

Responsible use of Artificial Intelligence to Enhance Society



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Garry Kasparov Avast Security Ambassador





1997

NATURAL 'PROZAC': DOES IT REALLY WORK?





Exem hum







2010/0C6206C6974746C65 16E642074616C7 FADE3100A16C20Data BreachE2048652 07 72 **Markin** 08 12202E6F6163686573204C697474CC 520 CBTOFFED DOBA 101 Cyber Attack696EA1 86FAF64206 E013921FC 1HC2 74023 106564207368 206E61C F766 6C792 Protection File Col B1 627 DeepAttacks 2AA261736B60142E204808100 EA BC010046368AF93010808B4FA017745C7A6 108B2C3 D5 1 0 0 FD C FOFOOFOOAFFA33C08E00F2A5697D011A56AFE64 07468652 C 1712E 4 81 LF1D01 02073 C732C20736852756B013 0AA21633 CT D O CI GEG (2001 A LE719 System Safety Compromised LA 200 CT avast





Rajarshi Gupta

Head of Artificial Intelligence, Avast



Data, data, data: What drives the world's largest consumer security network



MACHINE LEARNING

Every Month, Avast:









AVAST CLOUD ENGINE

Every Month, Avast:

Prevents +2 billion



malware attacks Pushes 50 PB of data

Monthly Engagement

> 10,000 Servers

Across 10 Locations Worldwide, processing monthly:

> 300M Files

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> 200Bn URLs



Dynamic malware detection engine

Purpose-built approach that takes < 12 hours to add new features, train, and deploy into production



COLLECTION

Goal: Harness as much data as possible

Avast6X more consumer PC users thanAdvantage:the nearest competitor⁽¹⁾

EXTRACTION

Deconstruct data into billions of artifacts

Proprietary Local Expert architecture leverages over 500+ features (e.g. size, origin, age, and file entropy)

TRAINING

Update models to understand the intention of a sample

New models can be trained on the entire historical dataset in less than 12 hours

EXECUTION

Precisely and quickly identify what is benign vs. malicious

Endpoint-based models are updated 200+ times per day



Al in Action: Using Neural Nets to Optimize the Classification Engine

Published at ICLR2018

Goal: augment our traditional handcrafted models with machine-generated features

- Train a Convolutional Neural Net using the raw sequence of bytes from the binary files
- Training set of 20 million Windows PE files

Results

- · Raw model achieves comparable accuracy to hand crafted features
- · Choosing machine-generated features makes it much harder to evade
- · Enriched features model shows extra gain of using both sets of features

CLASSIFIER	ACCURACY
MalConv	94.6%
Avast Raw Files ConvNet	96.0%
FNN on features crafted manually	96.2%
FNN on enriched features	97.1%



Al in Action: Deep Neural Net for IoT Traffic

Input: Flow statistics from millions of homes

Input device traffic information for many devices, in many homes over a long time period



Multi-Level Model

DEVICE TYPE IoT devices have very limited behavior

Identifying devices allow us to model their behavior



SERVICE TYPE

Many devices plus internet makes up services, e.g. Netflix



May focus on a device type, or servicetype

Output: Autonomously identify anomalous traffic

- Recognize unknown attacks •
- · Identify the device or service causing the attacks

Using AI to Defend Against AI-based DeepAttacks







Thank You

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Smart Home Networks in Mexico

In Dec 2018:

- Avast scanned 2.5M devices in 390K homes
- 47% of homes have at least one vulnerability
- 28% of homes have at least one router vulnerability
- 7% of home routers have weak WiFi password





Al is Security's Greatest Opportunity

VOLUME

SCALABILITY

Coping with the sheer volume of new threats would be impossible without ML

VELOCITYSPEEDMost threats have very shortlongevity; machines can act muchfaster

VARIETY

ACCURACY

ML is also really good at taking into account large amounts of contextual data

