

# Describing Malware via Tagging

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In collaboration with:

Ethan M. Rudd, Alex Long, Tad Heppner, Konstantin Berlin



**SOPHOS**

# Malware Detection

- We already know that ML can be very powerful to detect malicious software.
  - Malicious / Benign, Confidence score

# Malware Detection

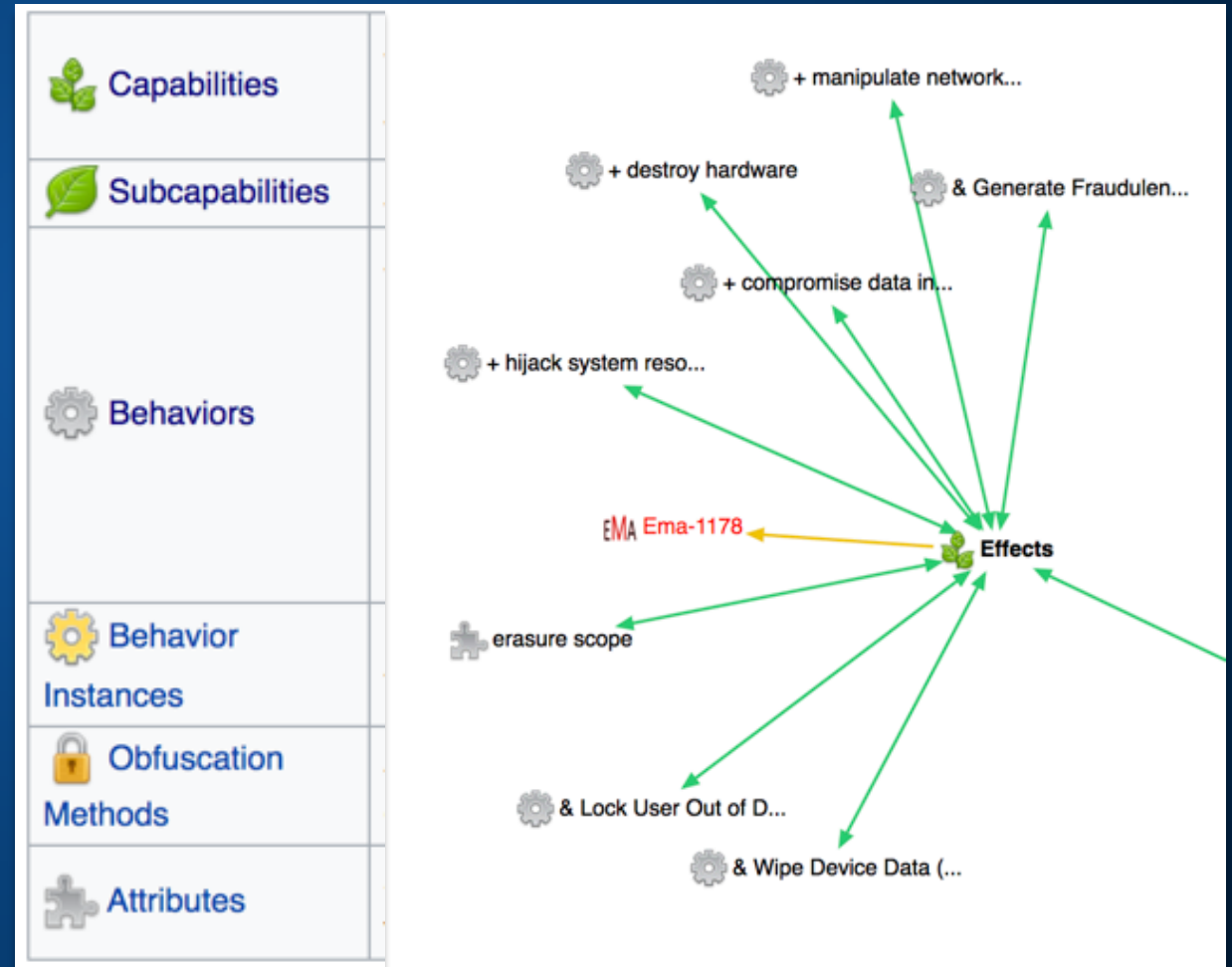
- We already know that ML can be very powerful to detect malicious software.
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# Malware Description

- Can we **also** use it to describe the type of malware it detected?

# Malware description: MAEC

- **Structured language** for encoding **high-fidelity** information about malware based on:
  - Behaviors
  - Artifacts
  - Relationships between samples



# Malware description: Families

- Trojan-FNET!CCD9055108A1
- Win32.Virlock.Gen.8
- W32.Virlock!inf7
- TR/Crypt.ZPACK.Gen
- Trojan ( 004d48ee1 )
- Virus:Win32/Nabucur.D
- W32/VirRnsm-F
- Virus.Win32.PolyRansom.k
- W32/Virlock.J
- a variant of Win32/Virlock.J

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ransom polyransom virlock virlock  
(ransomware)

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- virrnsn  
(file-infector)

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(ransomware)
- virrnsn  
(file-infector)
- crypt zpack  
(packed)



# Tags from Tokens

Detection name	Tokens	Tags
Artemis!4A26E203524C, Downloader, a variant of Win32/Adware.Adposhel.AM.gen, None, None, None, Gen:Variant.Razy.260309, None, Trojan ( 005153df1 ), Riskware/Adposhel	artemis, <b>downloader</b> , variant, win32, <b>adware</b> , <b>adposhel</b> , gen, gen, variant, razy, trojan, riskware, <b>adposhel</b>	<b>adware</b> <b>downloader</b>
W32.Virlock!inf7, TR/Crypt.ZPACK.Gen, Trojan ( 004d48ee1 ), Virus:Win32/Nabucur.D, W32/VirRnsm-F, Virus.Win32.PolyRansom.k, Win32.Virlock.Gen.8, W32/Virlock.J, Trojan-FNET!CCD9055108A1, a variant of Win32/Virlock.J	w32, <b>virlock</b> , inf7, tr, <b>crypt</b> , <b>zpack</b> , gen, trojan, win32, <b>nabucur</b> , w32, vir, <b>rsm</b> , <b>virrsm</b> , win32, poly, <b>ransom</b> , <b>polyransom</b> , win32, <b>virlock</b> , gen, trojan, variant, win32, <b>virlock</b>	<b>ransomware</b> <b>packed</b> <b>file-infector</b>

# Dataset Creation

- **Only malware samples** 
- **Train set**
  - M: 10,000,000
  - XL: 76,205,000
- **Val set**
  - 3,159,377
- **Test set**
  - 3,784,746

# Dataset Creation

- Only malware samples



- Train set

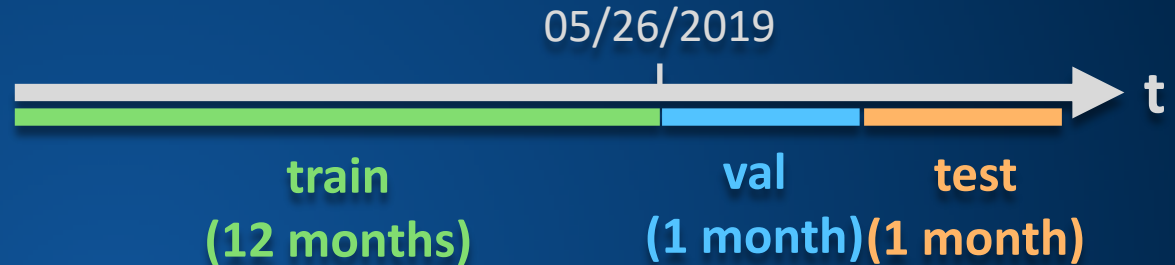
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- Only malware samples



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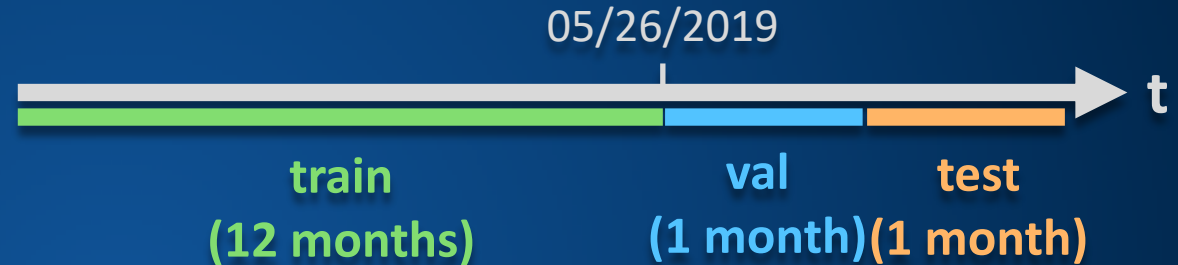
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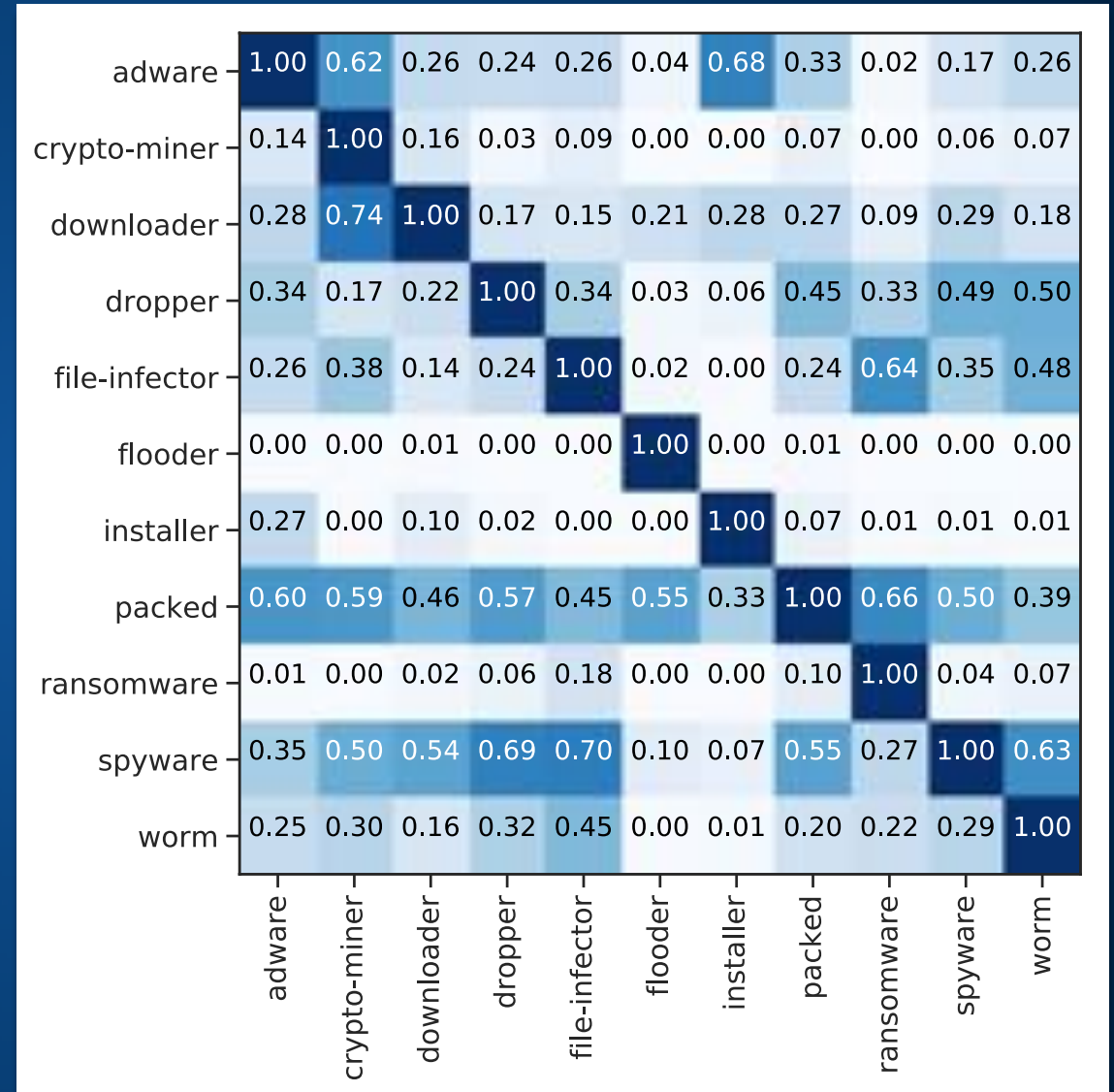
- 3,784,746



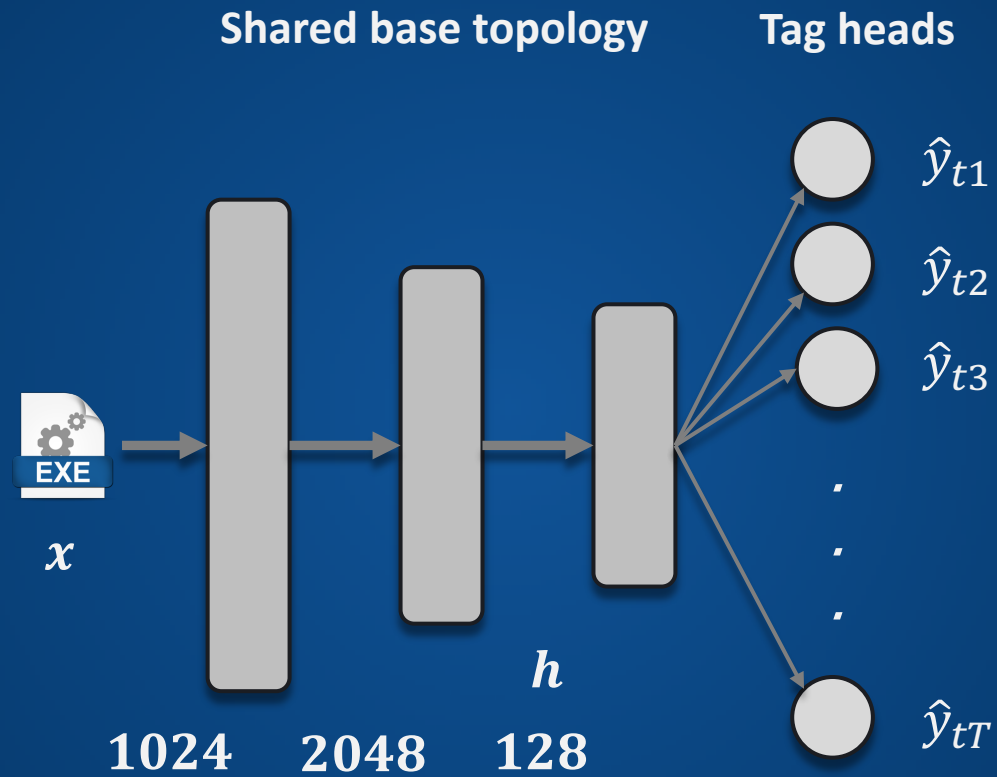
- 2D binary entropy **features** as in Saxe & Berlin, 2015.
- **Labels** from data-driven augmentation of token-to-tag mapping.

# Tags distribution

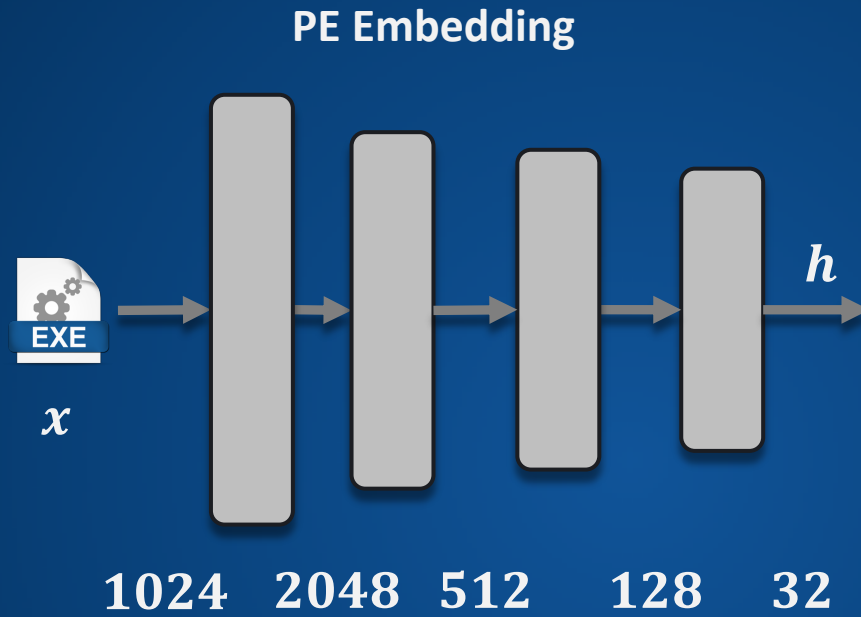
- **packed** 30.31%
- **spyware** 29.96%
- **dropper** 24.04%
- **worm** 20.63%
- **file-infector** 20.23%
- **downloader** 17.86%
- **adware** 16.95%
- **installer** 8.56%
- **ransomware** 8.13%
- **crypto-miner** 2.28%
- **flooder** 0.79%



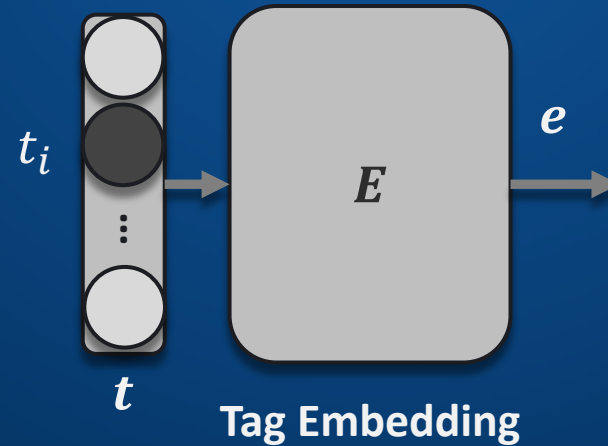
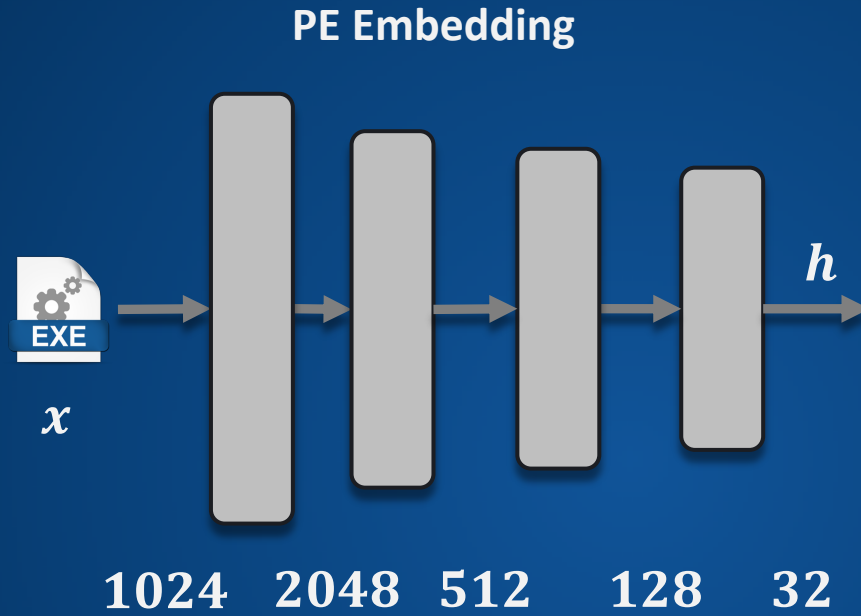
# Model 1: Multi-head



# Model 2: Joint Embedding

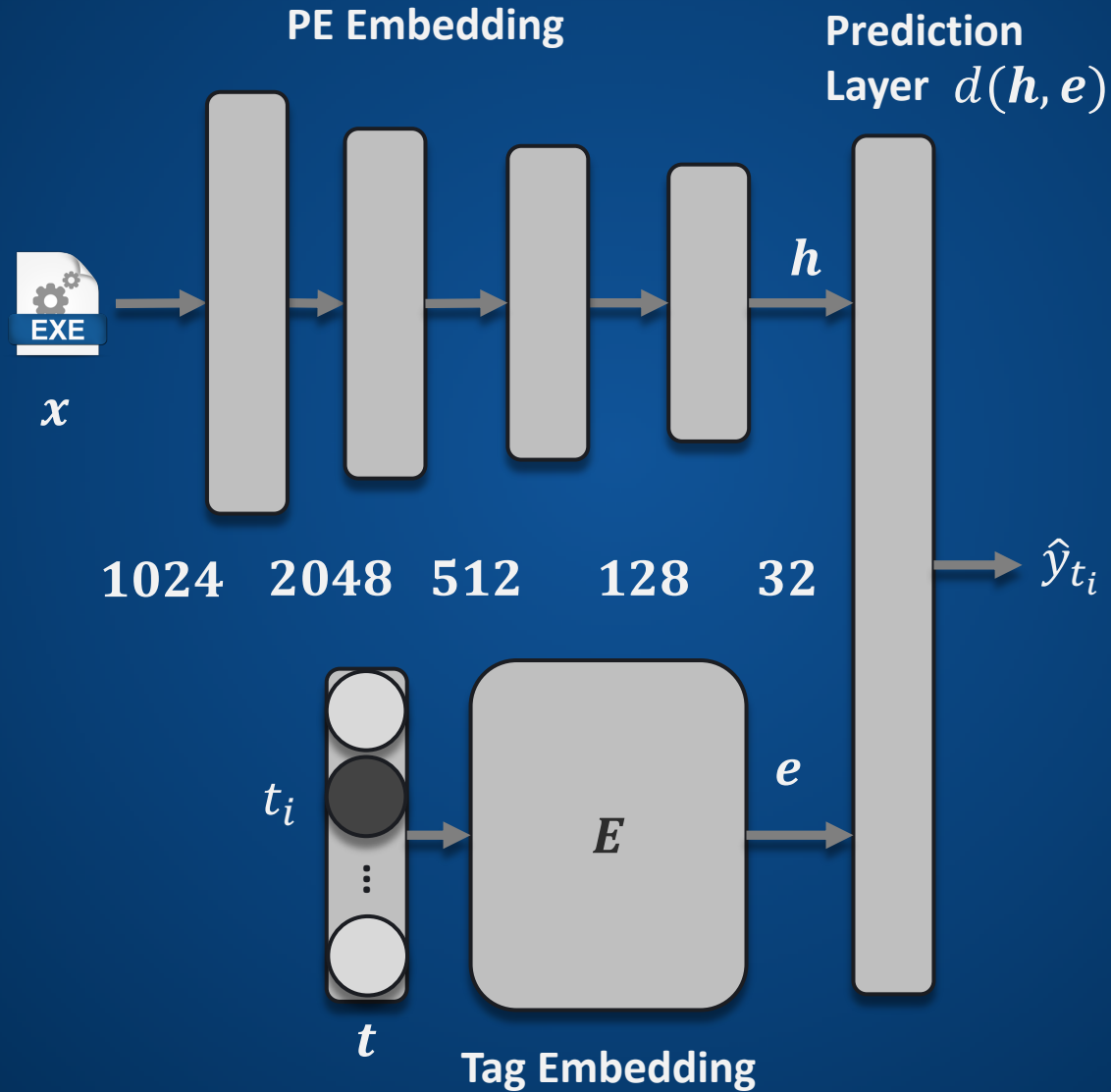


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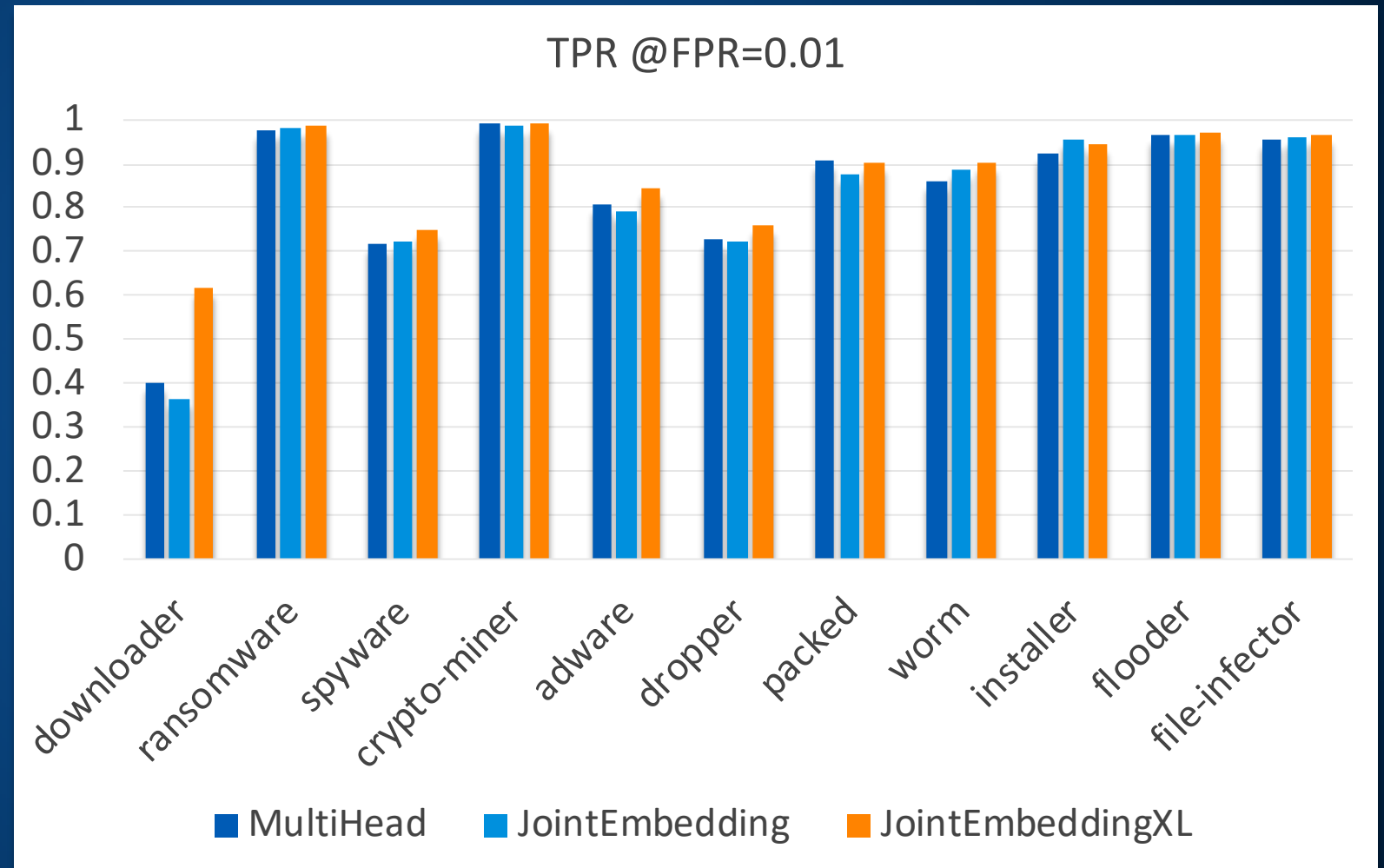


# Evaluation: Per Tag

- **JointEmbedding XL**

- Mean TPR: 0.88
- Mean AUC: 0.99

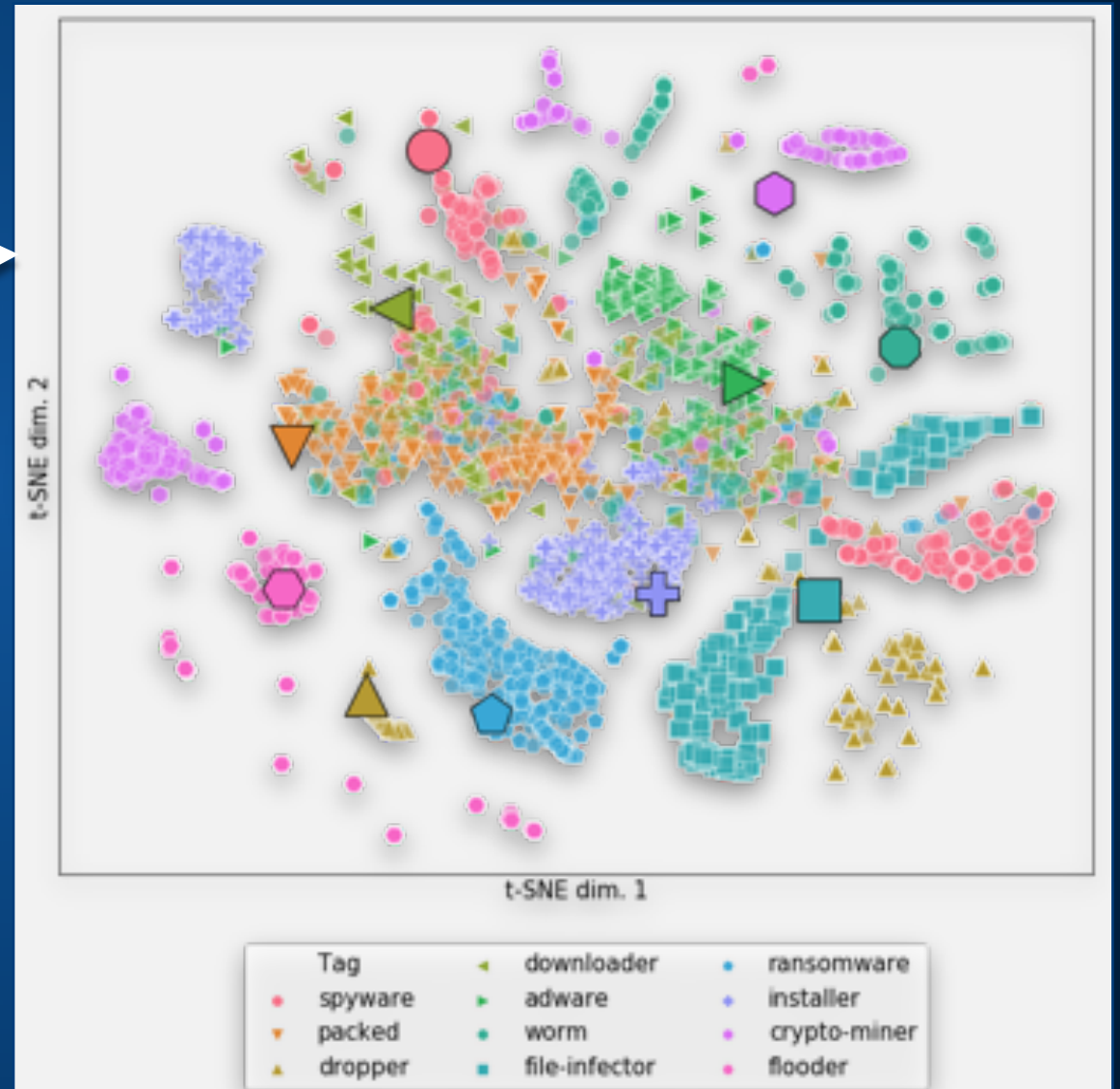
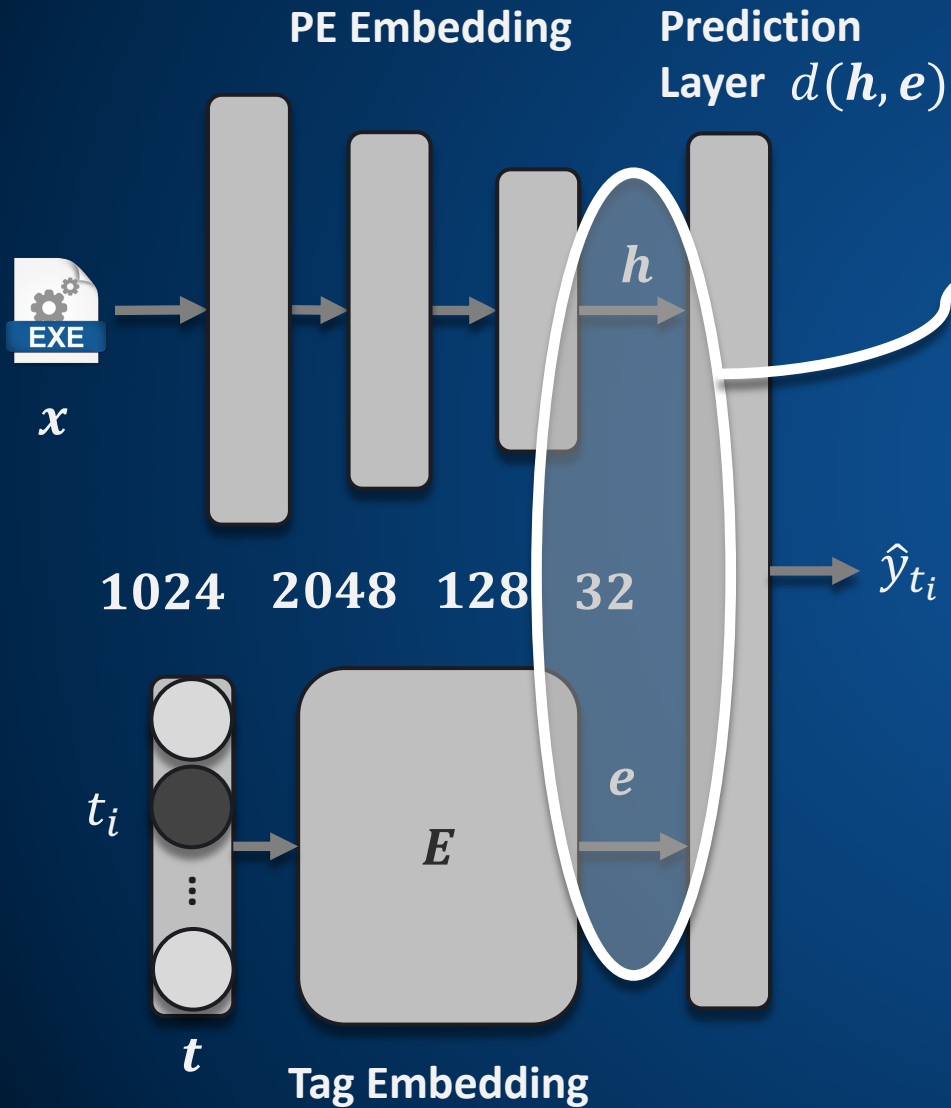
- 71% of the files for which all the tags are correctly predicted.



# Tagging : Use-cases

- Generate contextual information.
  - Provide to consumers
  - Use internally
- Ability of **prioritize** and **cluster** events based on threat characteristics.
- Common “light-weight” description framework.
  - Can be extended to various detection/analysis technologies.
- Improve standard Malware/Benign classification
  - **ALOHA: Auxiliary Loss Optimization for Hypothesis Augmentation**  
Ethan M. Rudd, Felipe N. Ducau, Cody Wild, Konstantin Berlin, and Richard Harang  
Usenix, 2019.

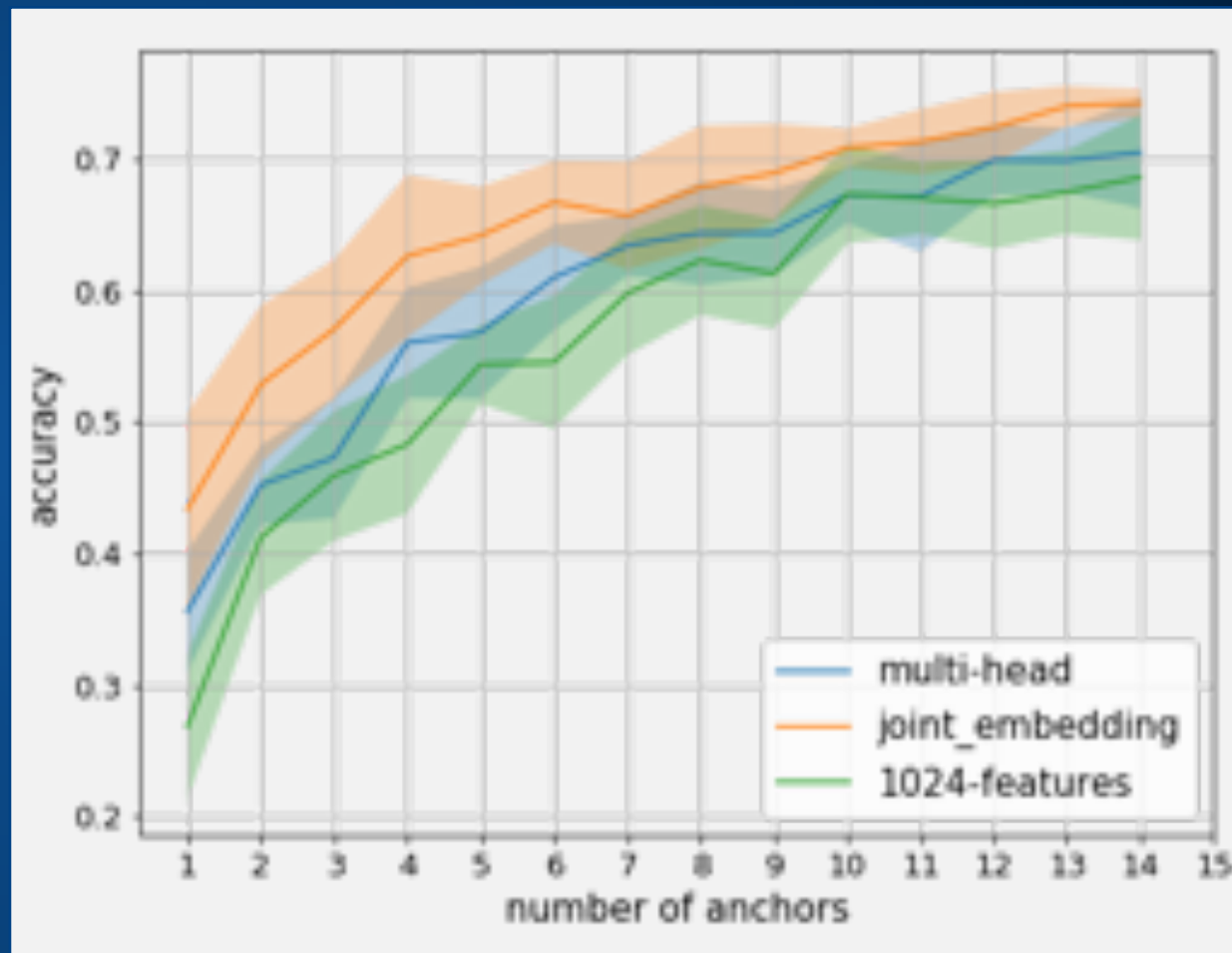
# Joint embedding space via t-sne



# Joint embedding space similarity index

- Select samples in test set for which we have a good knowledge about their families (~1,500).
  - upatre, bladabindi, nanocore, emotet, darkcomet, ...
- Measure similarity between test samples and **anchor** test samples.
- Predict **family** based on closest anchor in embedding space.

12-way accuracy



# That's all folks!

**SMART:** Semantic Malware Attribute Relevance Tagging

Felipe N. Ducau, Ethan M. Rudd, Alex Long, Tad Heppner, Konstantin Berlin

2019 pre-print, arXiv:1905.06262



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