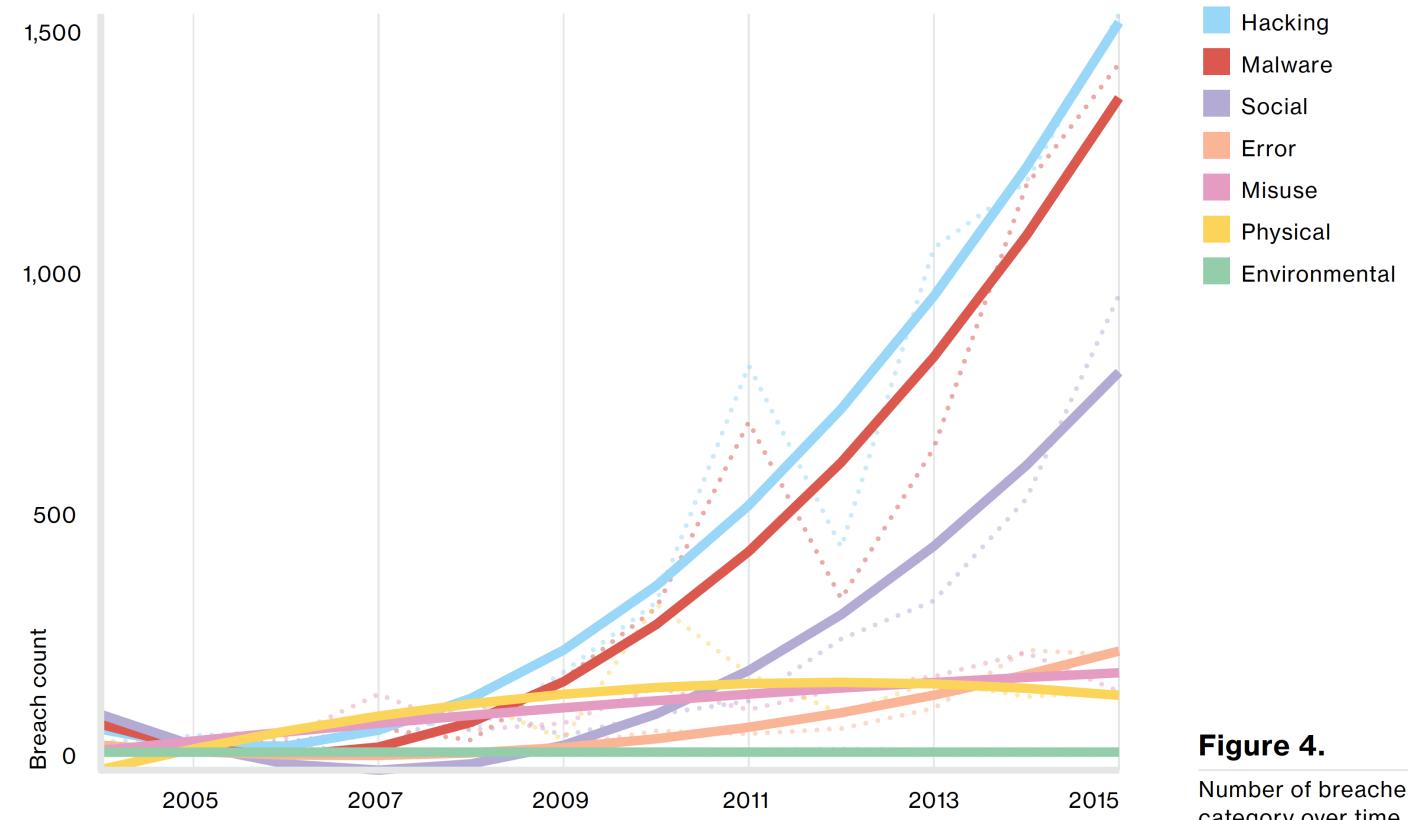
An AI Approach to Malware Similarity Analysis: Mapping the Malware Genome With a Deep Neural Network

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Lead Research Scientist
Invincea Labs

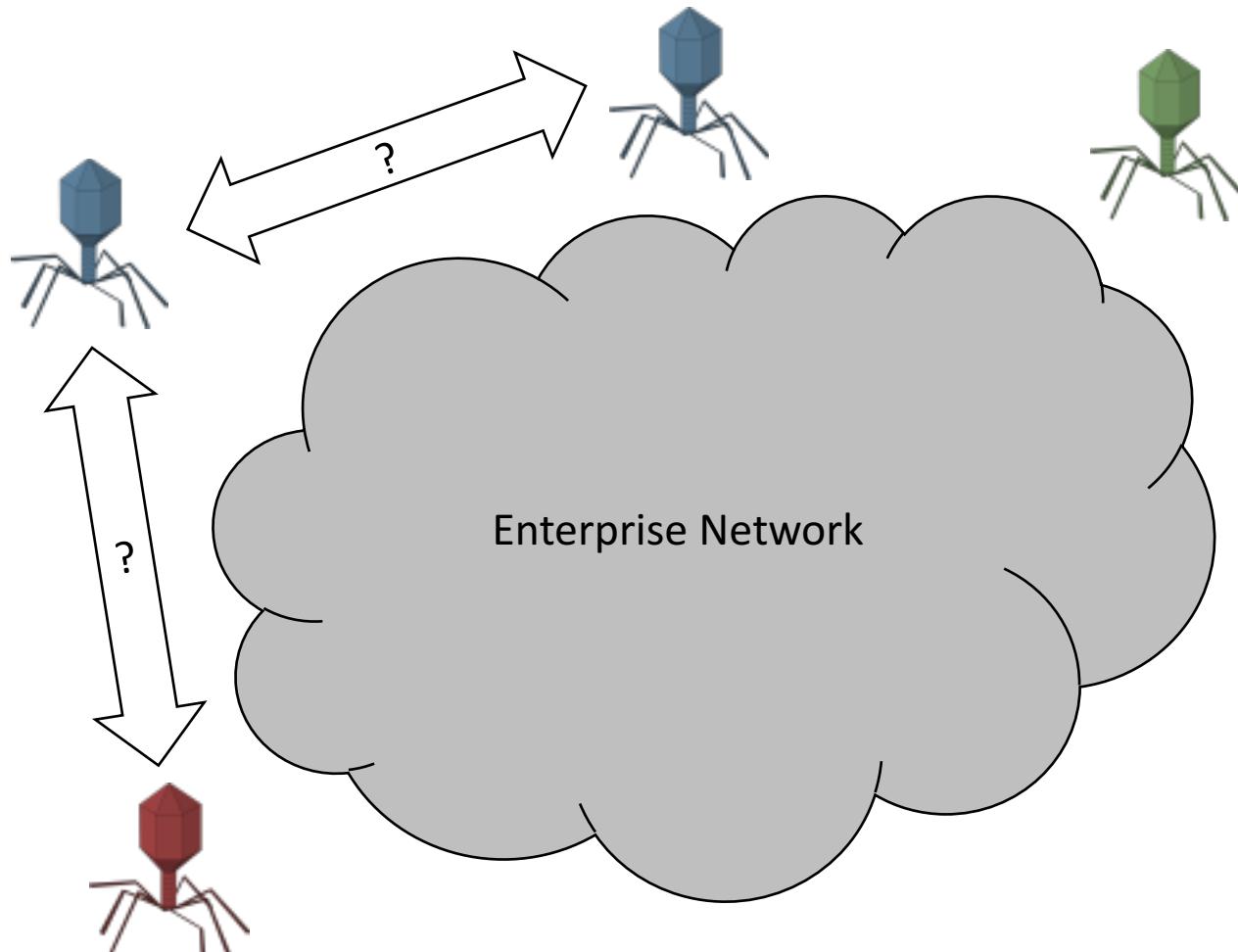
Why AI?

- Intelligence is critical for prevention and remediation
- AI is good at finding patterns in large data

Number of Network Breaches Per Year
(Verizon's 2016 Data Breach Investigations Report)



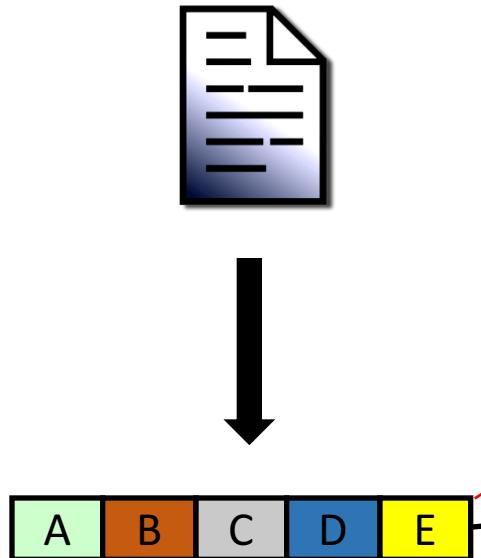
Intelligence through Similarity



- Benefits
 - Identify threat actors
 - Link various attacks to a single actor
 - Quickly understand functionality
 - Speed up reverse engineering
- Mitigation
 - Signatures
 - Network Rules

Finding Similar Malware

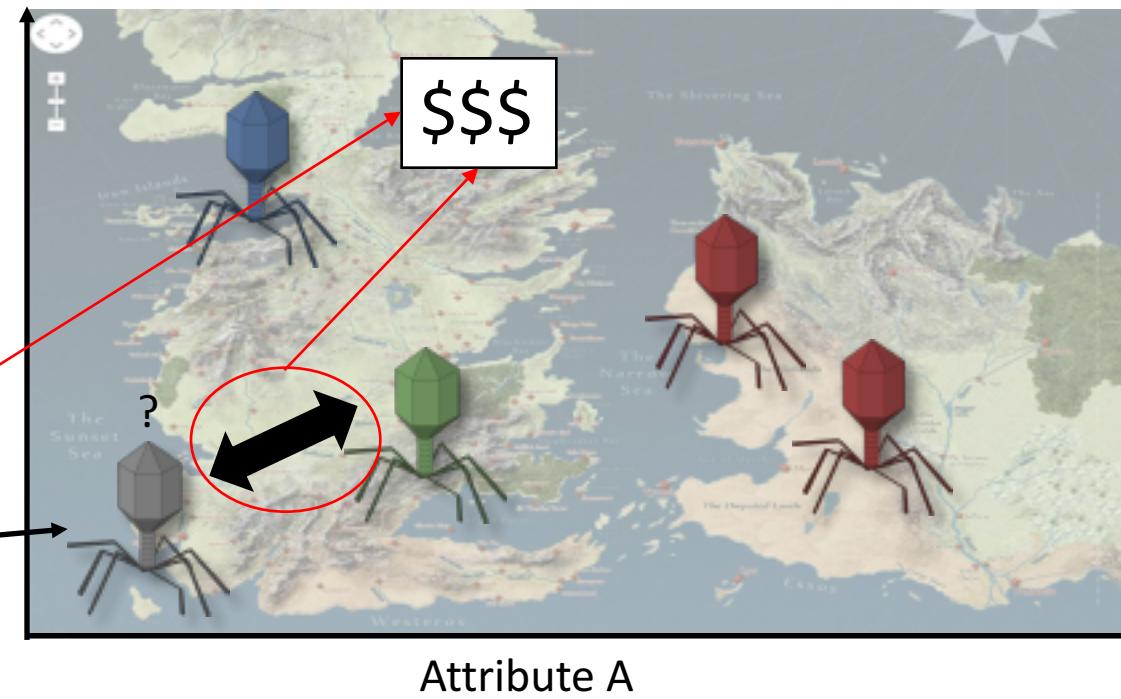
Attribute Extraction



Attributes

- Byte n-grams
- Opcode n-grams
- Printable strings
- System calls
- ...

Attribute Map (Embedding)

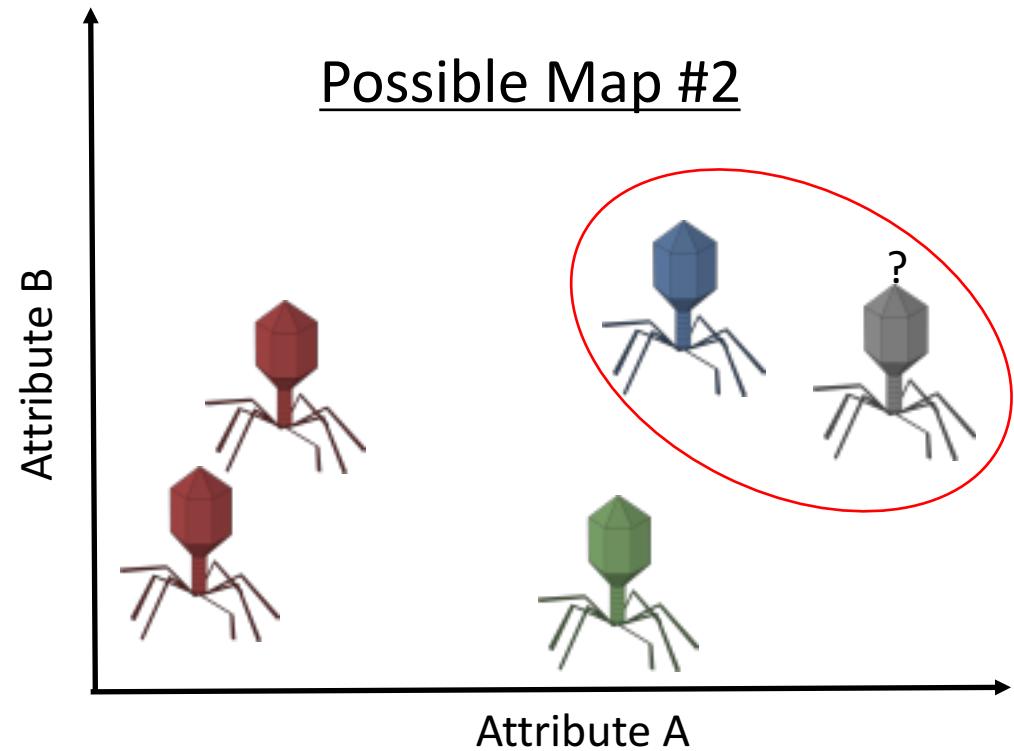
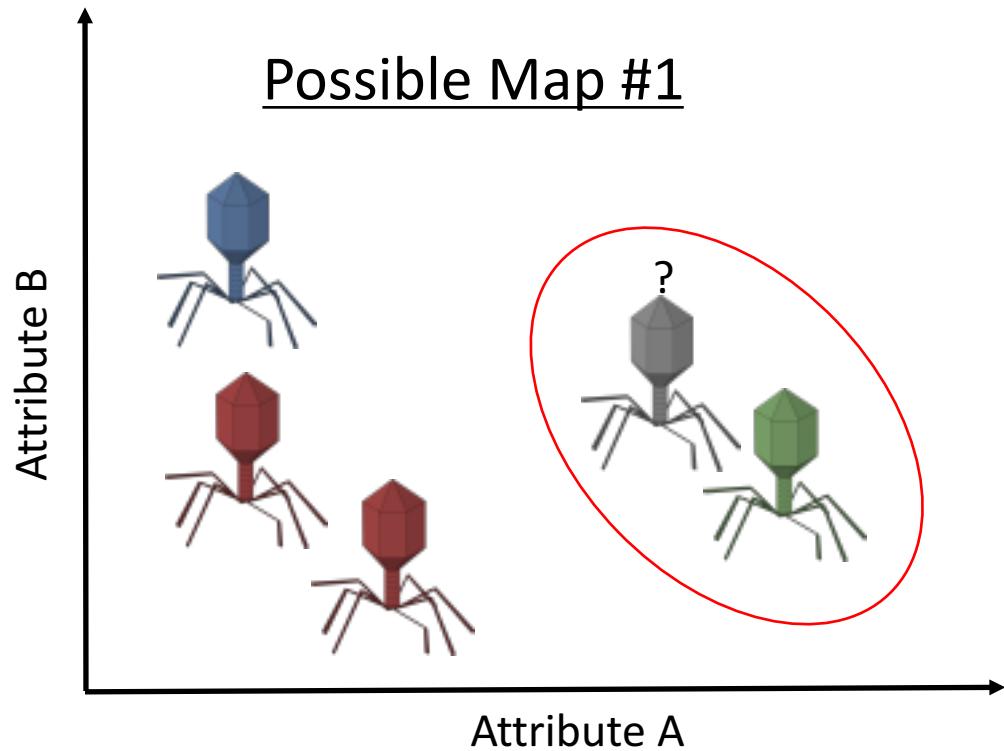


Similarity Search

- MinHash
- Feature hashing
- Other sketching
- ...

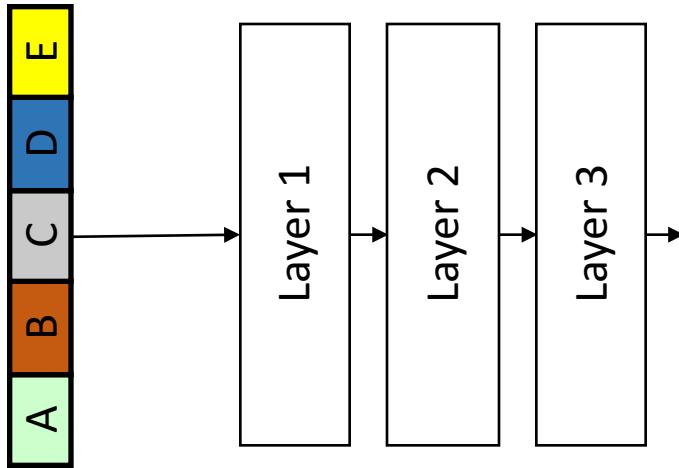
Jang, Jiyong et. al. *Proceedings of the 18th ACM conference on Computer and communications security*. ACM, 2011.
Sæbjørnsen, Andreas, et al. *Proceedings of 18th international symposium on Software testing and analysis*. ACM, 2009.
Bayer, Ulrich, et al. *NDSS*. Vol. 9. 2009.
...Many more

Issues with Attribute Maps



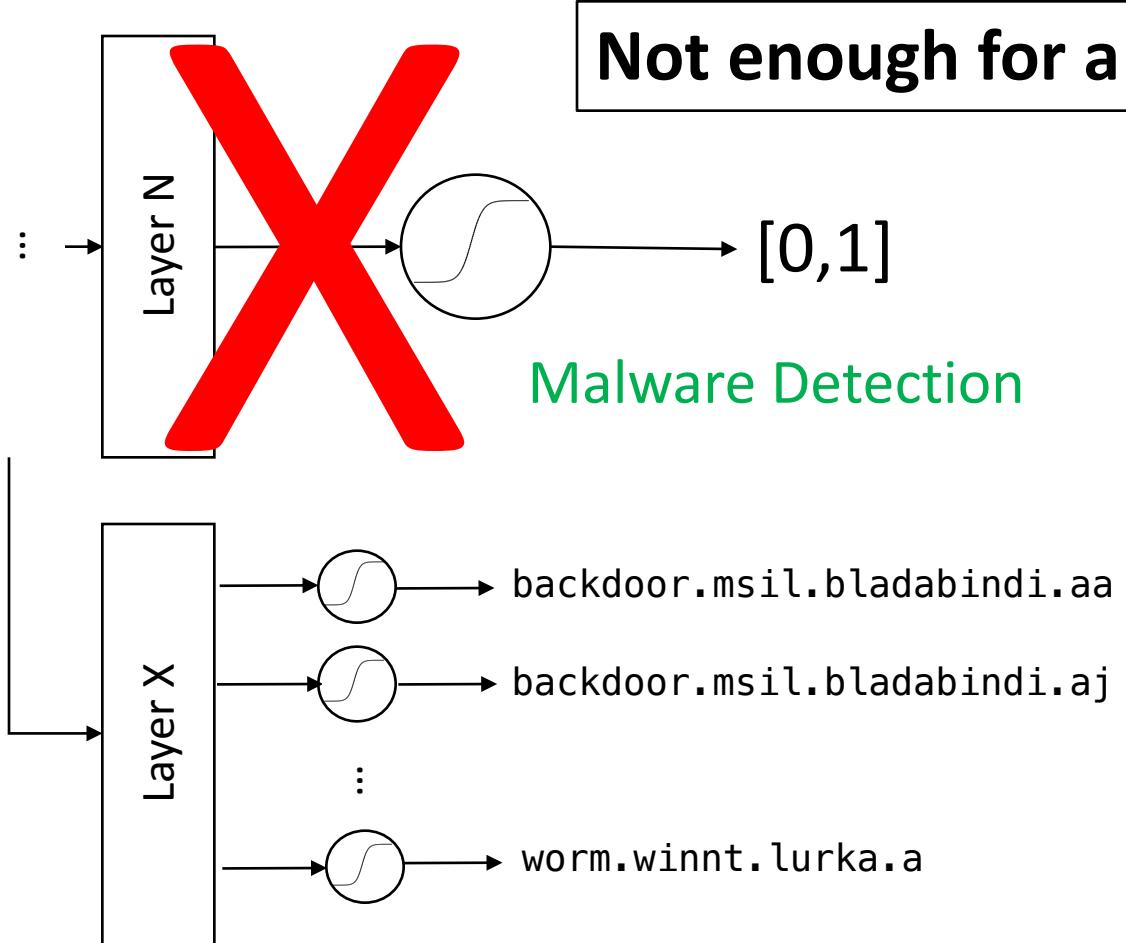
How to get consistent results, regardless of attributes?

Supervised Classification (Endpoint Solution)



Joshua Saxe and Konstantin Berlin, (*MALWARE*). IEEE, 2015.

Categorical Classification



Not enough for a triage system!

Malware Detection



0.97 F1-score (precision and recall)

- 1500 Microsoft Families
- 2.0M Training files

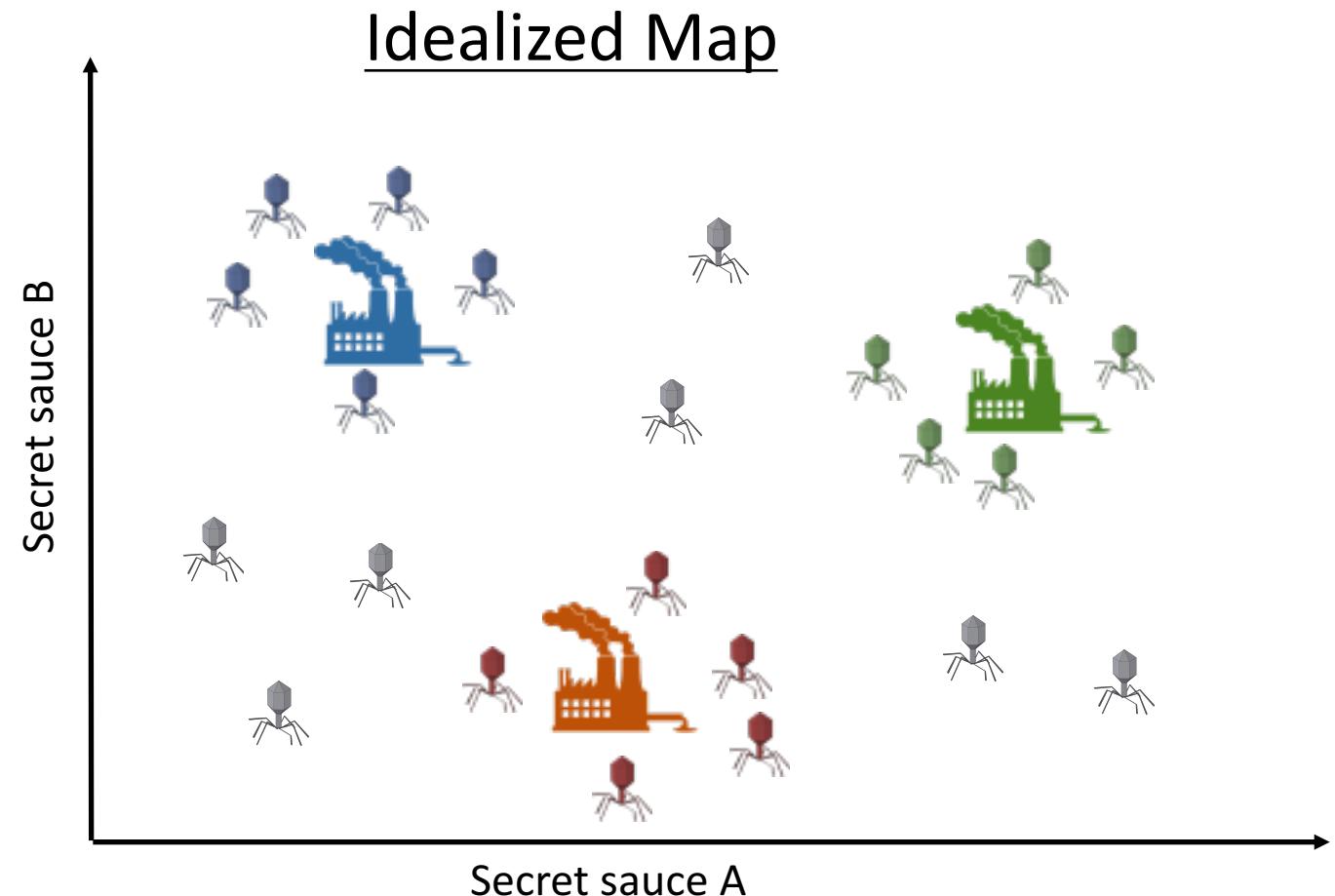


Given a set of attributes how do we create a “good” map?*

*No luck required

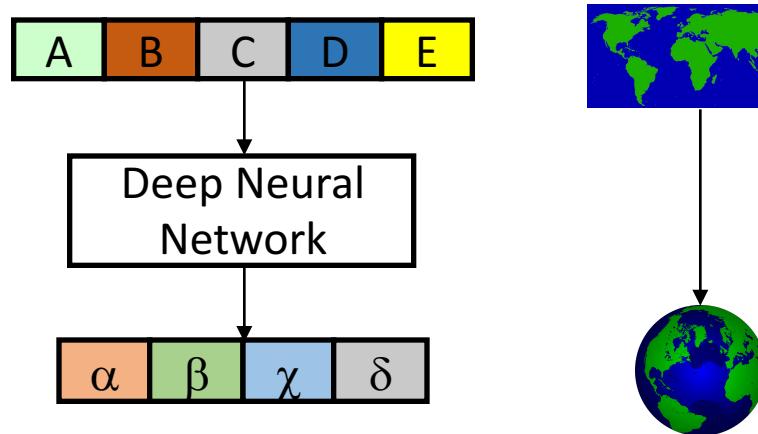
Imaginary World of Malware Factories

- Ideal World
 - Each hidden factory produces one malware family/variant
 - Factories are positioned relative to what and how they exploit vulnerabilities
- **...but this imaginary, no!?**



There is No Spoon Map...

- We created the map when we selected the attributes
- We can morph them in any way we choose
- One good way to morph the attributes is using a deep neural network

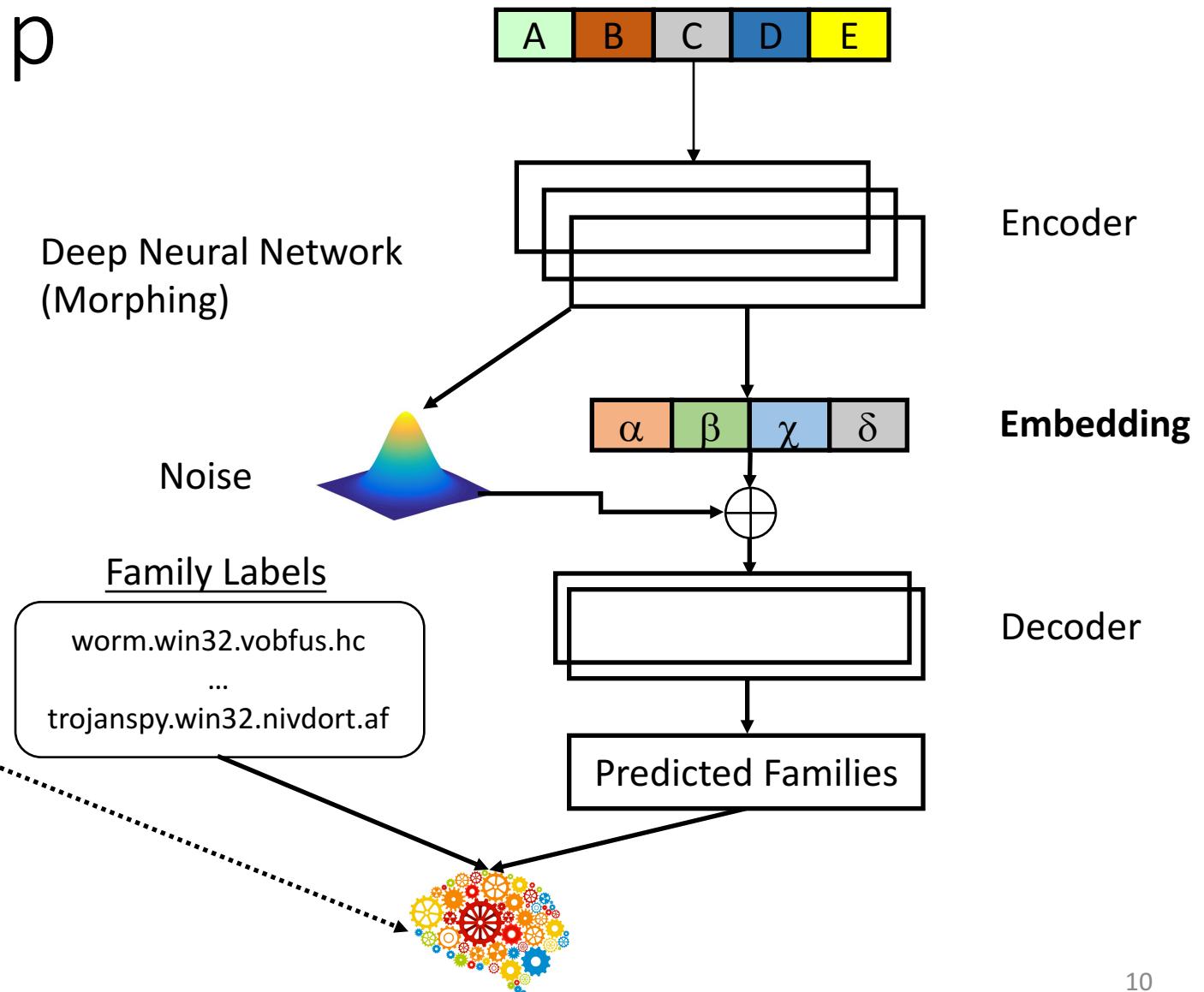
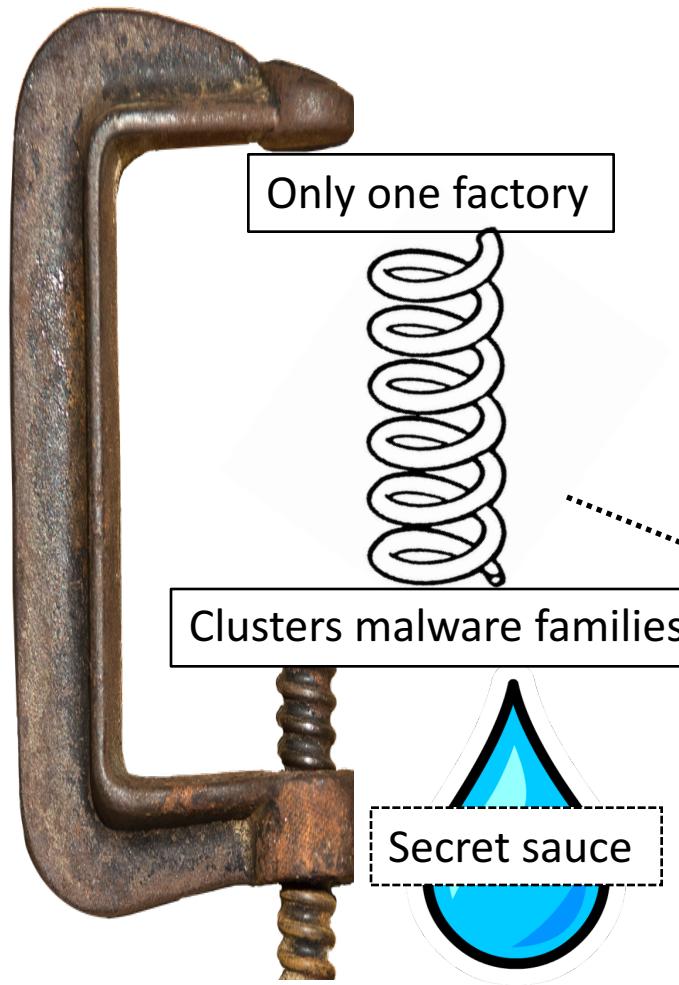


"The Matrix", 1999

Morphing the Map

Variational Autoencoder

Kingma, D. P., & Welling, M. (2013). *arXiv preprint arXiv:1312.6114*.



Toy Embedding Visualization

- Example



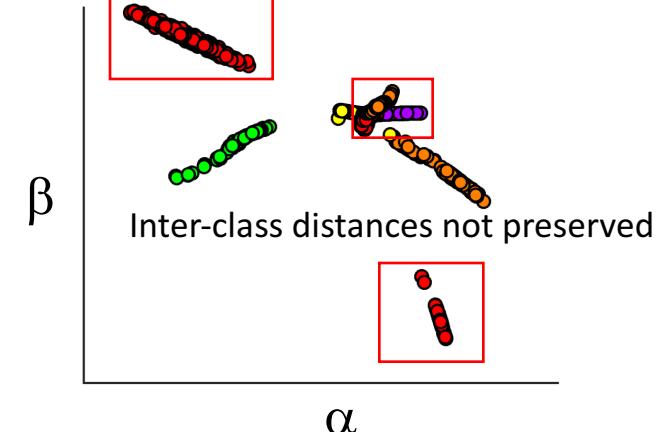
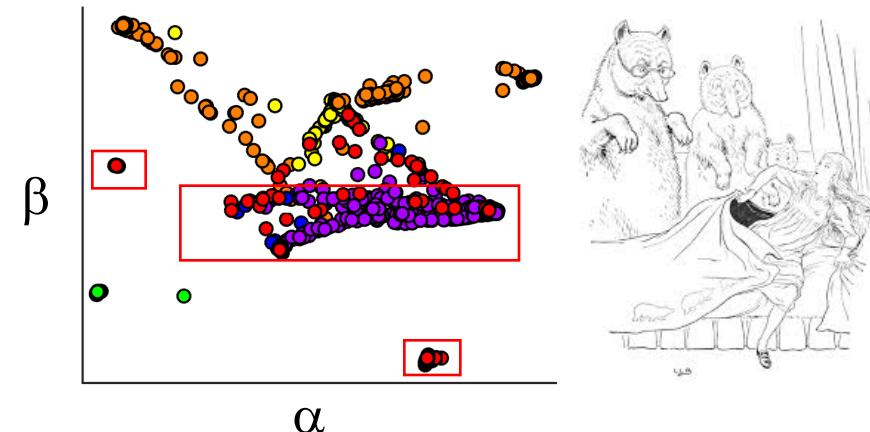
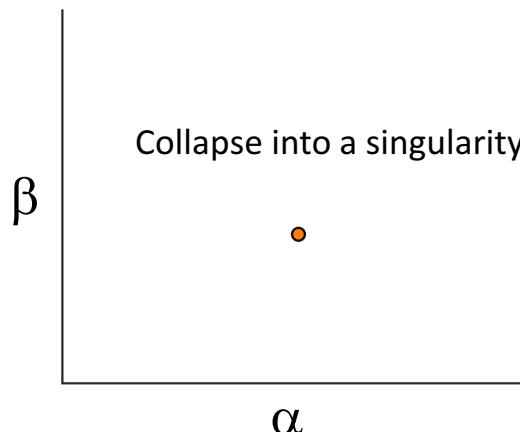
- 8 family/variant prediction
- 2D embedding

- virus.win32.nabucur.d
- virus.win32.ramnit.i
- virus.win32.sality.at
- virus.win32.shodi.i
- virus.win32.virut.ae
- virus.win32.virut.br
- virus.win32.virut.k
- worm.win32.allapple.a

Only one factory



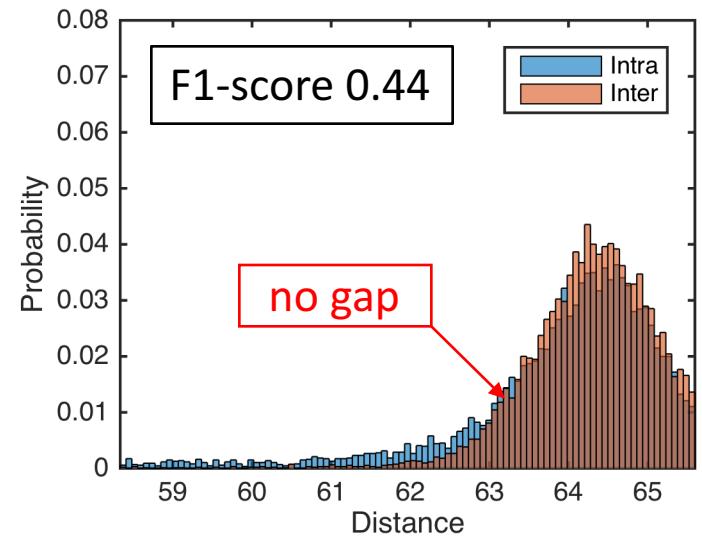
Clusters malware families



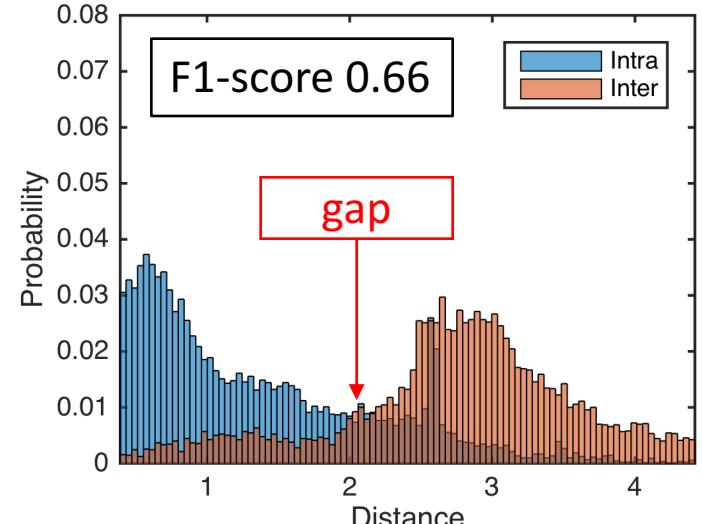
Results

- 800K samples
 - 1500 family/variants (99% coverage)
- Time-split Validation
 - Train on old data
 - Test on 30 days later
- Measure F1-score of 3-nearest neighbor classifier

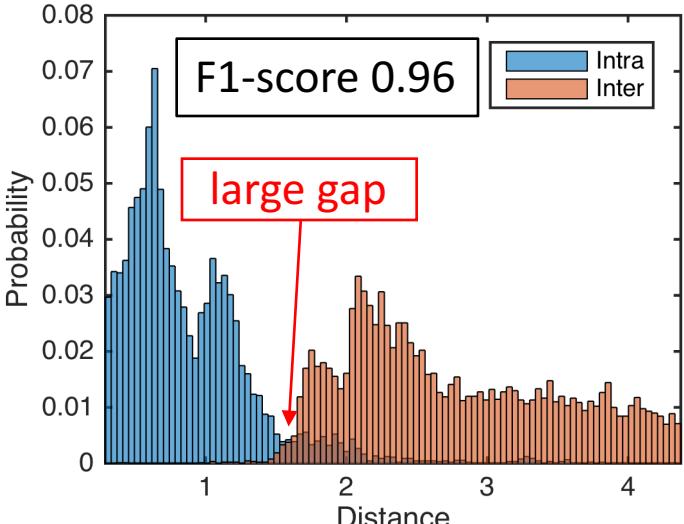
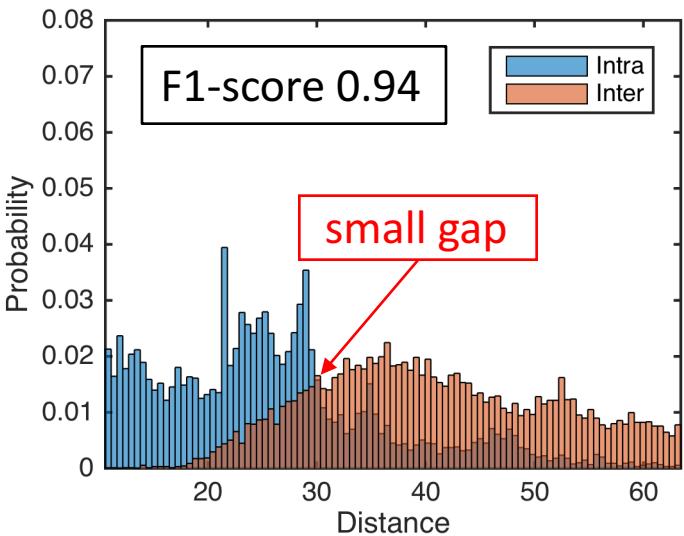
Attribute Vector



Semantic Embedding

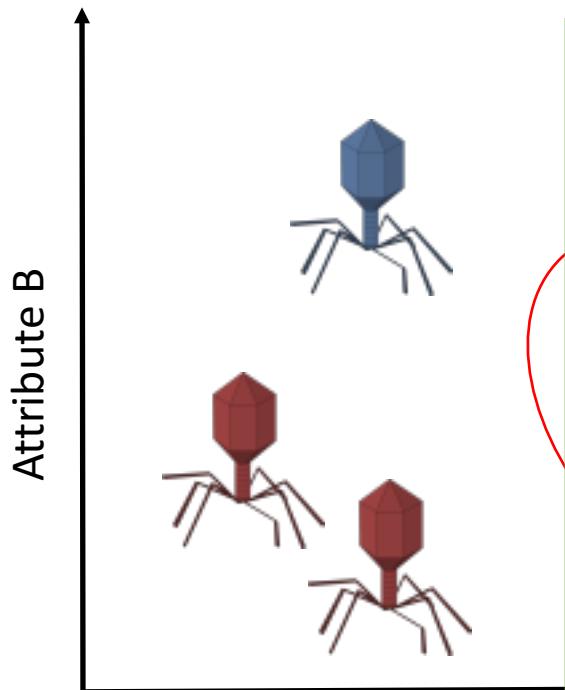


Deep-learning Features

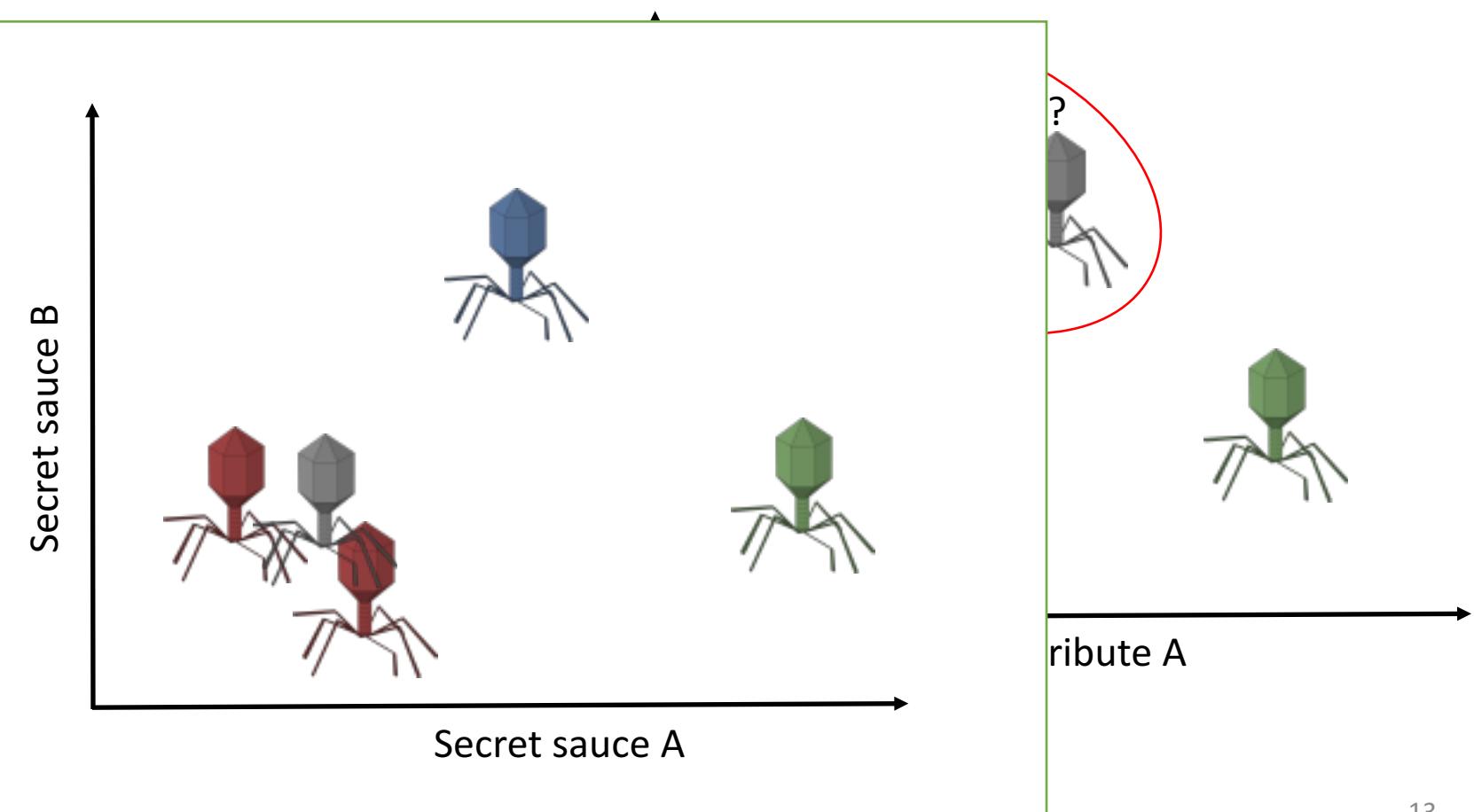


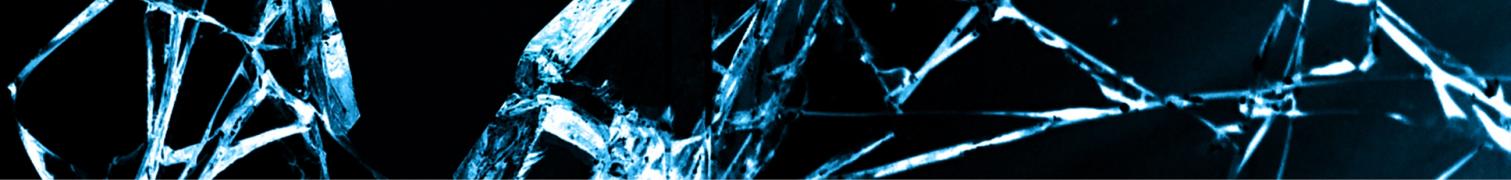
Issues with Attribute Maps

Possible Embedding #1



Possible Embedding #2





Conclusion

- Developing feature extraction is expensive and requires time consuming tuning to adapt to a specific domain
- Traditional approaches to malware similarity are hard to tune
- Using supervised-learning approaches we can improve existing features by embedding them into better maps
- **Automatic (re)tuning will improve attribution and reduce cost**

More Information

- Acknowledgement
 - Josh Saxe and Robert Gove
 - Invincea Inc.
- More information
 - Name: Konstantin Berlin
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- We are hiring!
 - Research Scientist
 - Senior Research Scientist
 - Principal Research Scientists