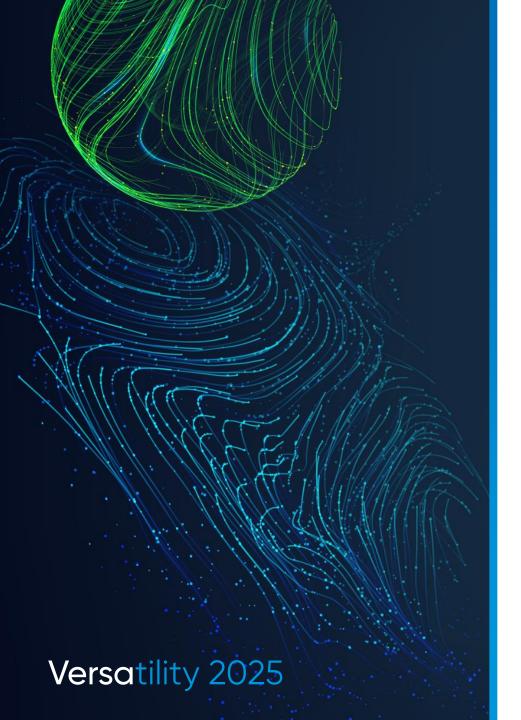


AI/ML for security use cases





Agenda

- Pain points in cybersecurity
- Key components
- AI/ML for Advanced Threat Prevention (ATP)
- AI/ML for Data Loss Prevention (DLP)
- AlOps for cybersecurity
- Questions



Today's Pain Points - Threat Prevention(ATP)

SaaS/laaS/PaaS sprawl – Increasing Attack surface

- Security teams overwhelmed by alert volume.
- Difficult to identify genuine threats amidst the noise.

Financial impact - TTRR

- Average cost of a data breach crossed \$4.5m in 2023, a 15% YoY increase.
- 277 days to Respond and Remediate breaches (204 to Respond, 73 to Remediate) in 2023.
- Loss of brand confidence.

Multiple point products - Integration complexity

- On average there are 76 different security solutions which creates problem of interworking the disparate security solutions
- 54% of SoC teams feel burnt out fielding security events, 45% of which were false positives.

Attack sophistication

- AI-created malware, zero-day exploits used by attackers to launch sophisticated and faster attacks.
- Ransomware attacks increased 24% YoY costing around \$5.3m to Respond and Remediate.



Today's Pain Points - Data Loss Prevention(DLP)

Data Visibility & Classification

- Identifying sensitive data across complex environments (on-prem, cloud, SaaS) is challenging.
- Complexities in data classification is a primary barrier to effective DLP implementation

Balancing Security & Productivity

- Strict DLP rules can block legitimate business activities.
- Employees/Users could potentially use insecure workarounds ("shadow IT").
- Tuning DLP policies to be effective without hindering productivity is difficult.

Data Protection in hybrid work environments

- Data sprawl across cloud services (IaaS, PaaS, SaaS) and accessed by remote workers.
- Data breaches cost about USD 1 million more on average at organizations.

Insider Threats

- Threats originating from within the Org (malicious or accidental) are harder to detect.
- Internal "privilege misuse" and "error" patterns remain significant factors in breaches.



Versa AI for file/stream-based Security – Key Components

Threat Protection

Data Protection

AI/ML Based Malware Detection

What is it?

Multi-stage AI/ML for realtime processing of files to identify malware



Key benefits

Eliminates zero-day attacks covering **90 %** of file types used to transmit malware while reducing sandbox load by **75%**

AI/ML Based Data Loss Prevention

For dynamic, adaptive protection from data loss (unintentional or malicious)



✓ Detection of sensitive data in text documents and image – enabling more accurate and dynamic data security.





AI/ML based Malware Detection — 1,000 ft view

Reduce reliance on static signatures:

- ➤ AI/ML-based malware detection systems can identify patterns and characteristics common to malware behaviors.
- Extract the way malware interacts with the system, such as auto-execution, data encryption, C&C communication etc.

AI/ML-based malware detection for common file-types:

- PE (Portable Executable)
- Docx, Pptx, Xlsx, RTF
- HTML/JS (JavaScript).

AI/ML-based lightweight inference at the Edge:

- Subset of file-types.
- Down-select files.
- Sent to cloud ATP for analysis beyond certain confidence score.



Augment signature-based detection

Multi file-

type support

AI/ML at the

Edge



models







Distinct ML models trained for each file-type

- Gradient-boosted trees.
- Deep Neural Networks.
- Multi-model Ensembles.
- Fine-tuned State-Of-The-Art transformer models.

AI/ML microservice – Integral part of **ATP**

- Receive file analysis request -> launch inference for file-type.
- Return results, [Clean, Suspicious, Malicious] with appropriate metadata.
- Depending on the result and confidence score one of the following actions:
 - ✓ Subsequent multi-vendor AV or sandboxing is triggered within ATP.
 - ✓ The result is returned instantly, and rest of the analysis is skipped.



Traditional Malware Detection

Reactive Detection

Traditional methods run regex/pcre on known signatures, yara rules etc. to identify malware which are error prone.

• Static Analysis

Examine and evaluate the code of a software program without executing

Signature based, Heuristic analysis, File hashing, Resource analysis, Static code analysis tools

• <u>"Point in Time" Remediation</u>

Detection and remediation based on known attacks at a specific point of time (e.g. signatures) which may become outdated

Traditional Malware Detection: Unable to detect Zero-Day
Exploits, Polymorphic and Metamorphic Malware and
Advanced Persistent Threats (APT) Identification

VERSA AI/ML BASED MALWARE DETECTION

Proactive Detection

- ✓ Al can identify threats based on behavior, catching zero-day and polymorphic malware.
- ✓ Eliminates zero-day attacks covering 90% of file used to transmit malware while reducing sandbox load by 75% (e.g PE, PDF, docx, pptx, xlsx, JS etc.)

Continuous Learning & Real Time Analysis

- ✓ Adapts over time to new threats from updated data and models, improving its detection capabilities for zero-day.
- ✓ Ensemble of Models: Trained gradient-boosted trees, Deep Neural

 Networks as well as fine-tuned transformer models for the different filetypes.

Sophisticated Threat Identification & Remediation

- ✓ Ingest file from any source (branch, users or SASE GWs, cloud services), with analysis & remediation using ATP cloud.
- ✓ Detect complex attack patterns and threats by training on diverse data.

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Malware Analysis - Structured

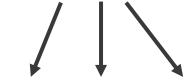


Data ingestion: Update data warehouse based on streaming data



- Data selection(coarse and granular).

- Feature ETL
- Prepare for training







Generate AI/ML models:

- Decision-tree XGB, RFC etc
- **DNNs**





Update metrics in MLOps dashboard for continuous training



Create an ensemble of models for the use case



Rate the models on performance and accuracy



Malware Analysis - Unstructured





Data ingestion: Update data warehouse based on streaming data

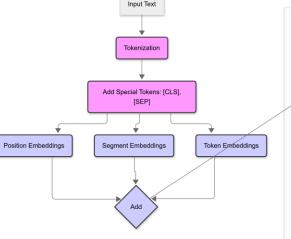


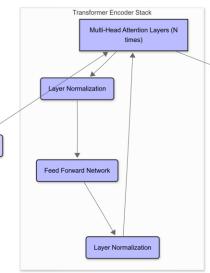
Update metrics in MLOps dashboard for continuous training

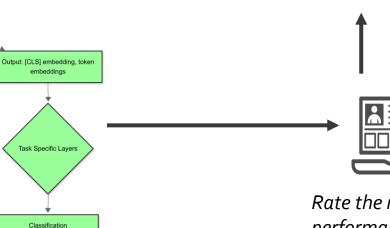


- Create an ensemble of models for the use case.
- 2) Optimize for inference
 - Deploy in ATP for inference

- 1) Preprocess unstructured data (ETL).
- 2) Train transformer models on processed data.



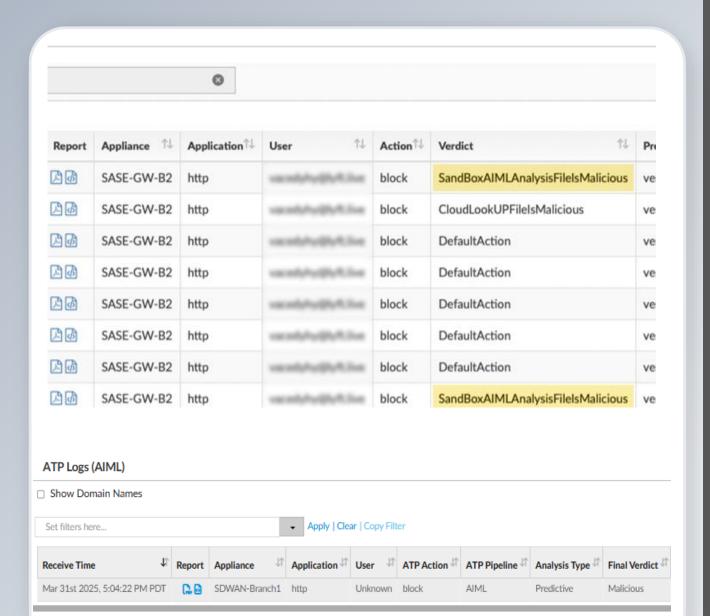




Rate the models on performance and accuracy







Showing 1 to 1 of 1 entries

Al based Malware Detection

- VersaAI™ deploys multi-stage AI/ML for realtime identification of malware
- Eliminates zero-day attacks covering 90 % of file types used to detonate malware while reducing sandbox load by 75%. Analytics dashboard shows detection by AI/ML and consequent policy action
- Consumed as part of ATP cloud (GA) as well as on-premise (2H'25).





AI/ML based DLP – high-level components

DLP for multiple document types: - Text, PDF, DOC(docx, pptx, xlsx), Images etc.

<u>Image and Text pre-processing</u> for DLP training and inference

<u>Bleeding edge transformer models</u> used to detect the type of Enterprise document (e.g. Credit cards, Passport, source-code,) and extract multi-modal content.

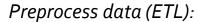
<u>Ability to fine-tune LLMs and smaller-footprint NER models</u> for classification and detection of sensitive/PII for customer-specific data.



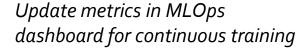
AI/ML based DLP - Model Training



Ingest data from various sources (Text, Image etc)



- Clean
- Normalize
- Upscale
- Augment
- Synthesize





- 1) Create an ensemble of models for the use case
- 2) Optimize for inference
- Deploy in cloud/on-prem for inference





Train AI/ML models:

- Embeddings models
- CNN based classifiers.
- NER models
- Transformer/LLM based models or its affiliates. All rights reserved.



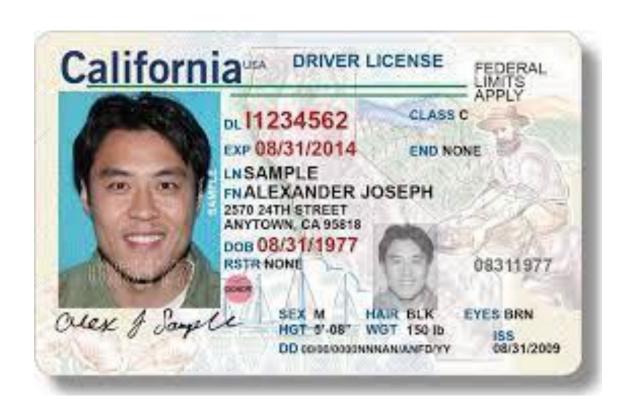
Rate the models on performance and accuracy



Versatility 2025

DLP Extraction and Classification – Example 1

```
{"detection":{
  "Enterprise": [
      "Document Type": "PERSONAL-IDENTITY",
      "Confidence Score": 99.0
  "Generic": [
      "Classification": {
        "Document Type": "DRIVER LICENSE"
      "Extraction": [
          "Identifier type": "NAME",
           "Identifier": "ALEXANDER JOSEPH",
           "Confidence Score": "100%",
           "Sensitivity": "high"
          "Identifier type": "DATE OF BIRTH",
           "Identifier": "08311977",
           "Confidence Score": "100%",
           "Sensitivity": "high"
```





DLP Extraction and Classification – Example 2

```
"Enterprise": [
    "Document Type": "FINANCE-REGULATORY",
    "Confidence Score": 98.5,
    "Sub-Document": {
      "Classification": {
         "Sub-document type": "Individual Tax Returns"
      "Extraction": [
           "Identifier type": "Taxpayer Name",
           "Identifier": "JOHN TAXPAYFR",
           "Confidence Score": 80,
           "Sensitivity": "Low"
           "Identifier type": "Taxpayer Social Security Number",
           "Identifier": "002-21-1252",
           "Confidence Score": 90,
           "Sensitivity": "High"
```

For	the year	Jan. 1-Dec. 31, 2010, or other tax year beginning , 2	010, ending		, 20	OMB No. 1545-0074
Name, R					Yourse	ocial security number
Address, N					002-21-1252	
and SSN 🖟 JO	HN '	TAXPAYER			Spouse's social security no. 010-25-5545	
See		acm 10 cmprem apm F				
eparate A 10 EAST 10 STREET APT 5 NEW YORK NY 10003				A	Make sure the SSN(s) above and on line 6c are correct Checking a box below will not change your tax or refund.	
Presidential						
Election Campaig		Check here if you, or your spouse if filing jointly, want \$3 to go to				ou Spouse
Filing Status	1 2	H Single Married filing joints annipled Return	the qualifying pers			person). (See inst.) If
_	3	X Married filing separately. Enter spouse's SSN above	enter child's name			not your dependent,
Check only one box.		and full name here. ► MARY MAYER 5				nt child (see instructions
Exemptions	6a	X Yourself. If someone can claim you as a dependent, do it				Boxes checked 1
If more than four dependents, see	b	Spouse				No. of children
dependents, see inst. & check here	С	Dependents: (2) Dependent's	(3) Dependent's	(4) √ if d	hild under	on 6c who:
(1) First nan	ne	Last name social security number	relationship to you	fying for ch credit (see	ild tax inst.)	did not live with you due
						to divorce or separation
						(see inst.) Dependents on 6c
						not entered above
						Add numbers on
	d	Total number of exemptions claimed				lines above
	7	Wages, salaries, tips, etc. Attach Form(s) W-2			7	
Income	8a	Touchts Island Attack Cabadala Differential			8a	250
Attach Form(s)	ba b	Taxable interest. Attach Schedule B if required	8b 6,	. 885	88	330
W-2 here. Also attach Forms W-2G and 1099-R if tax was withheld.	9a	Tax-exempt interest. Do not include on line 8a			9a	6,491
	b		9 b 5,	907	5a	0,421
	10	Taxable refunds, credits, or offsets of state and local income ta			10	1,359
	11	Alimony received			11	1,000
	12	Business income or (loss). Attach Schedule C or C-EZ			12	208,621
	13	Capital gain or (loss). Attach Schedule D if required. If not requ			13	-1,500
If you did not get a W-2, see instructions.	14	Other gains or (losses). Attach Form 4797			14	-27
	15a		Taxable amount		15b	
	16a	Pensions and annuities 16a b	Taxable amount		16b	
	17	Rental real estate, royalties, partnerships, S corporations, trust-			17	71
Enclose, but do not attach, any payment. Also, please use Form 1040-V.	18	Farm income or (loss). Attach Schedule F			18	
	19	Unemployment compensation			19	
	20a		Taxable amount		20b	
	21	Other income. NYC UBT REFUND 4971			21	4,971
	22	Combine amounts in the far right column for lines 7 through 21.		come▶	22	220,344
0 -1:	23		23		-	
Adjusted Gross	24	Certain business expenses of reservists, performing artists,	24			
Income	25		25		-	
income	26		26		-	
	27			.163	-	
	28		28	, 100	1	
	29		29 18.	811	1	
	30		30		1	
	31a		31a		1	
	32		32 5,	,000	1	
	33	Student loan interest deduction	33			
	34		34			
	35	Domestic production activities ded. Attach Form 8903	35			
	36	Add lines 23 through 31a and 32 through 35			36	32,974
	37	Subtract line 36 from line 22. This is your adjusted gross inco	me	▶	37	187,370



Traditional DLP

• No "context" in analysis

Relies on pre-defined pattern matches which may lead to false positives or negatives in certain attack scenarios

Static Detection Techniques

- a) Keyword Matching
- b) RegEx for Pattern Recognition (Social Security #s)
- c) Document Fingerprinting (unique hash or "fingerprint")
- d) File Type and Metadata Analysis
- e) Data classification by Tagging
 Leads to unnecessary False Positives, noisy alerting and can
 be easily circumvented.

• <u>Limited Adaptation</u>

Non-trivial real-time updates to DLP signatures (user or predefined)

VERSA AI/ML BASED DLP

Enhanced Detection with Contextual Analysis

✓ Understands the context (spatial or temporal) of data usage, allowing for more nuanced protection strategies that adapt to different scenarios.

Dynamic Adaptation with Real Time Protection

- ✓ Ability to fine-tune and learn from new data, improving its understanding of what constitutes sensitive information and potential threats over time
- ✓ Analyze data in-motion in real-time, providing immediate response to potential data leaks or unauthorized access.

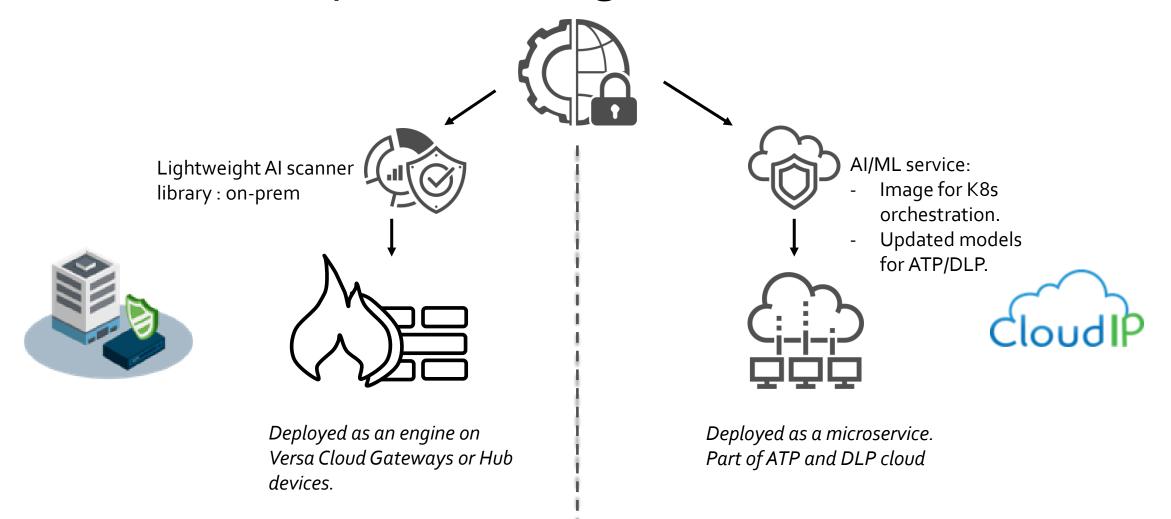
Advantage AI/ML

- ✓ Detection of Sensitive Information in Unstructured Data
- ✓ Adaptive Protection Against Evolving Data Types and Policies
- ✓ Anomaly Detection in Data exfiltration using UEBA triggered user risk analysis.

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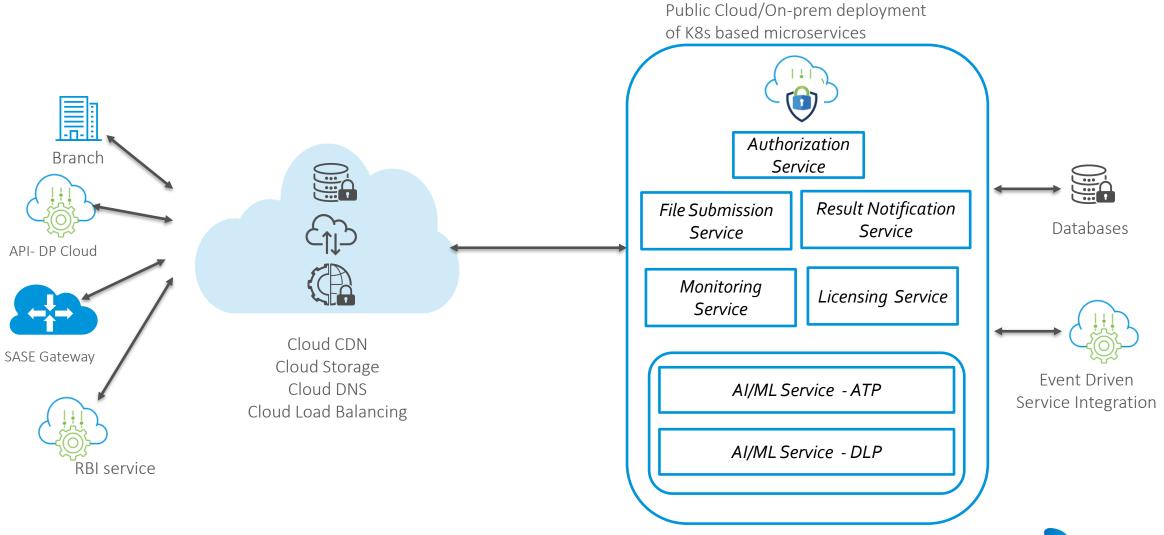
AIOps for ATP/DLP Versatility 2025

Model Consumption Strategies





Public Cloud Based AI/ML Services Flow Diagram



Versatility 2025

Public Cloud Based AI/ML ATP/DLP Data Pipeline

