

Becoming the Trainer: Attacking ML Training Infrastructure



Brett Hawkins (@h4wkst3r)

Adversary Services, IBM X-Force Red



Blog Post and Tool



Blog Post



MLOKit

Agenda



1. Introduction

2. Background

3. Attacking ML Training Environments

- Attack Scenarios
- Demos

4. Protecting ML Training Environments

5. Conclusion

6. Q&A

Introduction



Who am I – Brett Hawkins

<https://h4wkst3r.github.io>



Current Role

Team Lead,
Adversary Services
IBM X-Force Red



Open-Source Tool Author

SharPersist,
InvisibilityCloak,
SCMKit, ADOKit,
MLOKit



Conference Speaker

Black Hat (US & EU),
BlueHat,
ShmooCon,
DerbyCon, Wild
West Hackin' Fest,
BSides, Hackers
Teaching Hackers

Research Drivers



Threat actors
targeting AI/ML
environments



Lack of research on
attacking and
defending ML
infrastructure



Adoption of ML
technologies by
enterprises



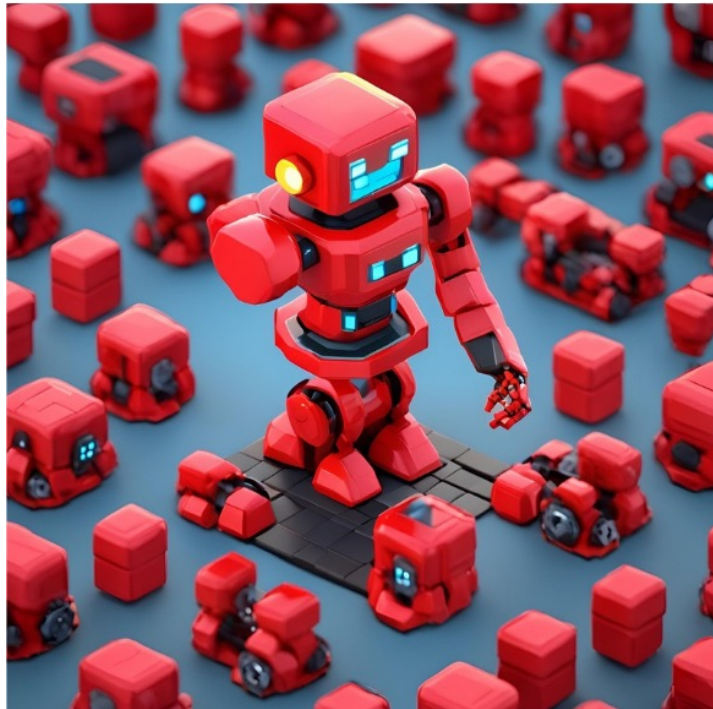
Lack of tooling to
simulate attacks
against platforms
used in ML
training envs

Threat Actor Motivation



Theft of models and weights, backdooring models for initial access or persistence, expanding access via lateral movement and privesc, sensitive data theft or deploying ransomware, model modification/poisoning for misclassification, degradation, fraud or ml-based detection evasion.

Attendee Takeaways



How to steal
models from
model registries



How to poison
models within ML
training platforms



How to defend
key components
of ML training
infrastructure



How to get code
execution via
attacks on ML
training
infrastructure

What is new in this research?

```
[*] INFO: Listing Model Artifact Location Info:
Account Name: testworkspace5178193999
Datastore Type: AzureBlob
Container Name: azureml
Path: ExperimentRun/dcid.AutoML_91114fd1-6657-4bf0-b51d-6f868e2c2033_42/outputs/mlflow-m
[*] INFO: Getting associated datastore for model artifacts:

      Account Name | Container Name | D
-----|-----|-----
testworkspace5178193999 | azureml | D
[*] INFO: Uploading model artifacts
[*] INFO: Uploading: conda.yaml
[*] INFO: Uploading: MLmodel
[*] INFO: Uploading: model.pkl
[*] INFO: Uploading: python_env.yaml
[*] INFO: Uploading: requirements.txt
[+] SUCCESS: Model has been poisoned with model artifacts specified in source directory
```



Advanced attacks
against ML training
environments



New detection
rules (Azure ML
and SageMaker)



MLOKit tool
updates – NEW
supported
platforms and
attacks

My Perspective



I am

Offensive
Cybersecurity
Specialist

I am not

Data Scientist

AI/ML Engineer

Cloud Engineer

Detection Engineer

DevOps Engineer

Software Engineer

Background



Prior Work

Links to prior work are provided in appendix slides

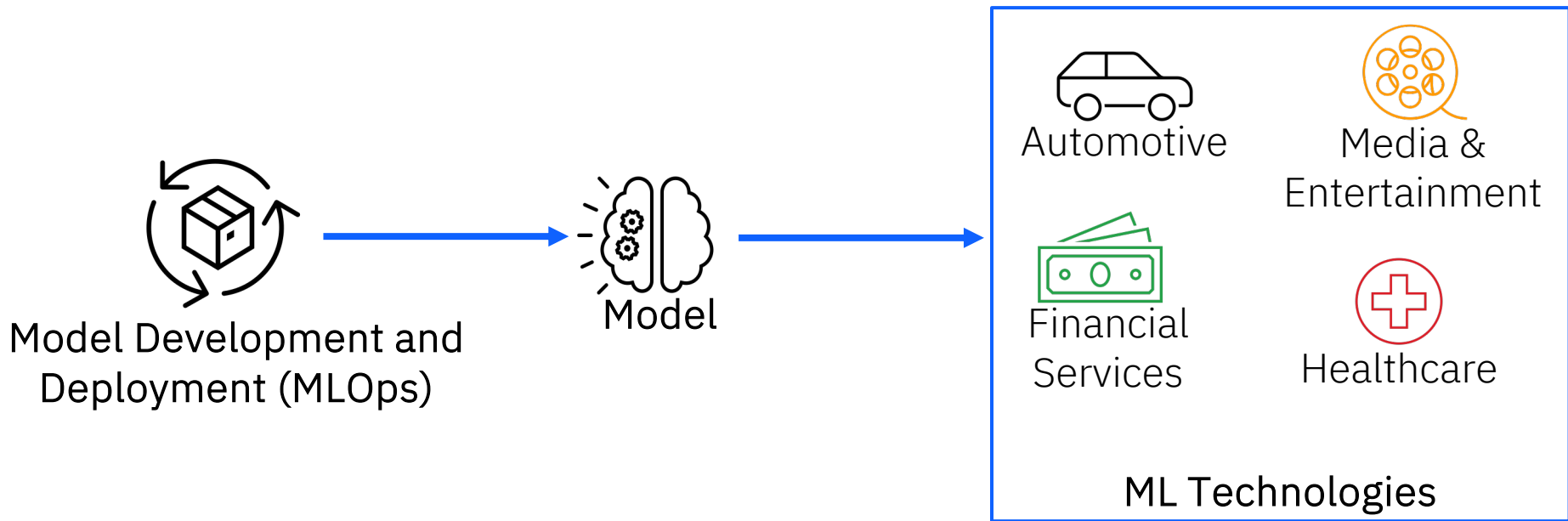
Chris Thompson ([@retBandit](#)) & I – ShmooCon 2025

Disrupting the Model: Abusing MLOps Platforms to Compromise ML Models and Enterprise Data Lakes

Or Azarzar ([@azarzaror](#)) – Blog Post (2021)

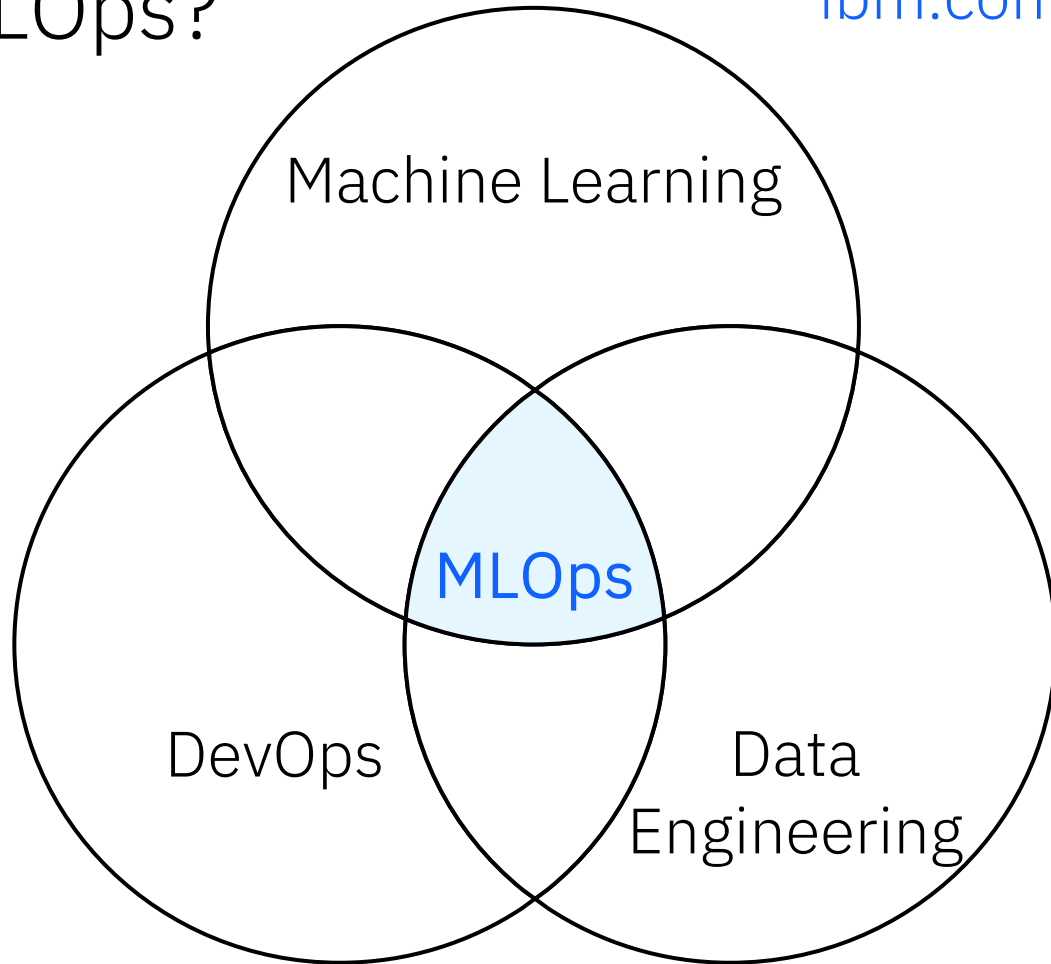
Protect Your Environment When Working with Amazon SageMaker

ML Technology Use Cases

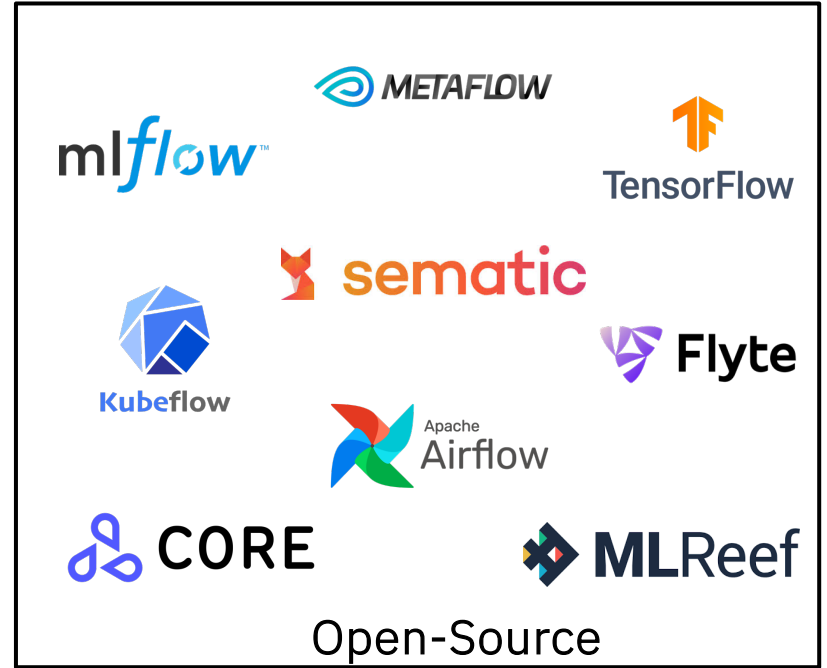


What is MLOps?

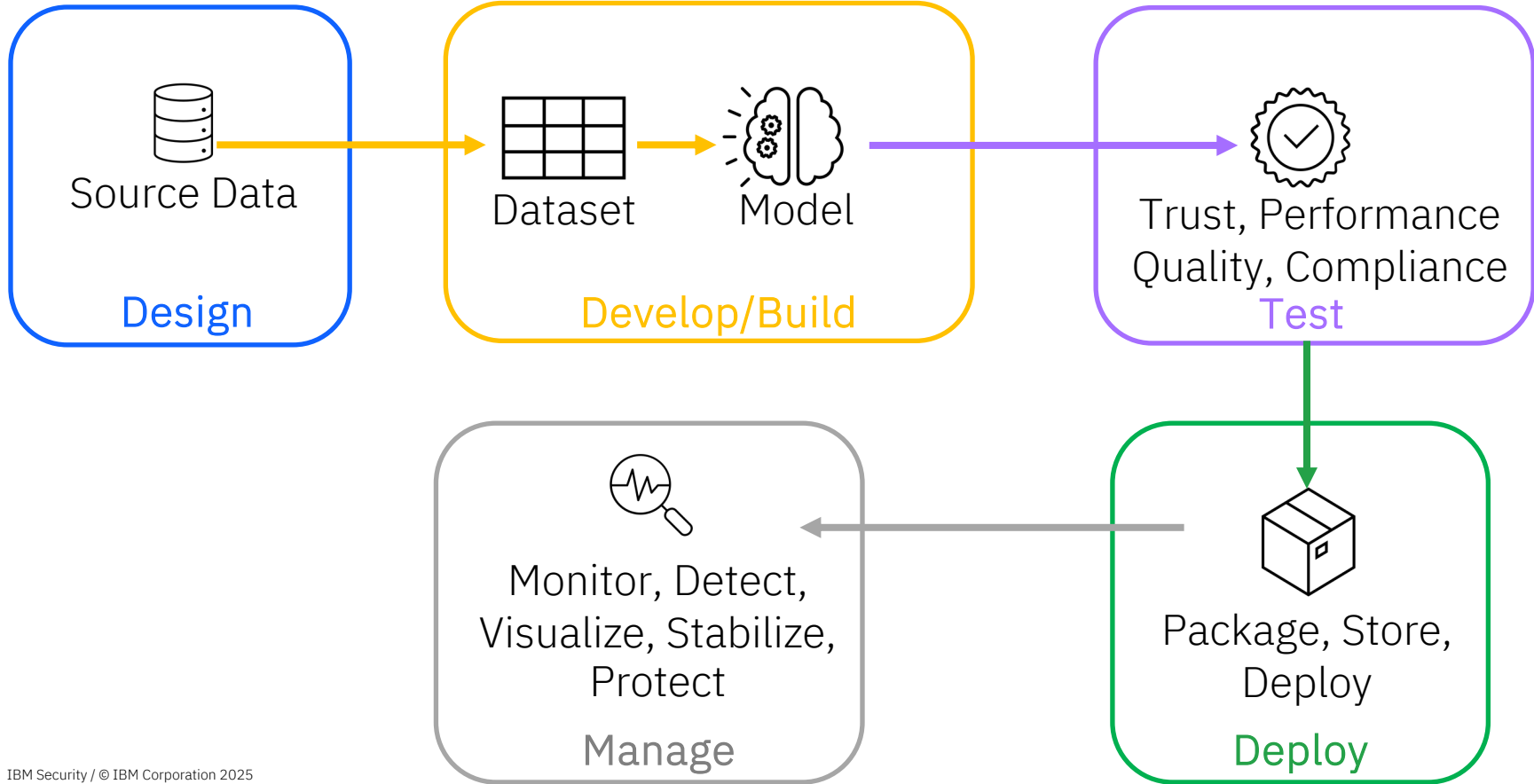
ibm.com/topics/mlops



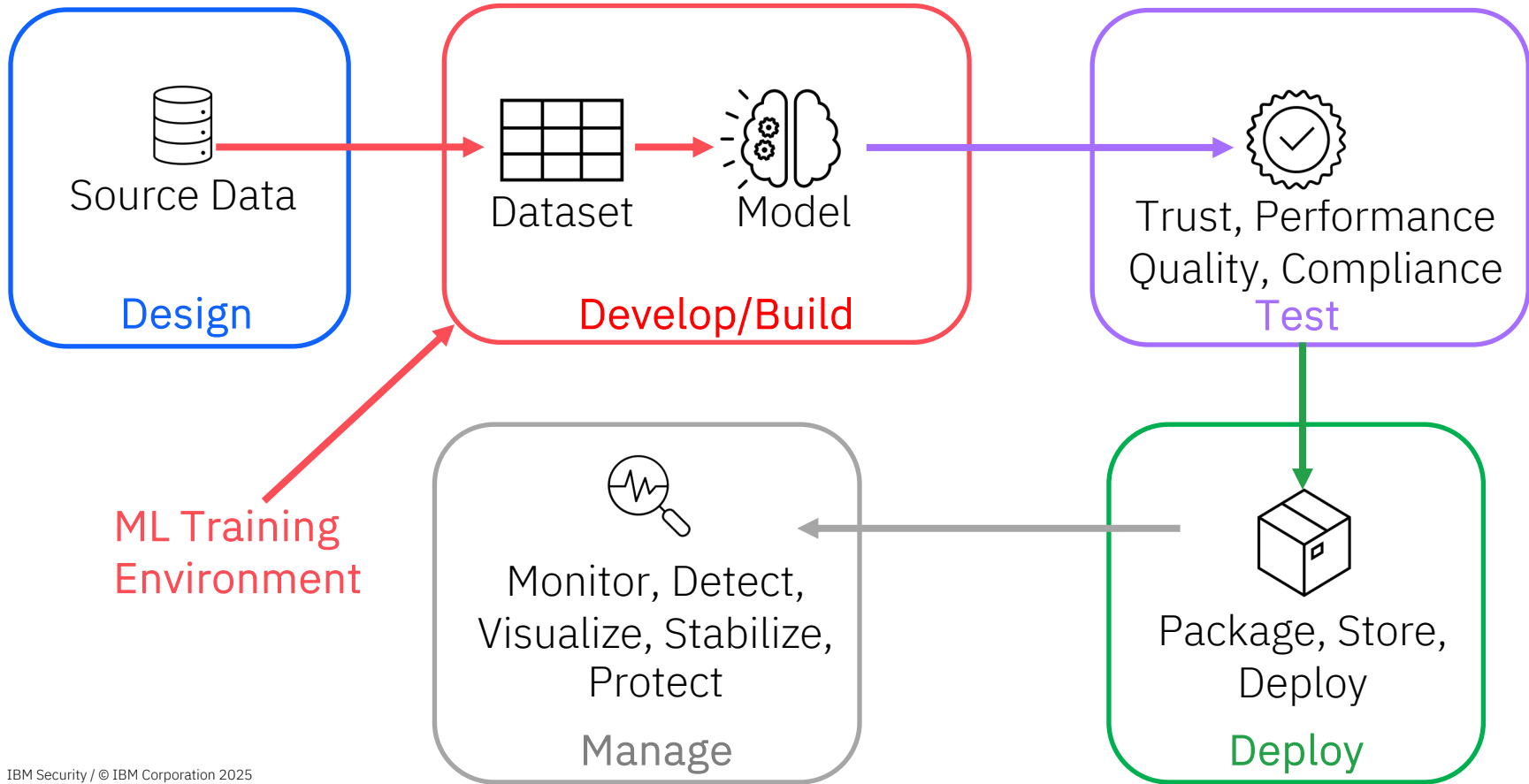
Popular MLOps Platforms



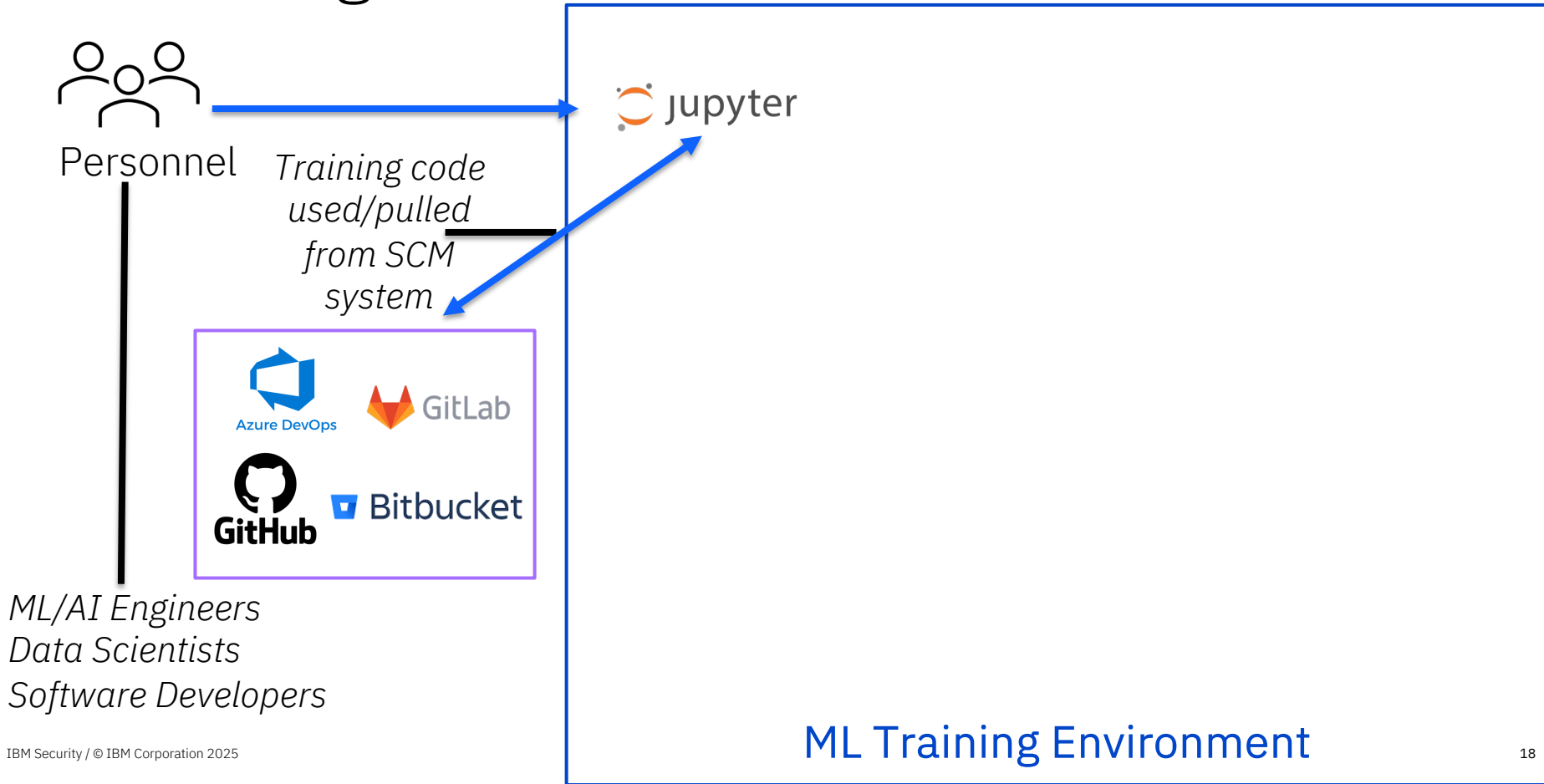
MLOps Lifecycle



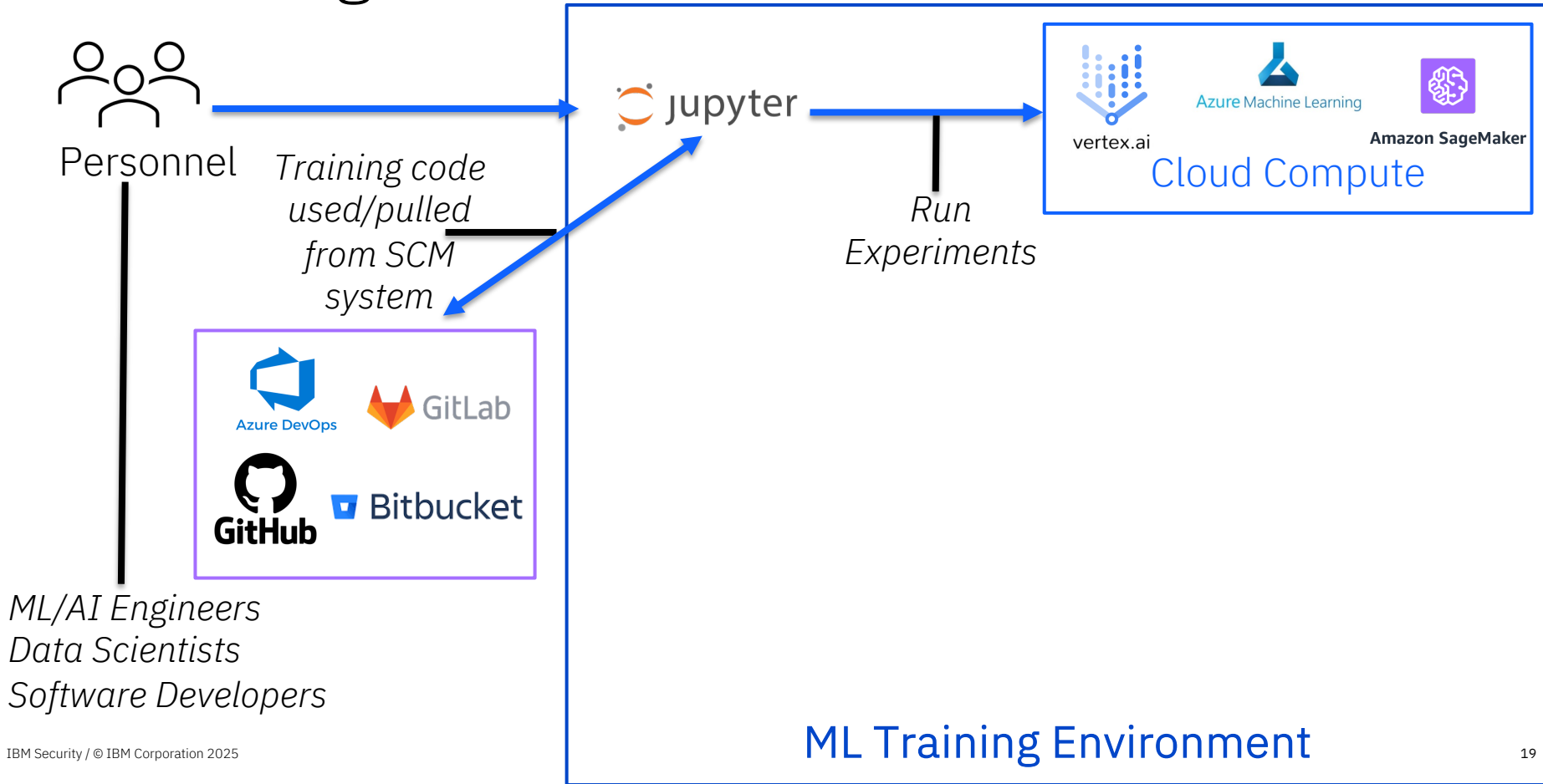
MLOps Lifecycle - ML Training Environment



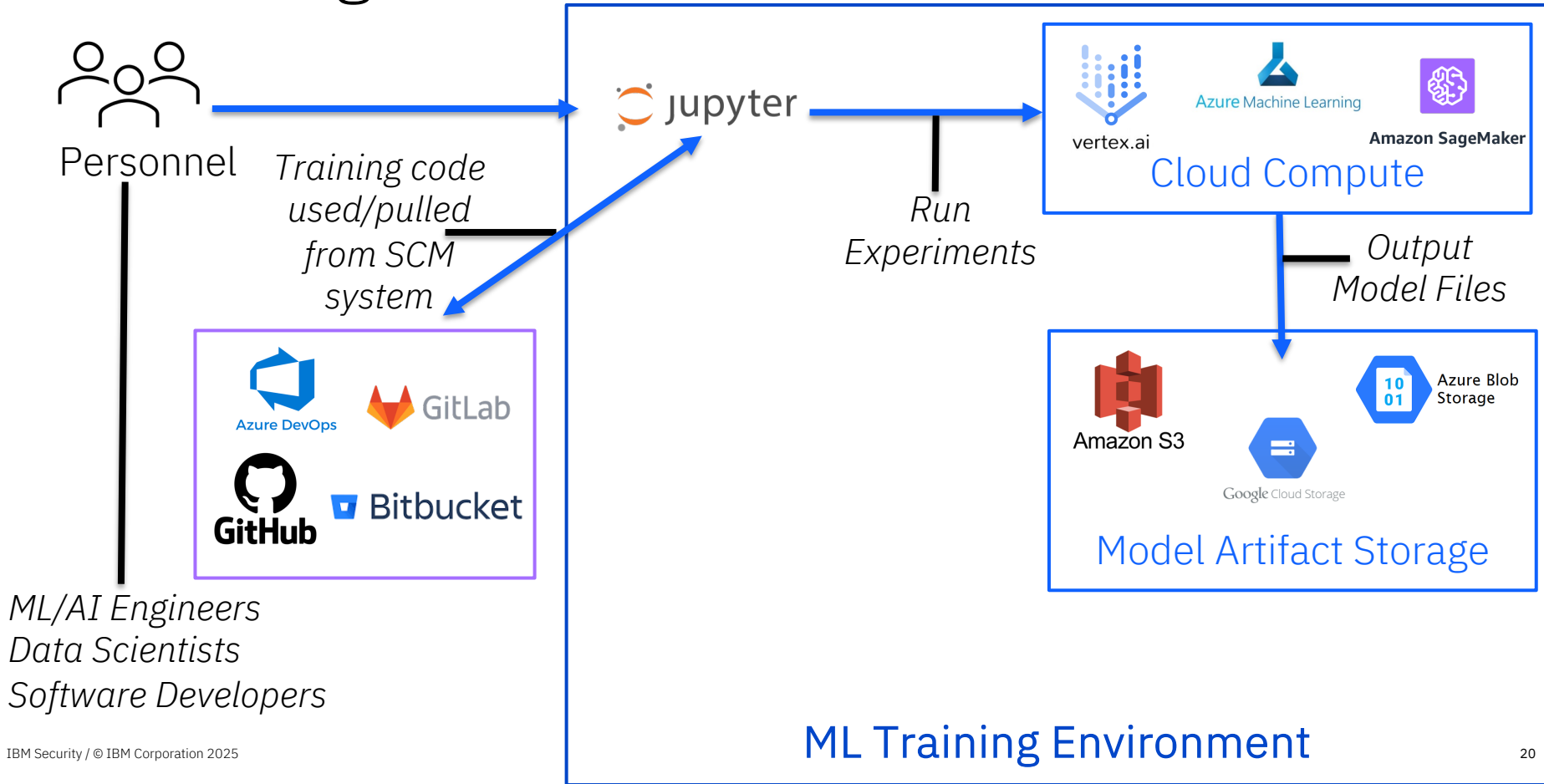
ML Training Environment Infrastructure



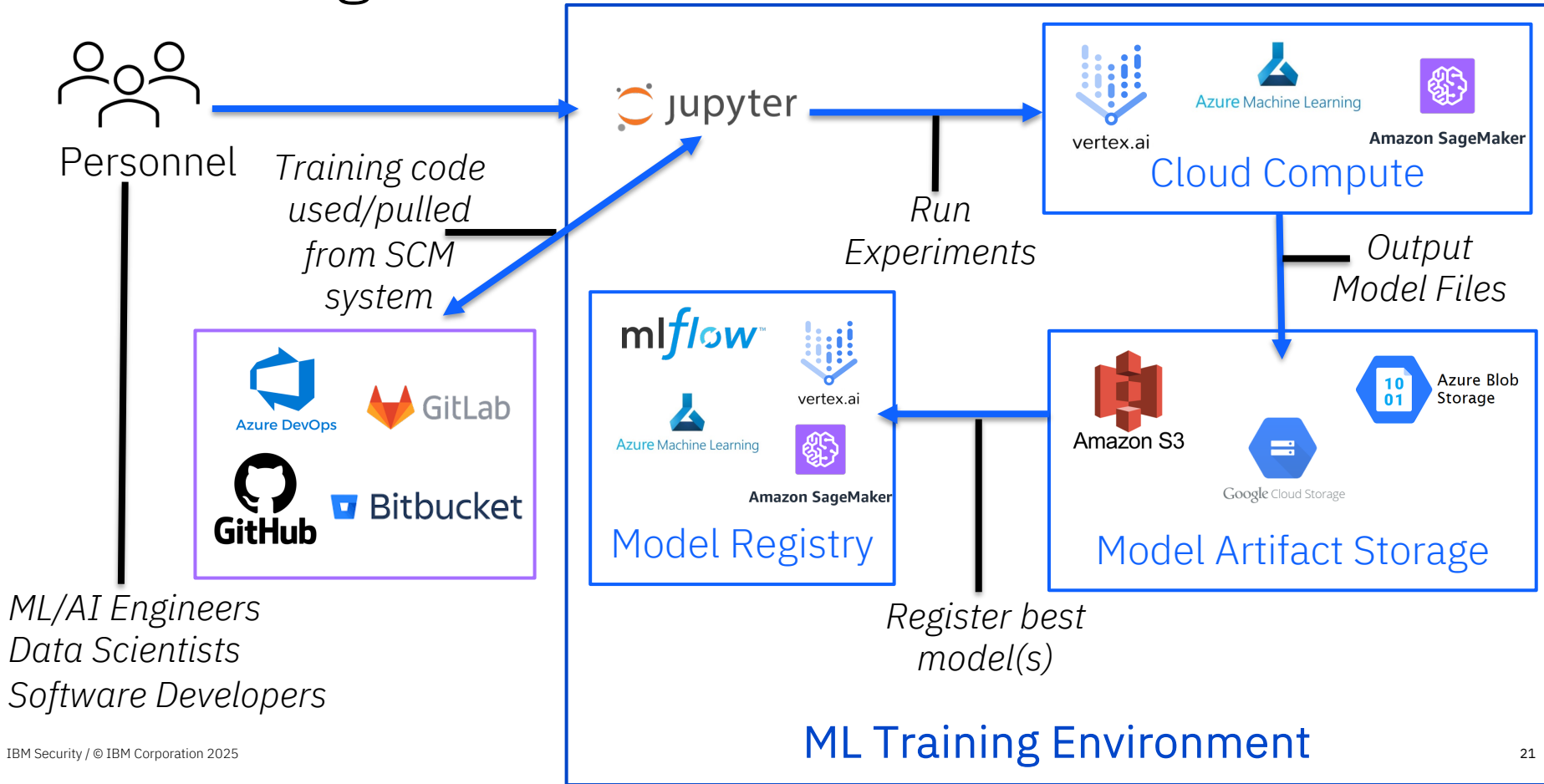
ML Training Environment Infrastructure



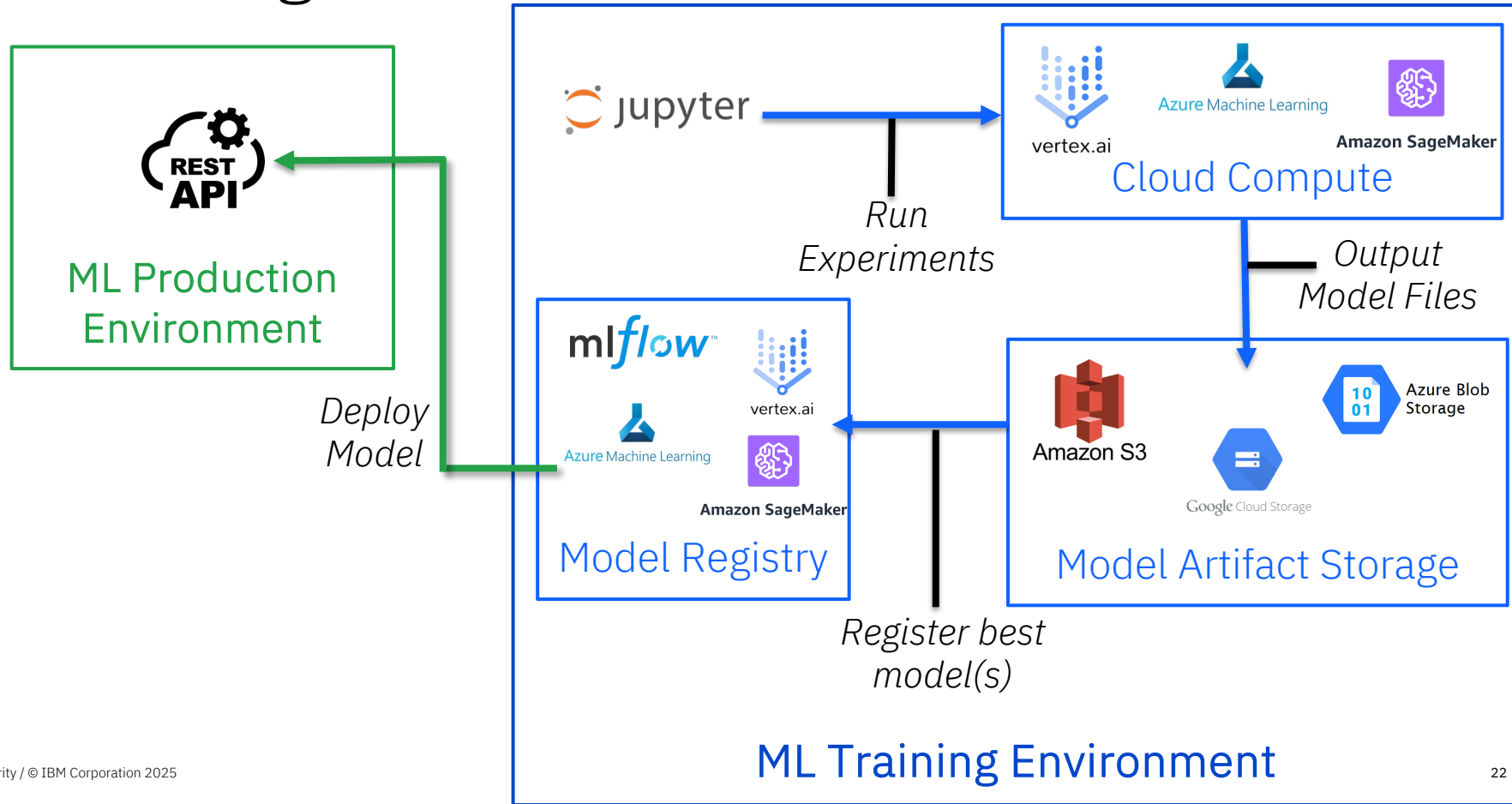
ML Training Environment Infrastructure



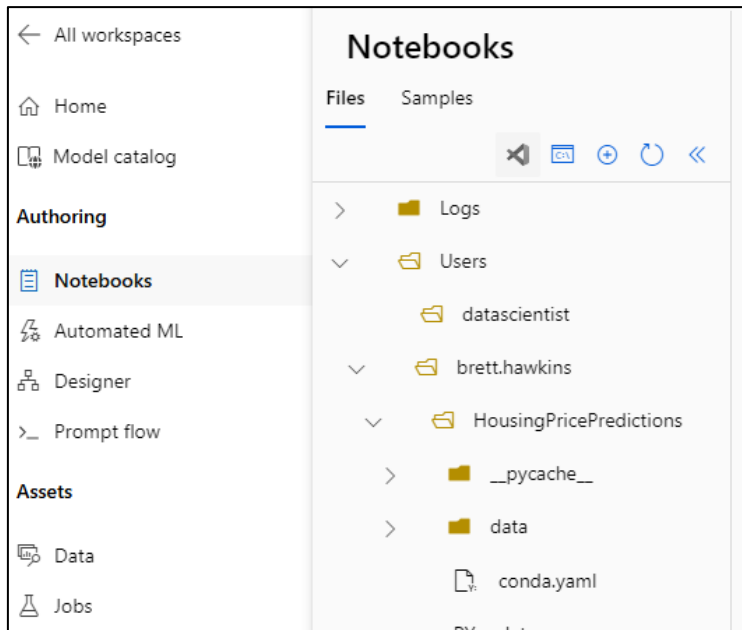
ML Training Environment Infrastructure



ML Training Environment Infrastructure



ML Training Infrastructure Components



Notebook Env
Contains ML training code



Model Registry
Track and version models



Cloud Compute
Infrastructure that performs ML training from ML training code



Model Artifact Storage
ML training artifact outputs (model weights, model files, etc.)

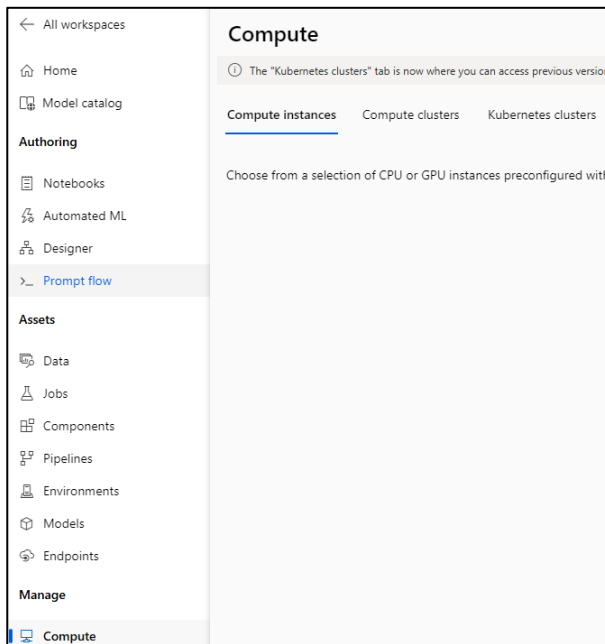
Notebook Env- SageMaker

The screenshot shows the SageMaker console interface. On the left is a navigation sidebar with sections like 'Notebooks', 'Admin configurations', and 'JumpStart'. The main area displays details for the notebook instance 'instance/brett-test-mlflow-notebook'. It includes lifecycle configuration, status (InService), creation time (Dec 04, 2024 19:43 UTC), last updated time (Dec 05, 2024 15:08 UTC), volume size (5GB EBS), platform identifier (Amazon Linux 2, Jupyter Lab 3), and minimum IMDS version (2). Below this, the 'Git repositories' section shows a table with one entry: 'brett-test-mlflow - (Default)' with the repository URL 'https://git-codecommit.us-east-2.amazonaws.com/v1/repos/brett-test-mlf'. A red arrow points from the 'Repository URL' column header to the file explorer on the right.

Name	Repository URL
brett-test-mlflow - (Default)	https://git-codecommit.us-east-2.amazonaws.com/v1/repos/brett-test-mlf

The screenshot shows a JupyterLab file explorer window. The top bar includes menus like 'File', 'Edit', 'View', 'Run', 'Kernel', 'Git', 'Tabs', 'Settings', and 'Help'. Below the menu bar is a search bar labeled 'Filter files by name'. The file list shows the directory structure: '/ brett-test-mlflow /'. The files listed are: 'data', 'mlruns', 'conda.yaml', 'data.py', 'deploy.py', 'MLproject' (highlighted in blue), 'params.py', 'run.py', 'test.py', 'train.py', and 'utils.py'. A red arrow points from the 'Repository URL' in the SageMaker console to the file explorer.

ML Training Infrastructure Components



Notebook Env
Contains ML
training code



Model Registry
Track and version
models



Cloud Compute
Infrastructure that
performs ML
training from ML
training code



Model Artifact
Storage
ML training
artifact outputs
(model weights,
model files, etc.)

ML Training Infrastructure Components



Notebook Env
Contains ML
training code



Model Registry
Track and version
models



Cloud Compute
Infrastructure that
performs ML
training from ML
training code



Model Artifact
Storage
ML training
artifact outputs
(model weights,
model files, etc.)

mlflow 2.18.0 Experiments Models

Experiments ⊕ ⌵

Search Experiments

- ✓ Default ✎ 🗑
- ✓ ElasticNet ✎ 🗑
- ✓ Some_Experiment_2 ✎ 🗑
- ✓ some-random-experiment ✎ 🗑
- ✓ Some_Experiment_3 ✎ 🗑

Displaying Runs from 5 Experiments

Runs Evaluation **Experimental** Traces **Experimental**

📄 📈 ⓘ Time created ▾

⬇️ Sort: Created ▾ 📊 Columns ▾ ☐ Expand rows 📁 Group by ▾

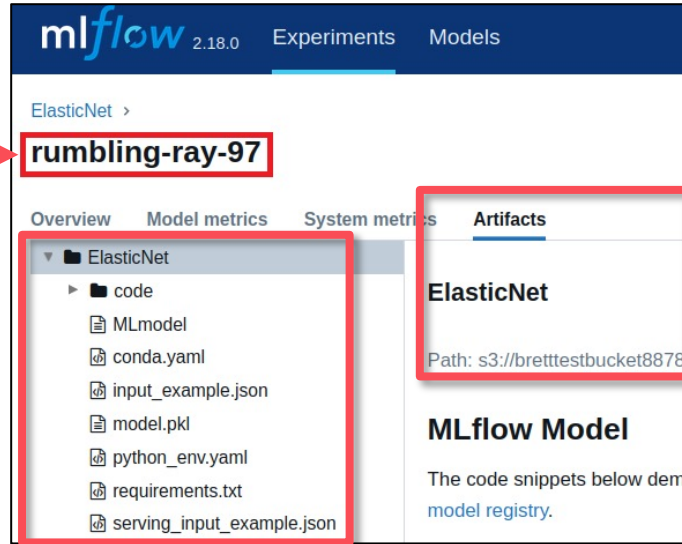
<input type="checkbox"/>	Run Name	Created ⬇️	Dataset
<input type="checkbox"/>	● silent-kite-450	🕒 10 seconds ago	📁 dataset (ea8b3b58) Training Datas...
<input type="checkbox"/>	● enthused-goat-111	🕒 22 seconds ago	📁 dataset (ea8b3b58) Training Datas...
<input type="checkbox"/>	● efficient-pig-620	🕒 34 seconds ago	📁 dataset (ea8b3b58) Training Datas...
<input type="checkbox"/>	● secretive-wolf-621	🕒 48 seconds ago	📁 dataset (ea8b3b58) Training Datas...
<input type="checkbox"/>	● bustling-steed-895	🕒 1 minute ago	📁 dataset (ea8b3b58) Training Datas...
<input type="checkbox"/>	● traveling-perch-716	🕒 1 minute ago	📁 dataset (ea8b3b58) Training Datas...
<input type="checkbox"/>	● unequaled-donkey-781	🕒 1 minute ago	📁 dataset (ea8b3b58) Training Datas...

Model Registry and Artifact Storage - MLFlow

Points to Experiment Run



Model Registry

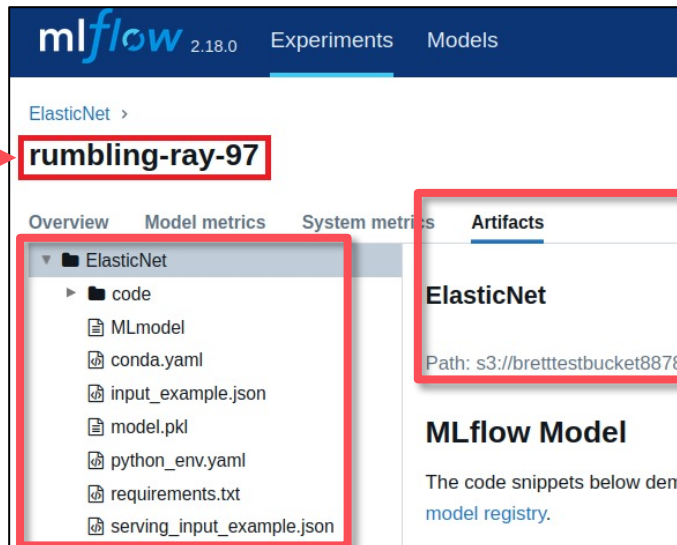


Model Registry and Artifact Storage - MLFlow

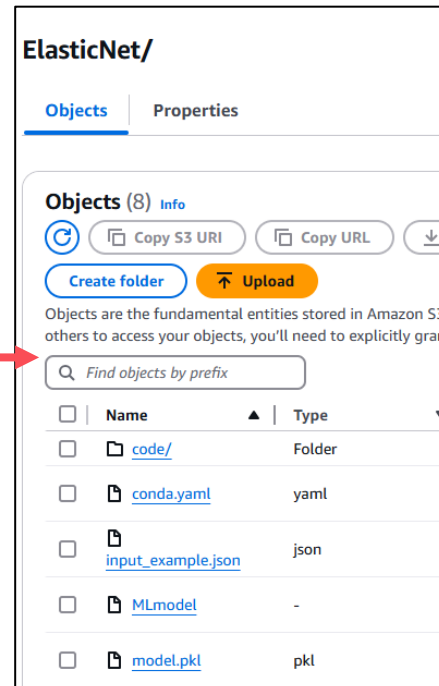
Points to Experiment Run



Model Registry



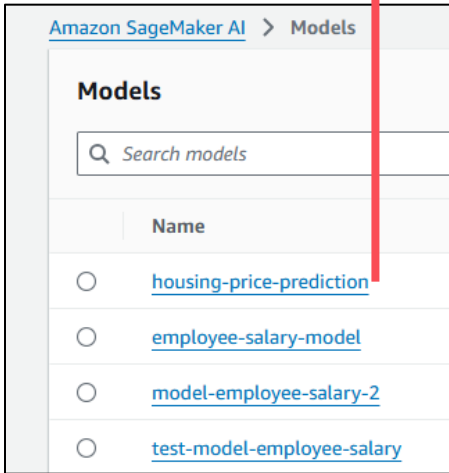
Model Artifacts from Experiment Run



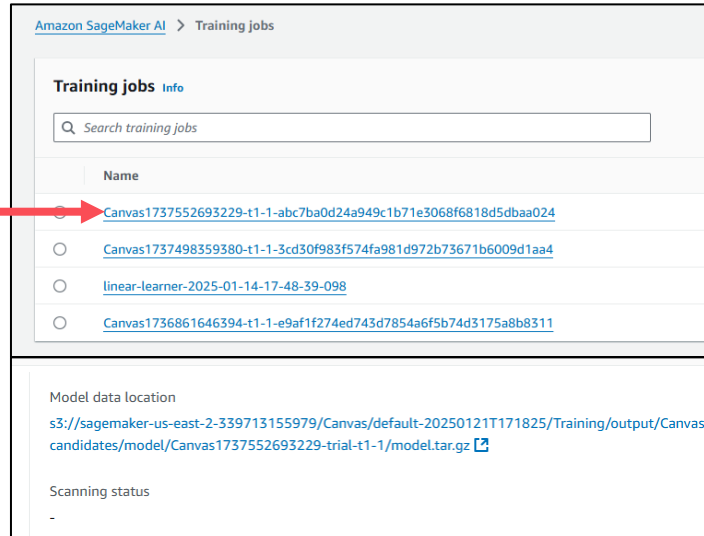
Artifact Storage (S3)

Model Registry and Artifact Storage - SageMaker

*Points to
Training Job*



Model Registry



Model Registry and Artifact Storage - SageMaker

*Points to
Training Job*

Amazon SageMaker AI > Training jobs

Training jobs [Info](#)

	Name
<input checked="" type="radio"/>	Canvas1737552693229-t1-1-abc7ba0d24a949c1b71e3068f6818d5dbaa024
<input type="radio"/>	Canvas1737498359380-t1-1-3cd30f983f574fa981d972b73671b6009d1aa4
<input type="radio"/>	linear-learner-2025-01-14-17-48-39-098
<input type="radio"/>	Canvas1736861646394-t1-1-e9af1f274ed743d7854a6f5b74d3175a8b8311

Model data location

<s3://sagemaker-us-east-2-339713155979/Canvas/default-20250121T171825/Training/output/Canvas/candidates/model/Canvas1737552693229-trial-t1-1/model.tar.gz>

Scanning status

-

Artifact
Storage (S3)

Canvas1737552693229-trial-t1-1/

[Objects](#) [Properties](#)

Objects (1)

Objects are the fundamental entities stored in Amazon S3. You can

<input type="checkbox"/>	Name	Type
<input checked="" type="checkbox"/>	model.tar.gz	gz

Model Registry

*Model Artifacts from
Training Job Run*

Model Registry and Artifact Storage – Azure ML

*Points to
Training Job*

Model List				
+ Register ▾ Refresh Delete Archive Deploy Compare (preview) ▾				
Search				
Name	☆	Version	Type	Job (Run ID)
salary-model		1	MLFLOW	helpful_leg_n2slbwmckq_40
heart-failure-model		1	MLFLOW	kind_guava_06ftffx2s2_38
taxifare-output-model		1	MLFLOW	AutoML_91114fd1-6657-4bf0

Model Registry

helpful_leg_n2slbwmckq Completed

Overview Data guardrails Models + child jobs Outputs + logs

Refresh Edit and submit (preview) + Register model

outputs

_automl_internal

featurization

data

pipeline

conda.yaml

MLmodel

model.pkl

python_env.yaml

requirements.txt

Model Registry and Artifact Storage – Azure ML

*Points to
Training Job*

Model List				
+ Register ▾ Refresh Delete Archive Deploy Compare (preview) ▾				
Search				
Name	☆	Version	Type	Job (Run ID)
salary-model		1	MLFLOW	helpful_leg_n2slbwmckq_40
heart-failure-model		1	MLFLOW	kind_guava_06ftfx2s2_38
taxifare-output-model		1	MLFLOW	AutoML_91114fd1-6657-4bf0

Model Registry

helpful_leg_n2slbwmckq Completed

Overview Data guardrails Models + child jobs **Outputs + logs**

Refresh Edit and submit (preview) + Register model

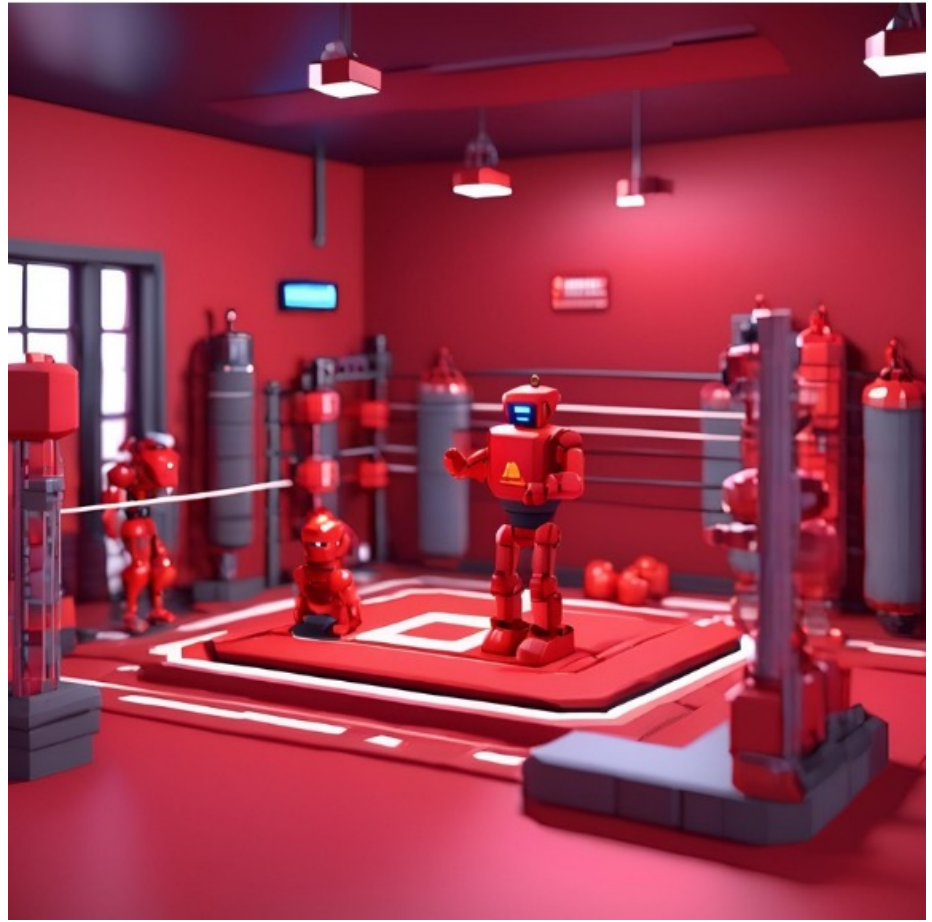
- outputs
 - _automl_internal
- featurization
 - data
- pipeline
 - conda.yaml
 - MLmodel
 - model.pkl
 - python_env.yaml
 - requirements.txt



Azure Blob Storage

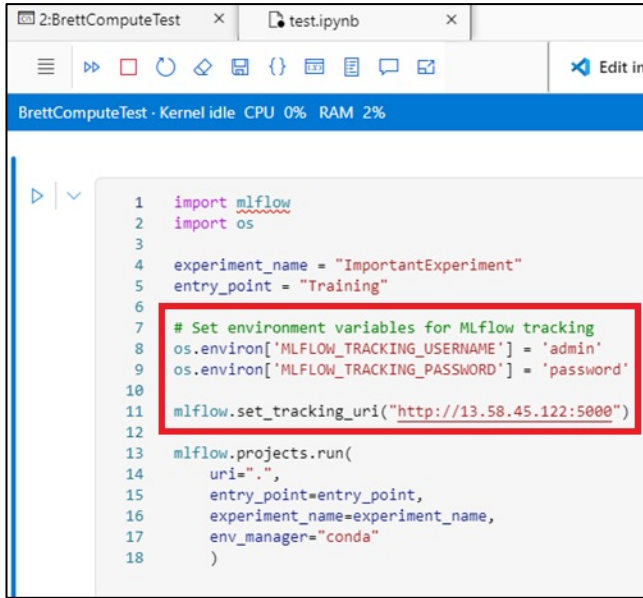
*Model Artifacts
from Training
Job Run*

Attacking ML Training Environments



Key Components – Attacker Perspective

*Credentials and
info on other
infrastructure*



```
1 import mlflow
2 import os
3
4 experiment_name = "ImportantExperiment"
5 entry_point = "Training"
6
7 # Set environment variables for MLflow tracking
8 os.environ['MLFLOW_TRACKING_USERNAME'] = 'admin'
9 os.environ['MLFLOW_TRACKING_PASSWORD'] = 'password'
10 mlflow.set_tracking_uri("http://13.58.45.122:5000")
11
12
13 mlflow.projects.run(
14     uri=".",
15     entry_point=entry_point,
16     experiment_name=experiment_name,
17     env_manager="conda"
18 )
```

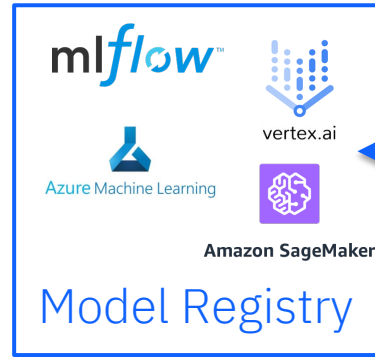


Jupyter

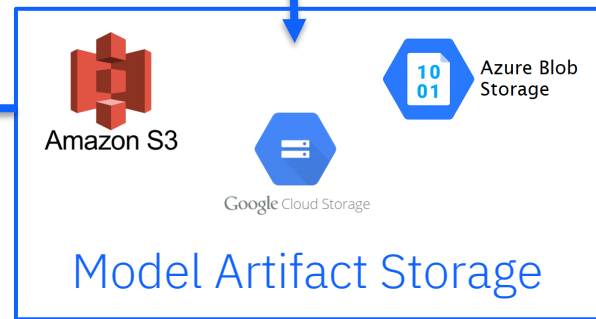
Run
Experiments



Output
Model Files

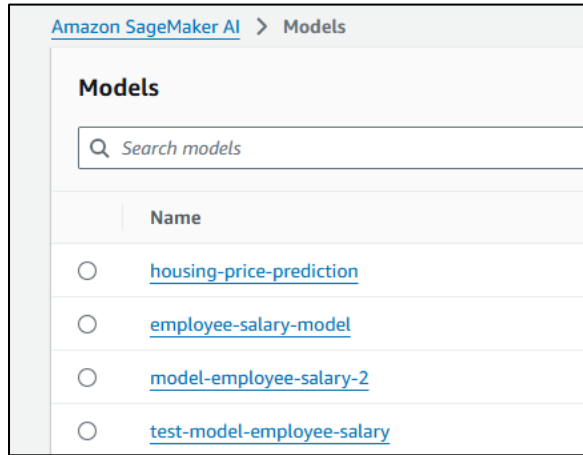


Register best
model(s)

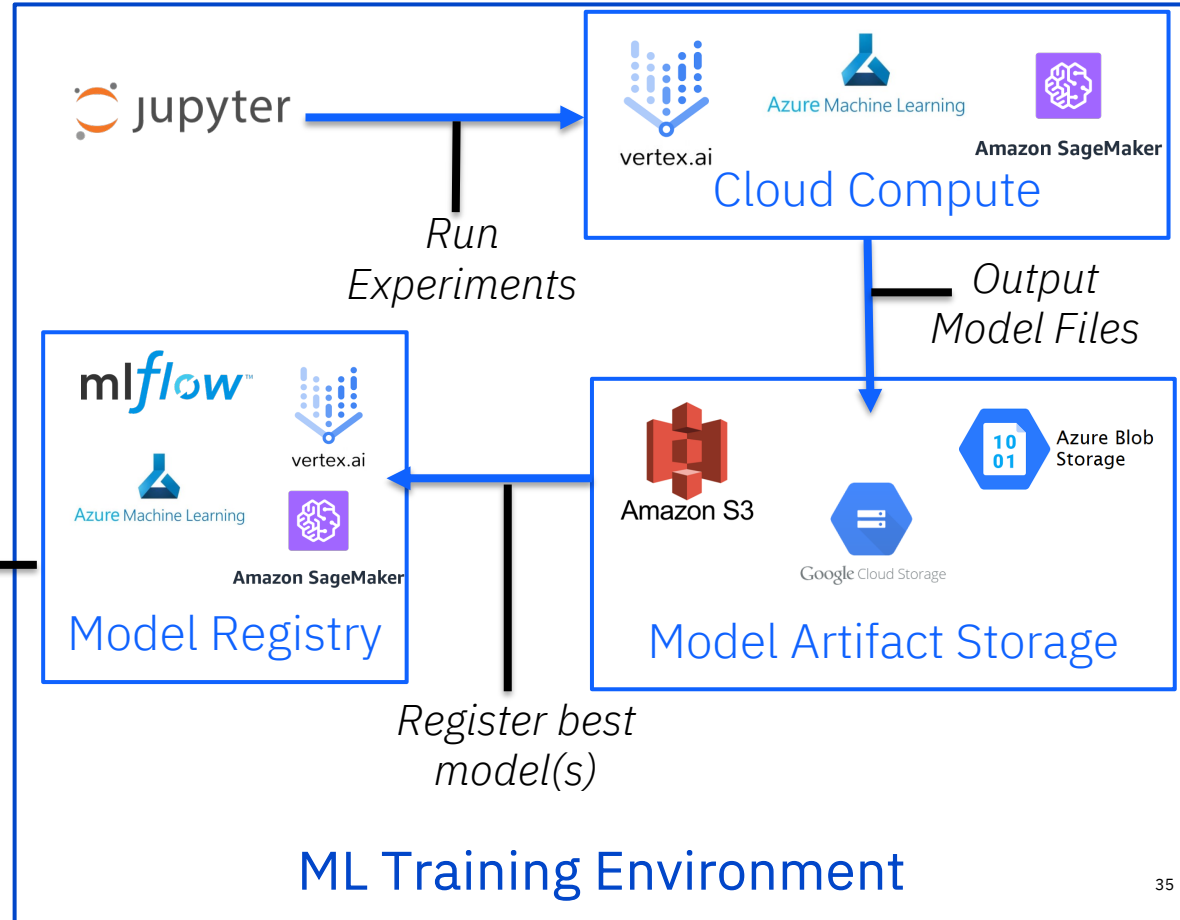


ML Training Environment

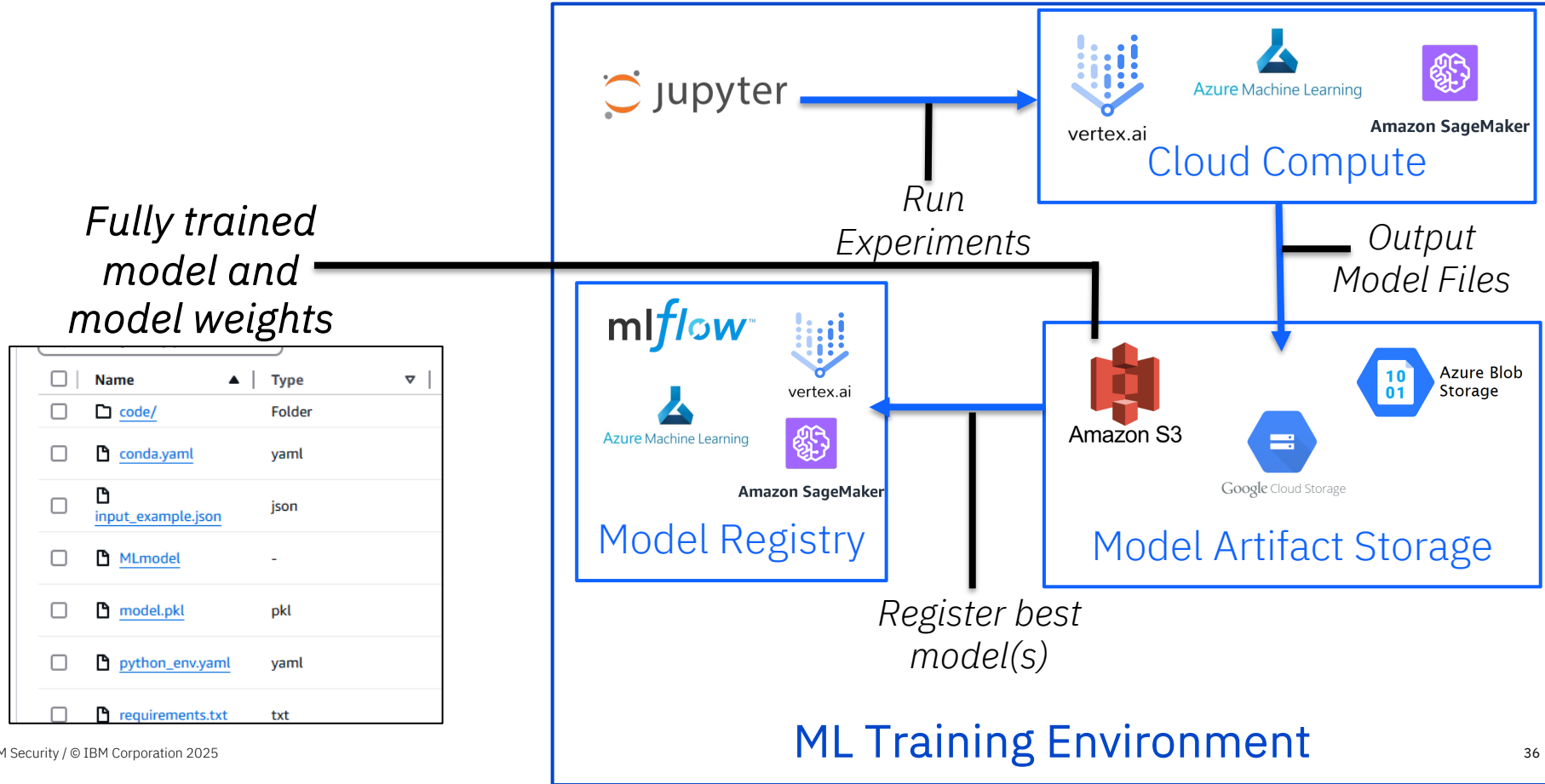
Key Components – Attacker Perspective



Useful for model reconnaissance



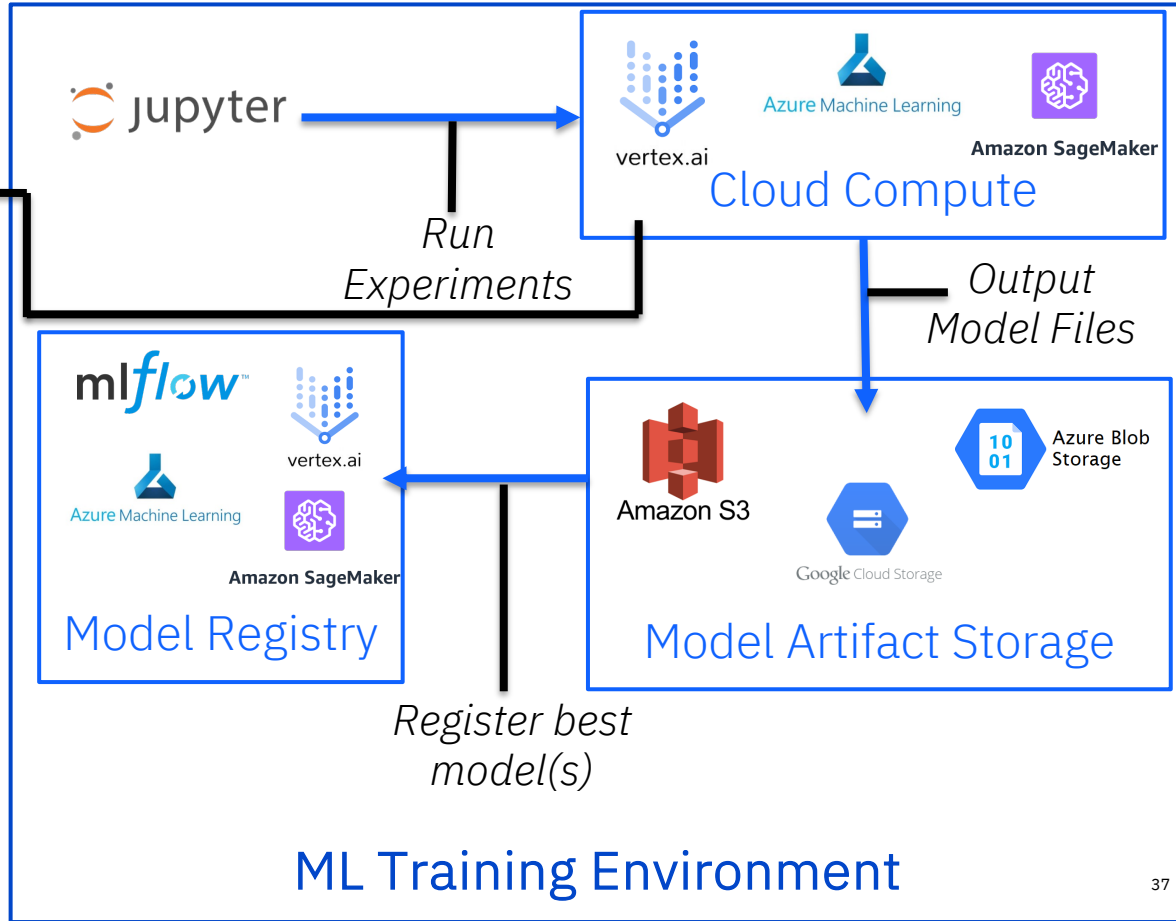
Key Components – Attacker Perspective



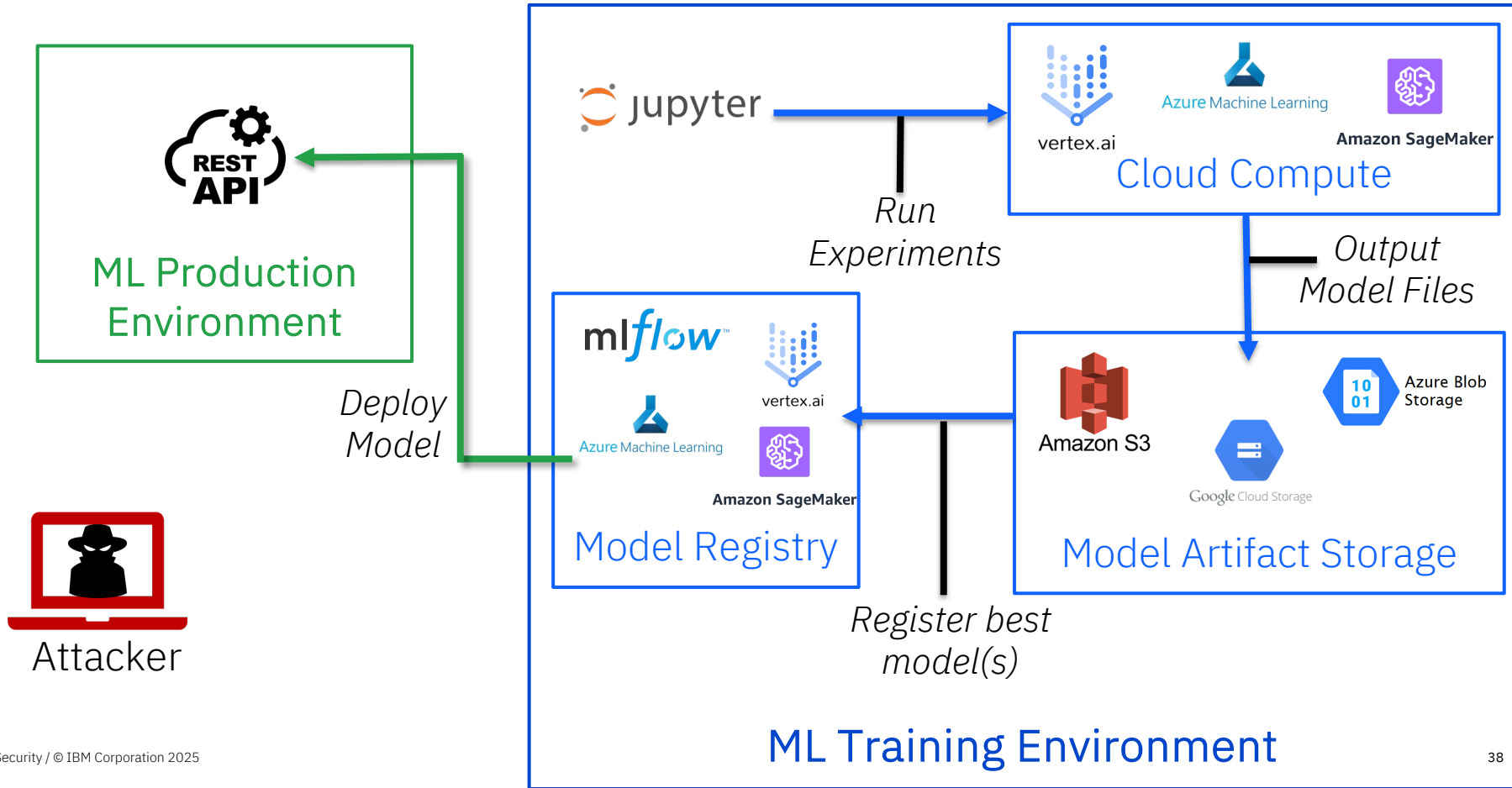
Key Components – Attacker Perspective

Sensitive environment variables and ML training code

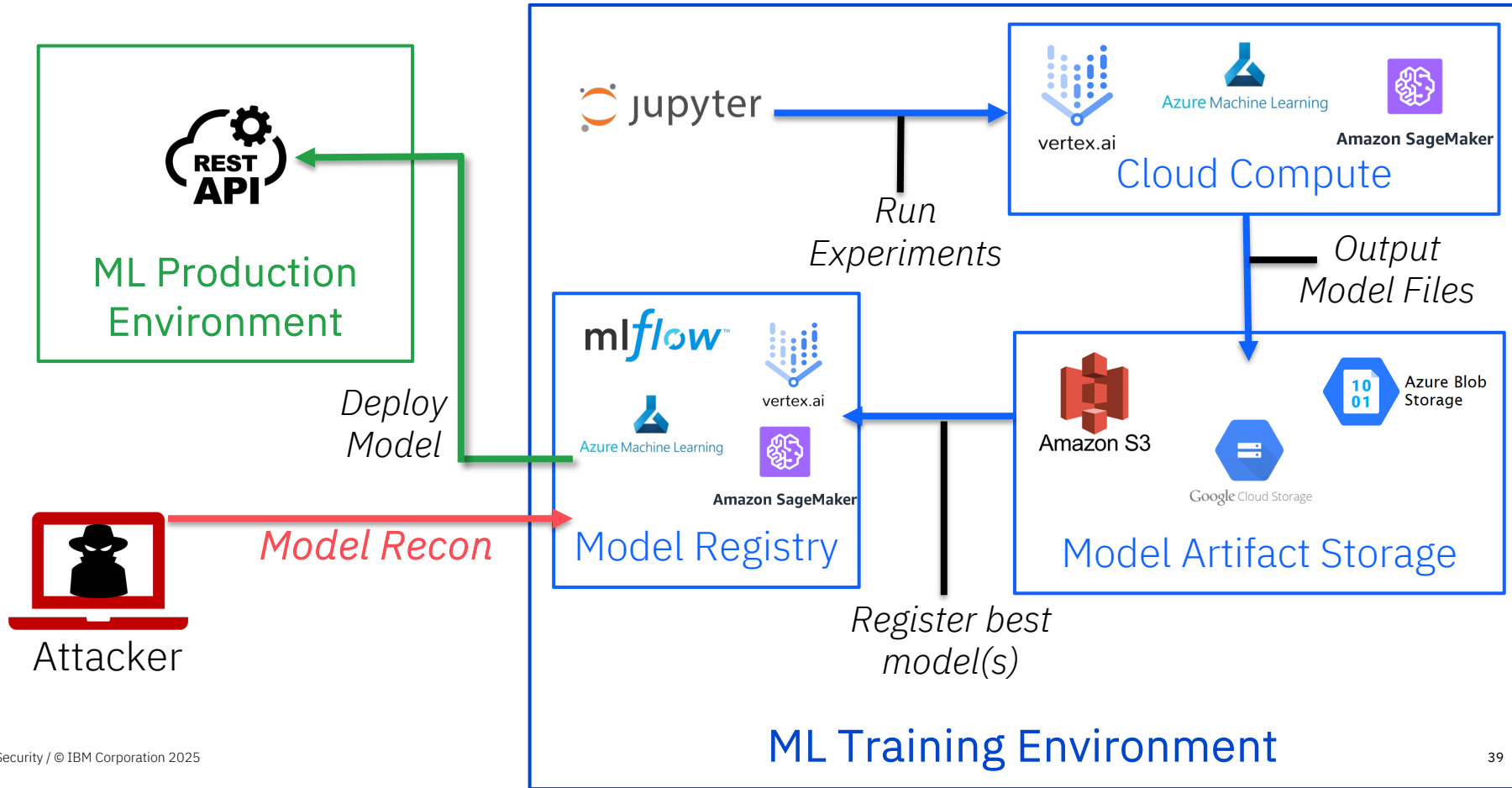
```
(base) [ec2-user@ip-172-16-11-6 brett-test-mflow]$ ls -la
ls -la
total 64
drwxrwxr-x 7 ec2-user ec2-user 4096 Dec 5 15:35 .
drwxr-xr-x 6 ec2-user ec2-user 4096 Dec 5 15:08 ..
-rw-rw-r-- 1 ec2-user ec2-user 1292 Dec 4 19:48 conda.yaml
drwxrwxr-x 2 ec2-user ec2-user 4096 Dec 4 19:48 data
-rw-rw-r-- 1 ec2-user ec2-user 1485 Dec 4 19:48 data.py
-rw-rw-r-- 1 ec2-user ec2-user 879 Dec 4 19:48 deploy.py
drwxrwxr-x 8 ec2-user ec2-user 4096 Dec 5 15:35 .git
drwxrwxr-x 2 ec2-user ec2-user 4096 Dec 5 14:30 .ipynb_checkpoints
-rw-rw-r-- 1 ec2-user ec2-user 131 Dec 5 15:35 MLproject
drwxrwxr-x 4 ec2-user ec2-user 4096 Dec 5 15:15 mlruns
-rw-rw-r-- 1 ec2-user ec2-user 986 Dec 4 19:48 params.py
drwxrwxr-x 2 ec2-user ec2-user 4096 Dec 4 19:58 __pycache__
-rw-rw-r-- 1 ec2-user ec2-user 270 Dec 5 14:30 run.py
-rw-rw-r-- 1 ec2-user ec2-user 520 Dec 4 19:48 test.py
-rw-rw-r-- 1 ec2-user ec2-user 1342 Dec 4 19:48 train.py
-rw-rw-r-- 1 ec2-user ec2-user 409 Dec 4 19:48 utils.py
(base) [ec2-user@ip-172-16-11-6 brett-test-mflow]$ ls -la data
ls -la data
total 904
drwxrwxr-x 2 ec2-user ec2-user 4096 Dec 4 19:48 .
drwxrwxr-x 7 ec2-user ec2-user 4096 Dec 5 15:35 ..
-rw-rw-r-- 1 ec2-user ec2-user 452865 Dec 4 19:48 test.csv
-rw-rw-r-- 1 ec2-user ec2-user 462137 Dec 4 19:48 train.csv
(base) [ec2-user@ip-172-16-11-6 brett-test-mflow]$
```



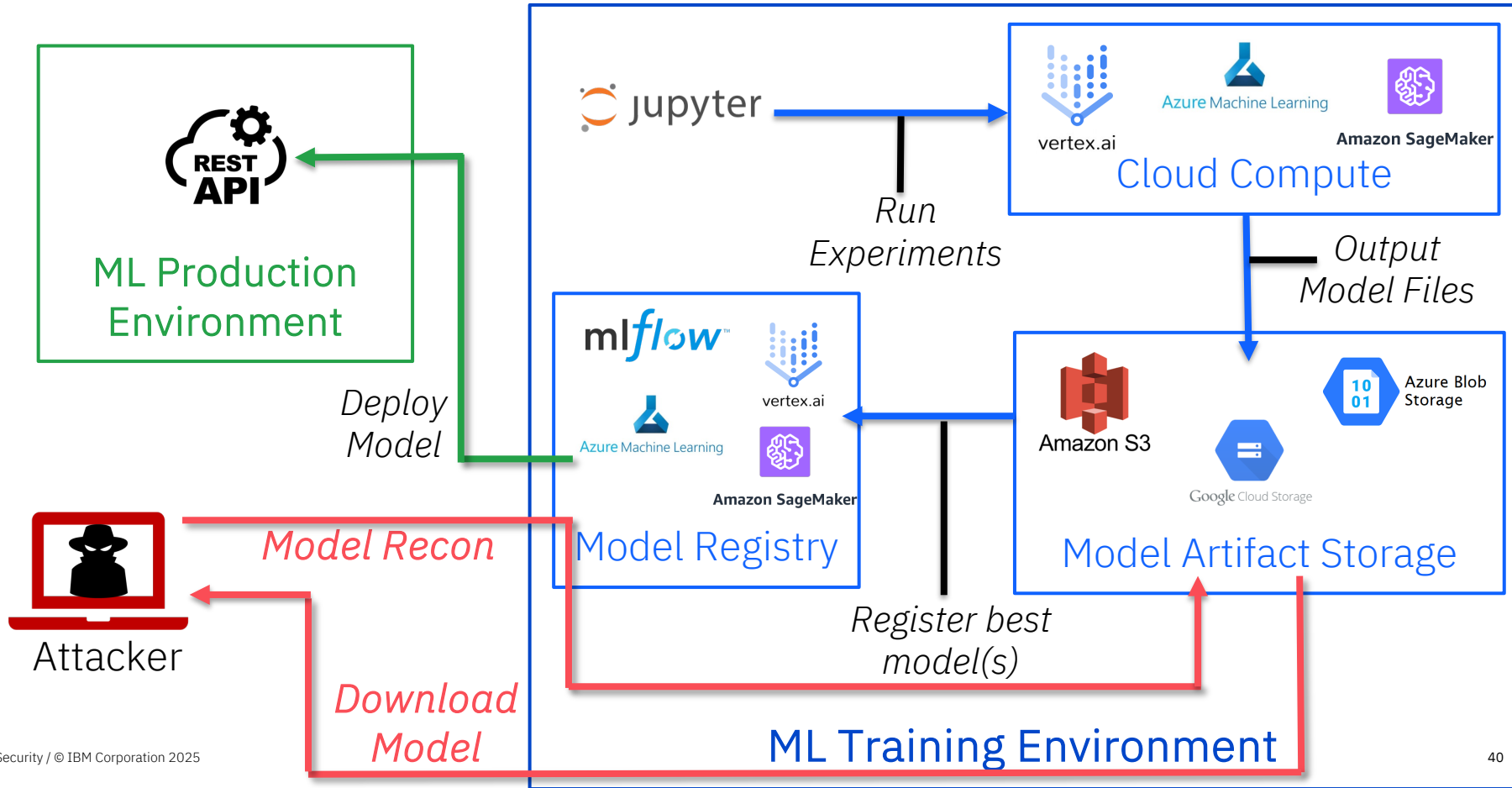
Model Theft



Model Theft



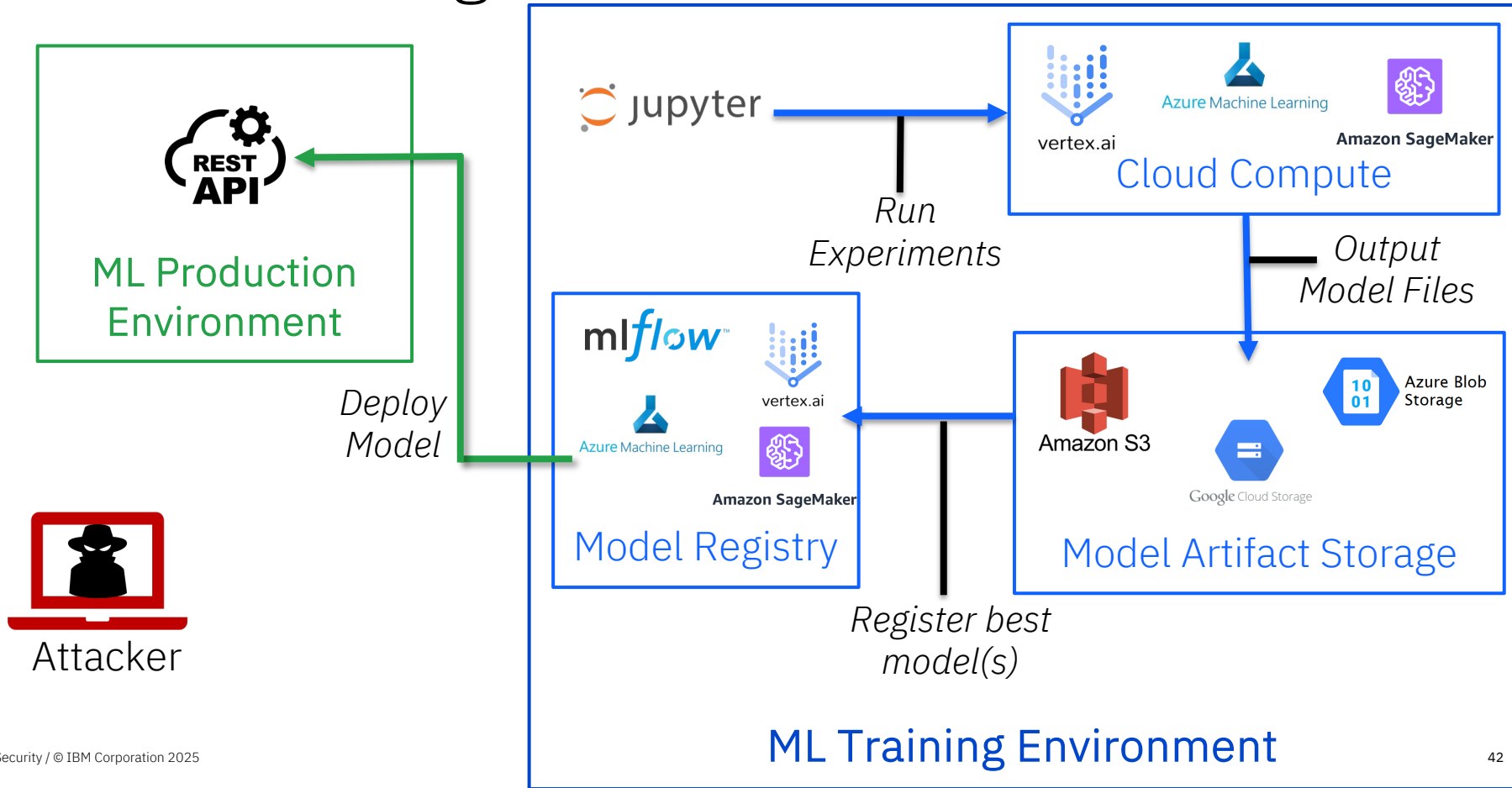
Model Theft



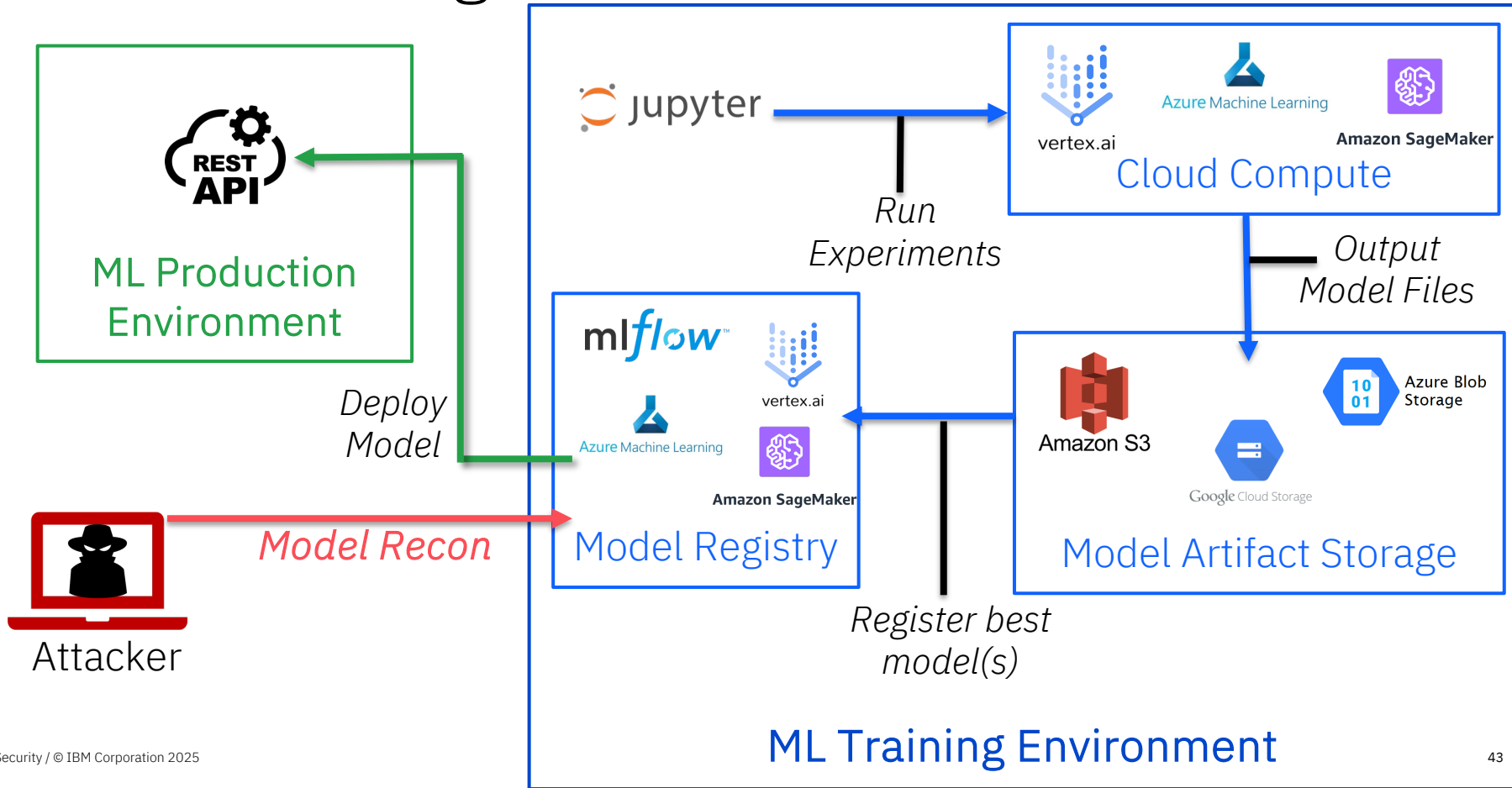
Model Theft - Impact



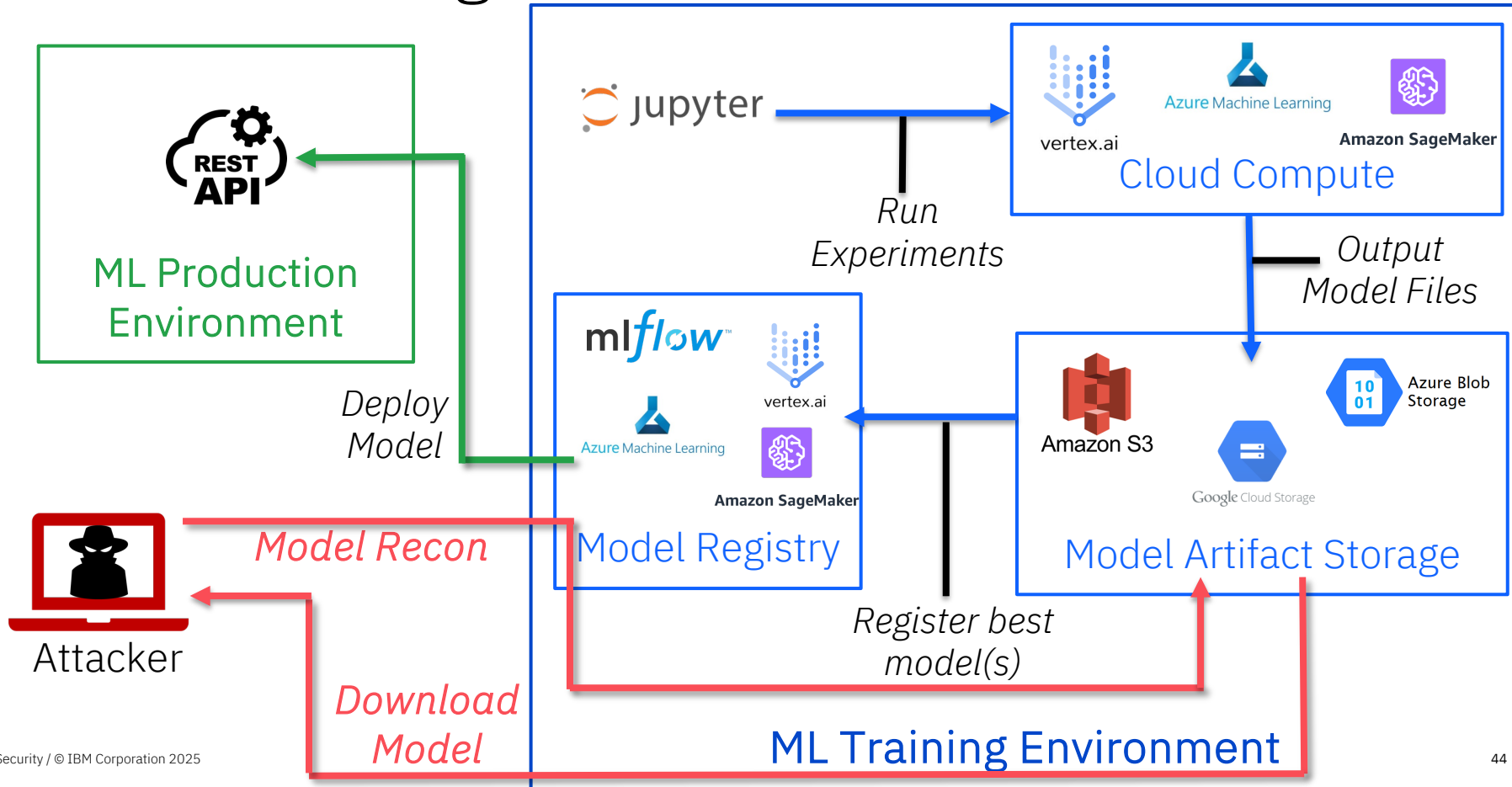
Model Poisoning – Code Execution



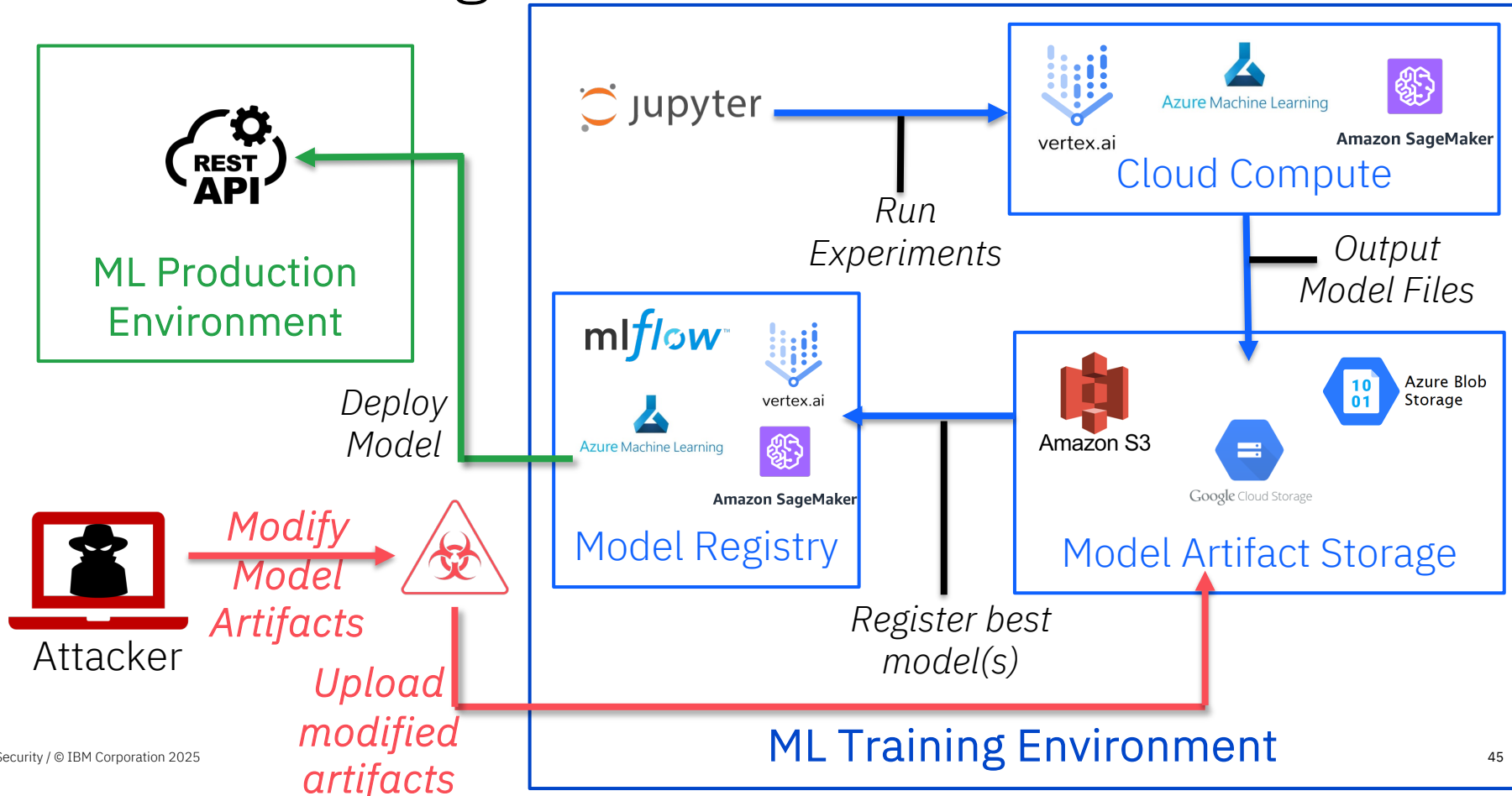
Model Poisoning – Code Execution



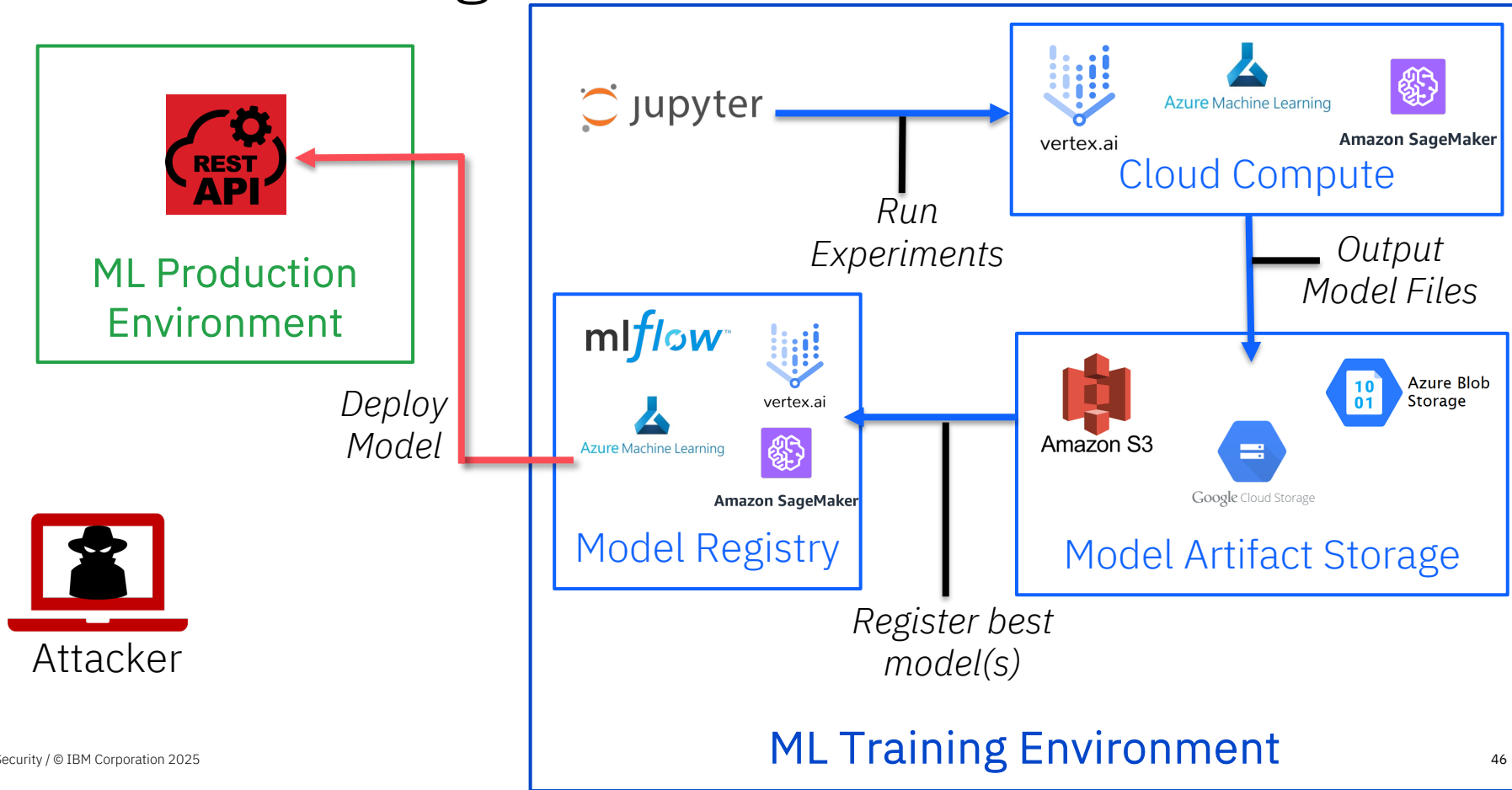
Model Poisoning – Code Execution



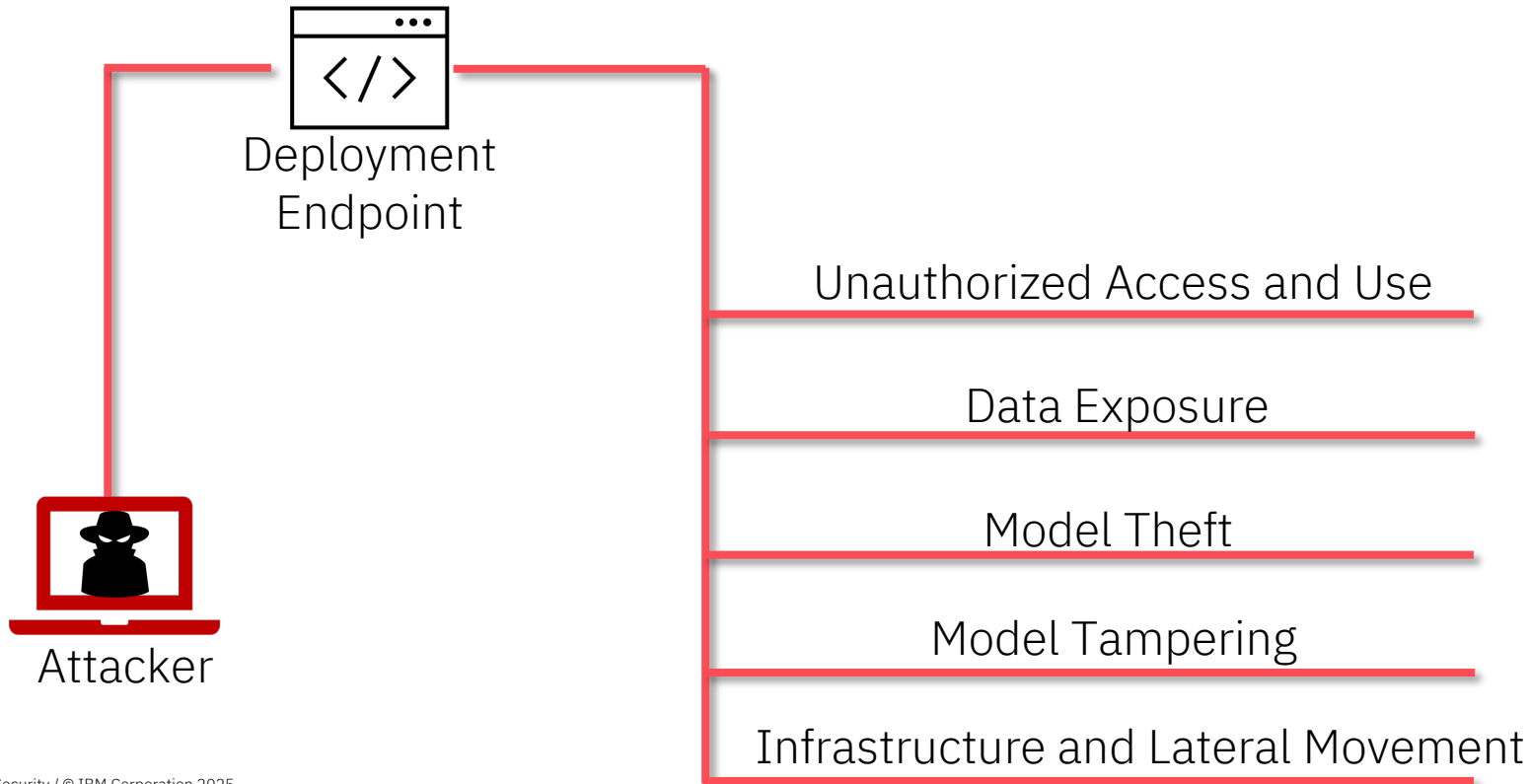
Model Poisoning – Code Execution



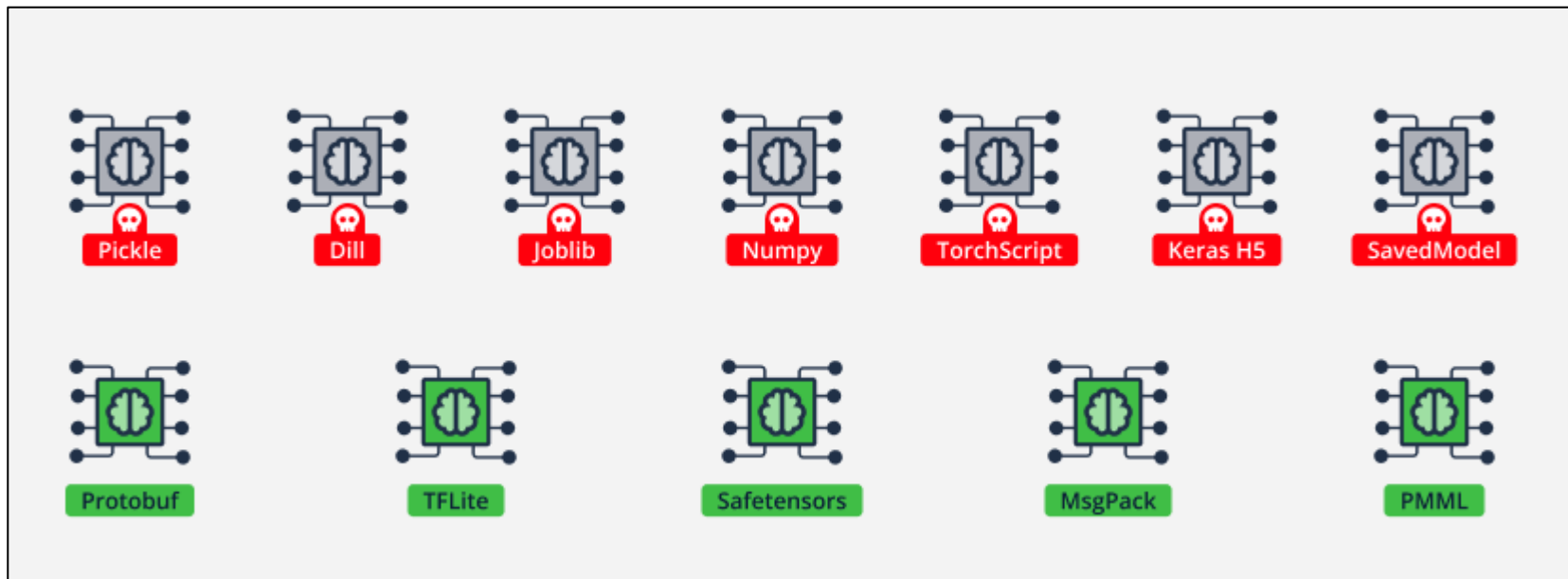
Model Poisoning – Code Execution



Model Poisoning – Code Execution - Impact



Model Formats – Support Code Execution on Load



<https://jfrog.com/blog/from-mlops-to-mloops-exposing-the-attack-surface-of-machine-learning-platforms/>

Creating Malicious Models for Code Execution

- MaliciousPickles - <https://github.com/coldwaterq/MaliciousPickles>
- Charcuterie - <https://github.com/moohax/Charcuterie>
- Fickling - <https://github.com/trailofbits/fickling>
- HiddenPickle - <https://github.com/hiddenlayerai/HiddenPickle>

MLOKit

github.com/xforced/MLOKit

```
[*] INFO: Performing download-model module for sagemaker
[*] INFO: Checking credentials provided
[+] SUCCESS: Credentials are valid

Model Name | Creation Date
-----|-----
employee-salary-model | 1/22/2025

[*] INFO: Downloading model artifacts
[*] INFO: Model artifacts location
s3://sagemaker-us-east-2-339713155979/Canvas/default-20250121T171825/Training/output/model.tar.gz
[*] INFO: Checking access to S3 bucket with name: sagemaker-us-east-2-339713155979
[+] SUCCESS: You have access to S3 bucket with name: sagemaker-us-east-2-339713155979
[*] INFO: Listing all files in prefix of: Canvas/default-20250121T171825/Training/output/
Canvas/default-20250121T171825/Training/output/Canvas1737498359380/sagemaker-model.tar.gz
[*] INFO: Downloading file at: Canvas/default-20250121T171825/Training/output/Canvas1737498359380/sagemaker-model.tar.gz
[+] SUCCESS: model.tar.gz written to: C:\Demo\MLOKit-OiEJQGbz
```



REST API Abuse
Conduct actions
programmatically



Authentication
API Key, Access
Token, Security
Creds, User/Pass

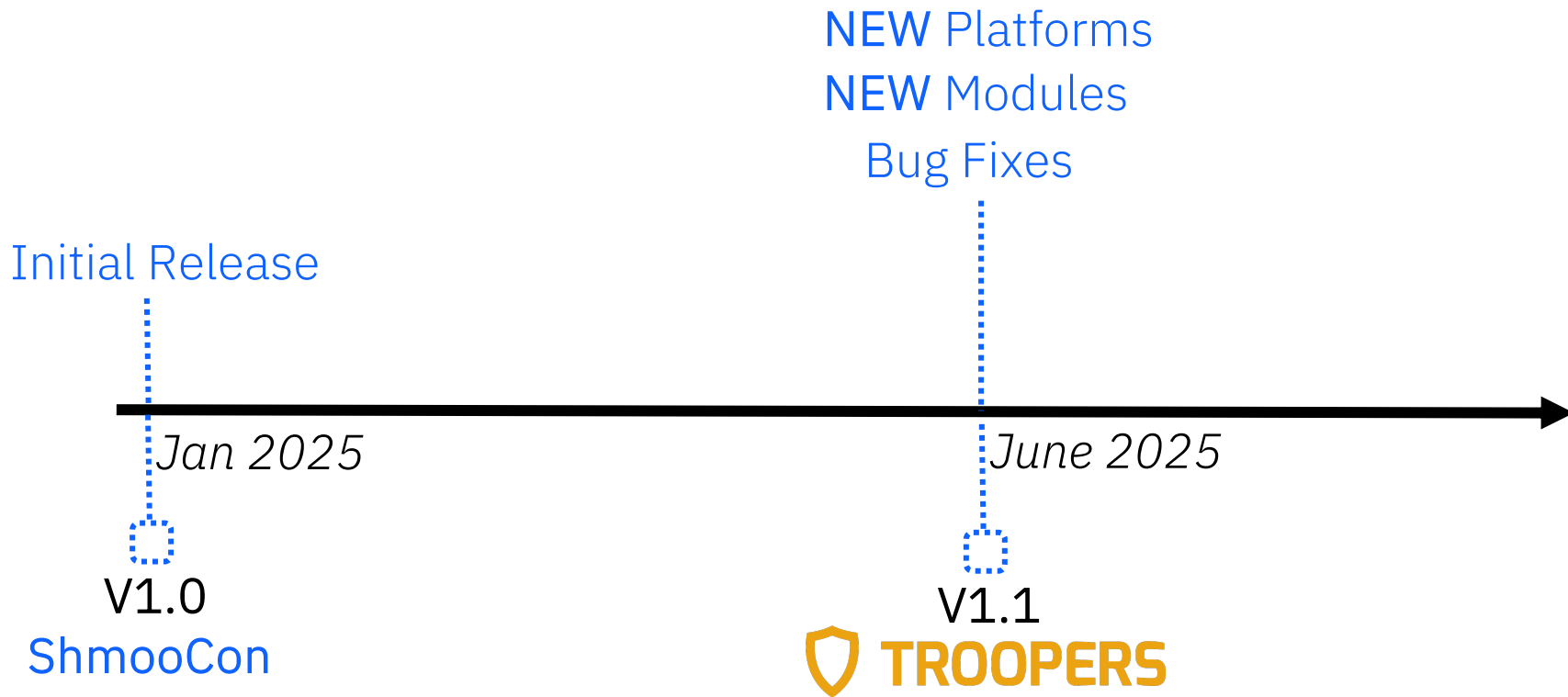


9 Modules
Recon, Training Data
Theft, Model Theft,
Model Poisoning,
Notebook Attacks



5 Supported Platforms
Azure ML, BigML,
Vertex AI, MLFlow,
SageMaker

MLOKit - History



Demos: Attack Scenarios



Obtaining Credentials

File Shares

Intranet Sites (e.g., internal wikis)

User Workstations

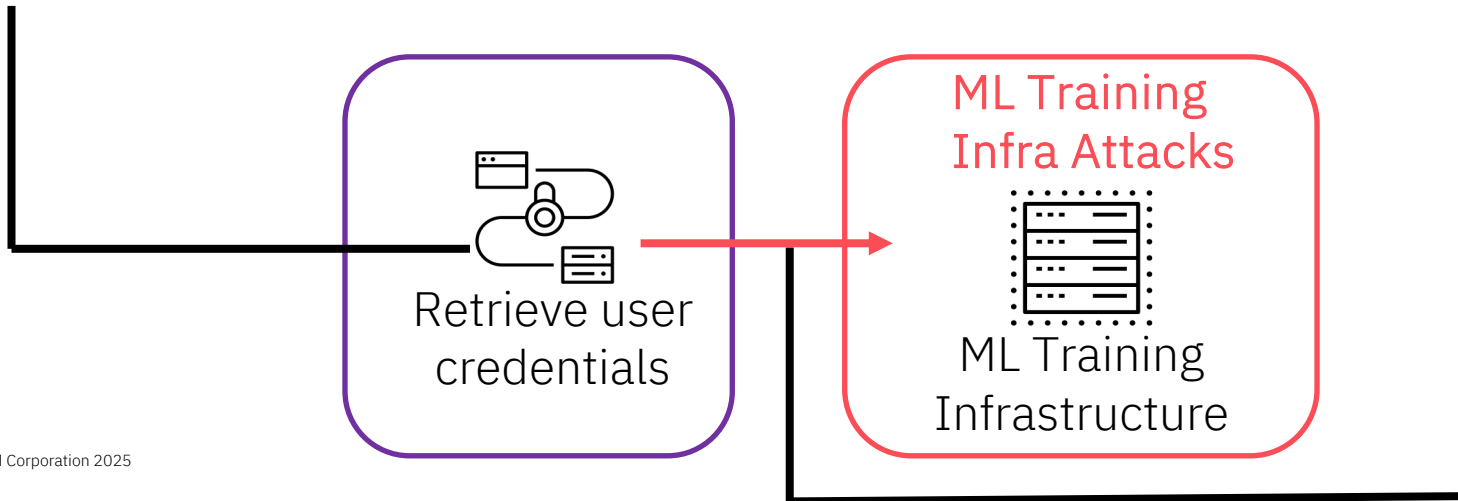
Social Engineering

Public Resources (e.g., Code Repos)

Unauthenticated Access

Public Data Breach Leaks

This research focuses
on attack paths
possible from here



Demos: Attack Scenarios

#1: MLFlow - Initial Access and Model Theft from Model Registry

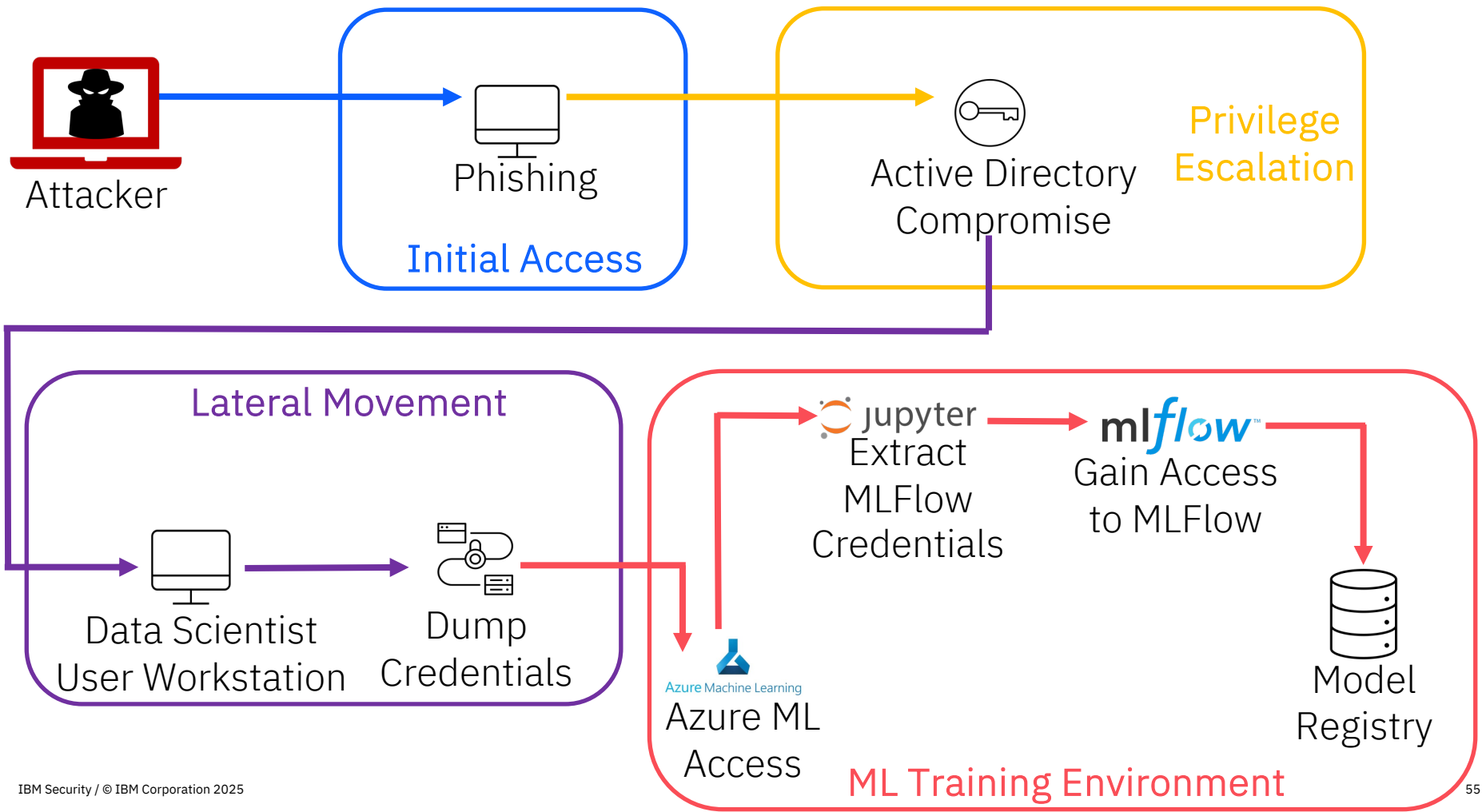
#2: SageMaker - Lateral Movement from SCM System to Cloud Compute

#3: SageMaker - Lateral Movement to Cloud Compute using Malicious Lifecycle Configuration

#4: SageMaker - Model Theft from Model Registry

#5: SageMaker - Model Poisoning to gain Code Execution

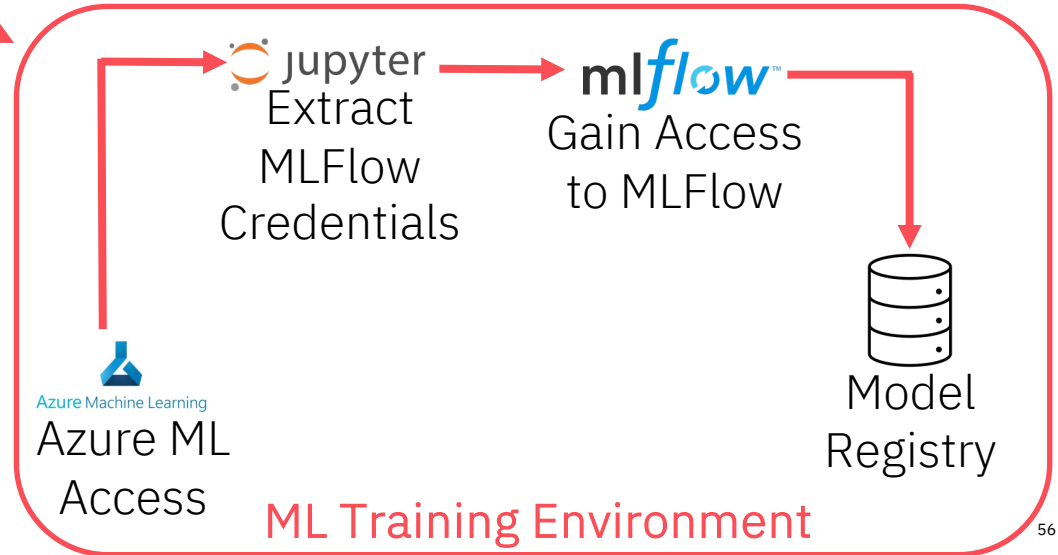
#6: Azure ML - Model Poisoning to gain Code Execution

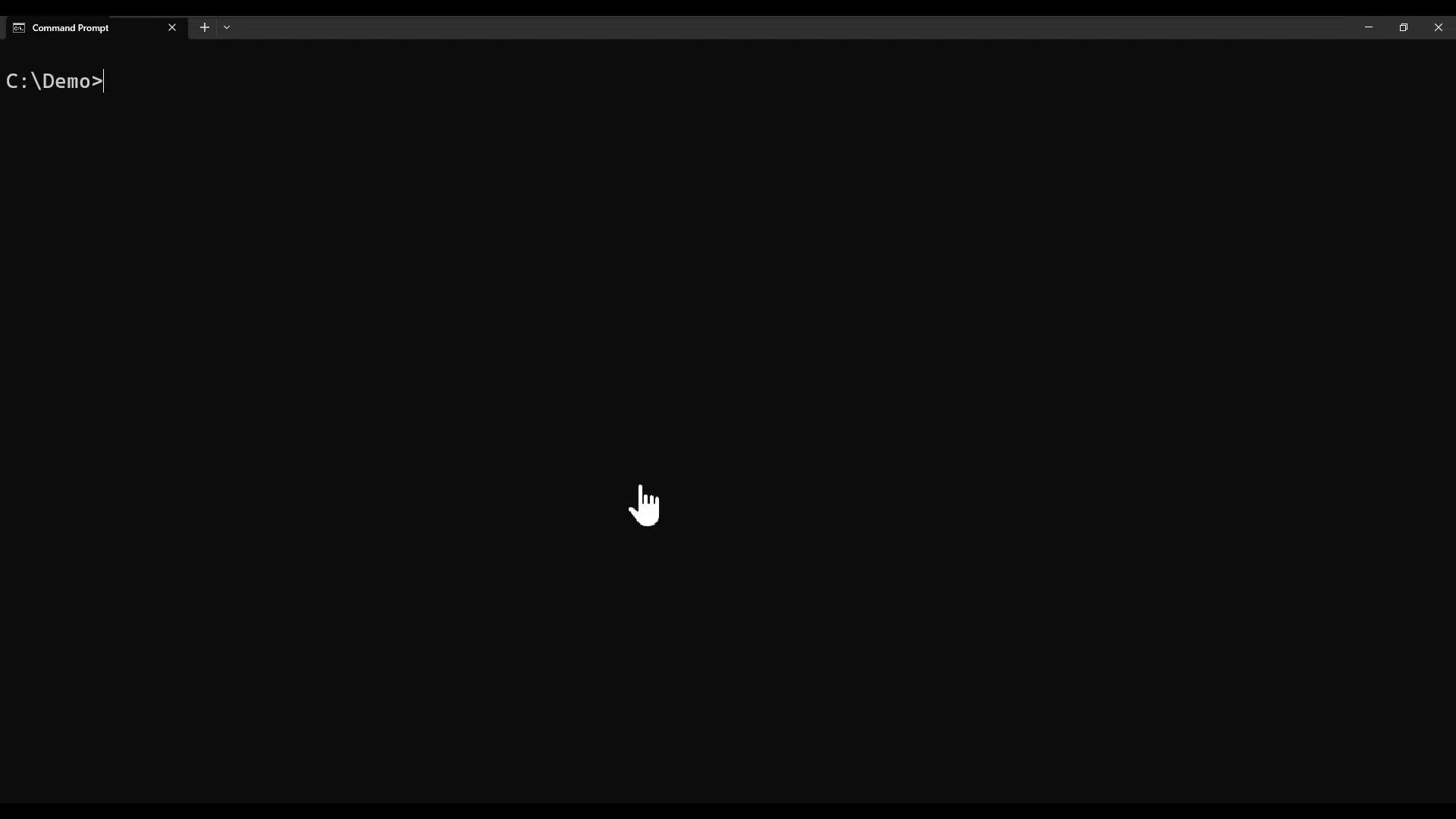




Attacker

Demo





Demos: Attack Scenarios

#1: MLFlow - Initial Access and Model Theft from Model Registry

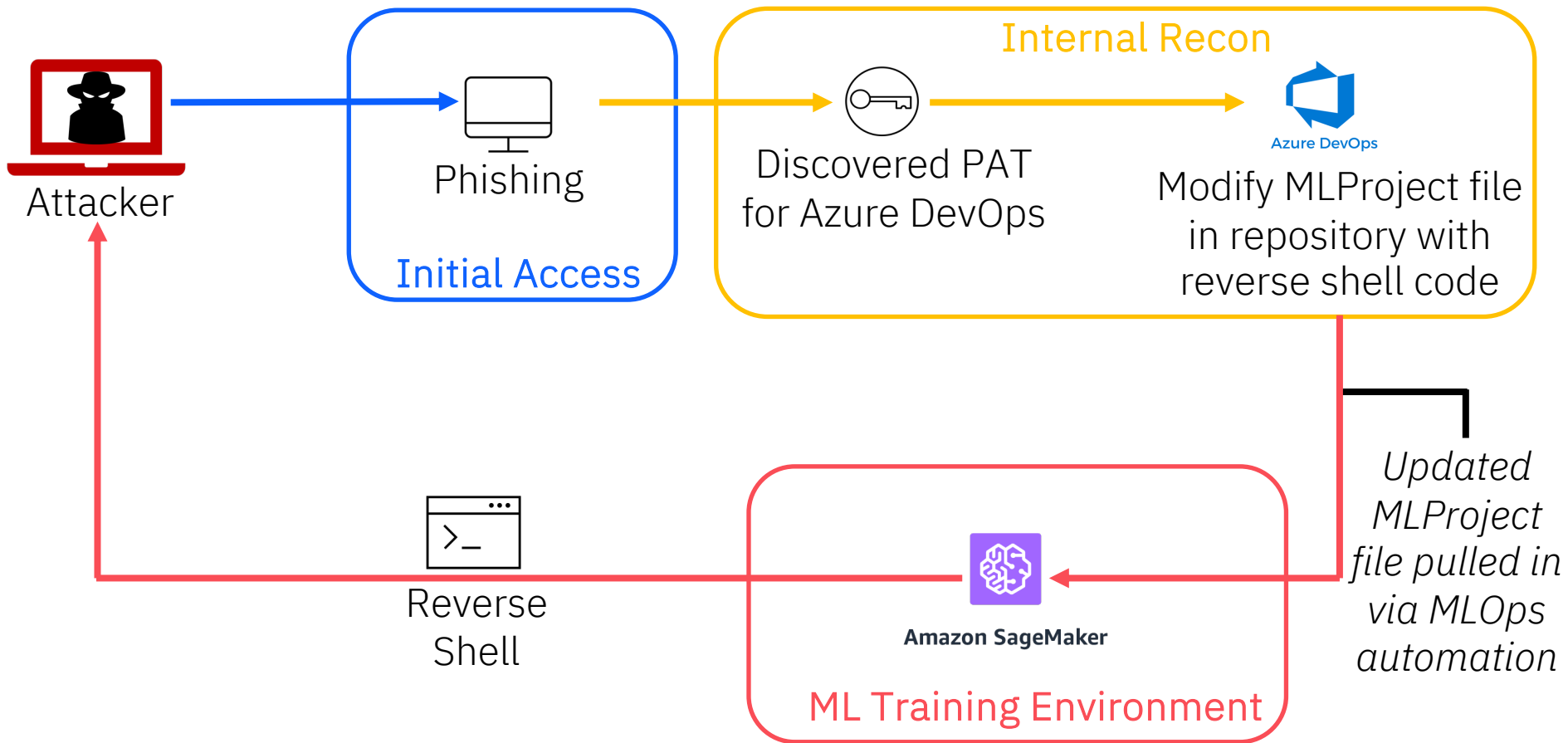
#2: SageMaker - Lateral Movement from SCM System to Cloud Compute

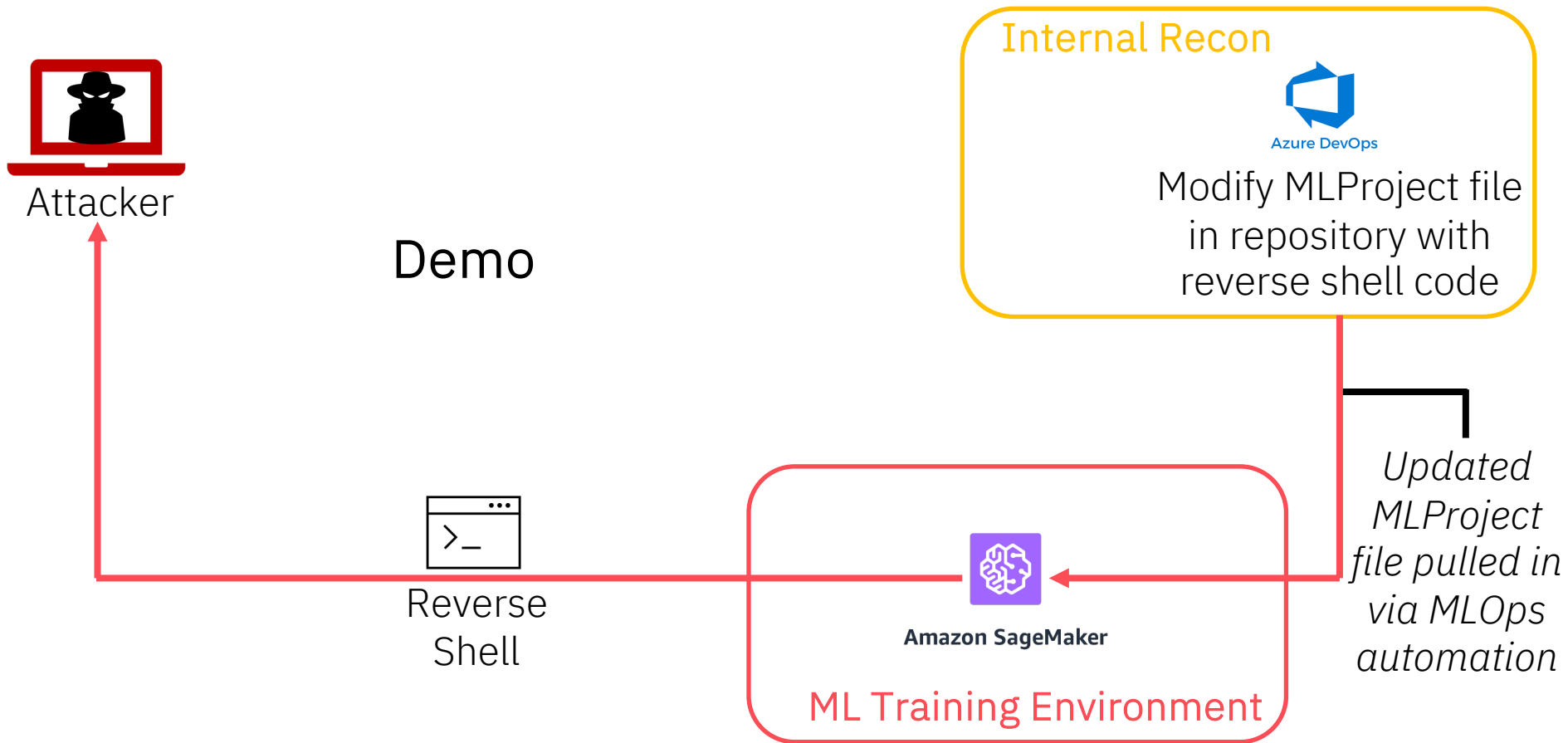
#3: SageMaker - Lateral Movement to Cloud Compute using Malicious Lifecycle Configuration

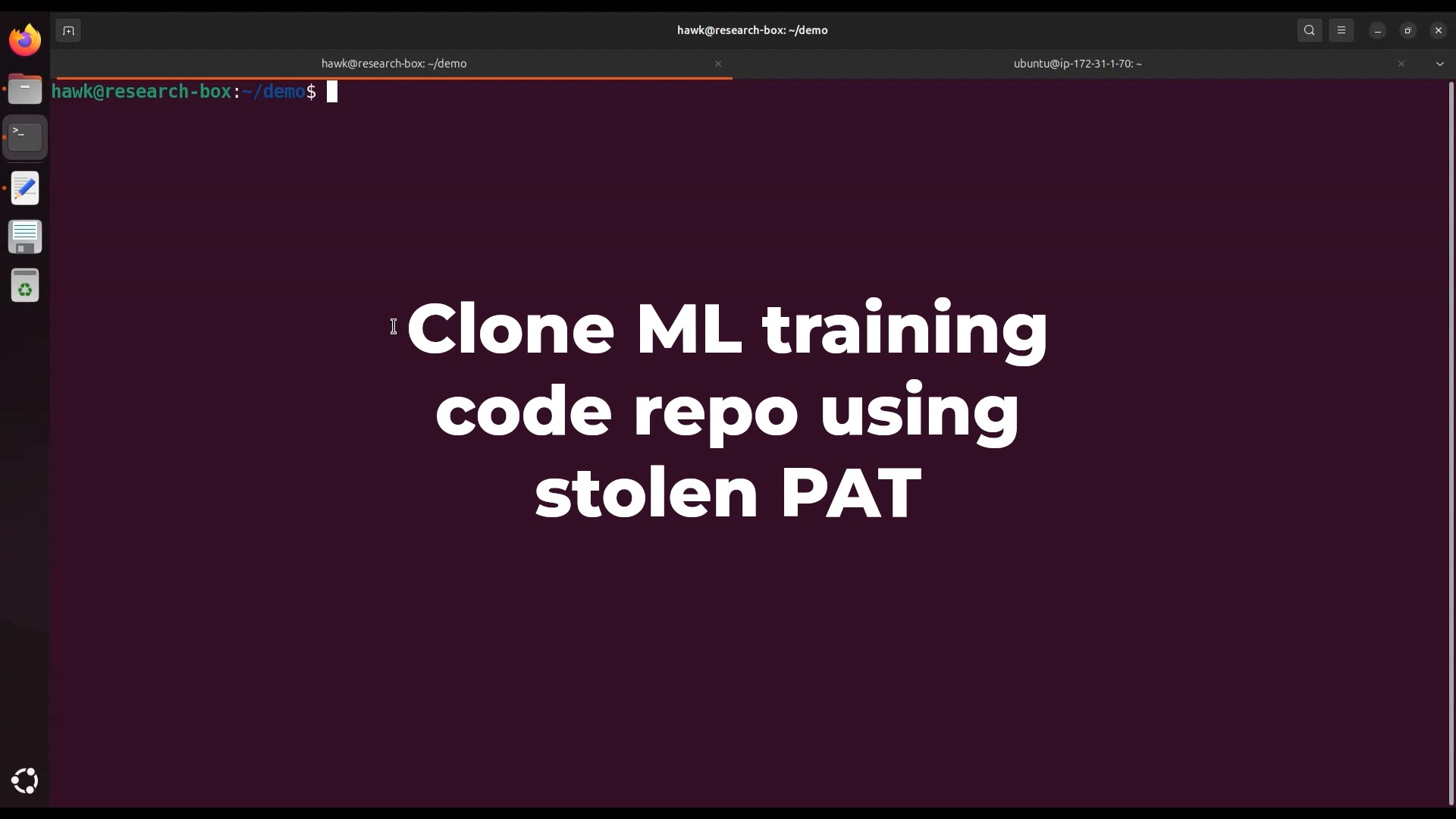
#4: SageMaker - Model Theft from Model Registry

#5: SageMaker - Model Poisoning to gain Code Execution

#6: Azure ML - Model Poisoning to gain Code Execution







**Clone ML training
code repo using
stolen PAT**

Demos: Attack Scenarios

#1: MLFlow - Initial Access and Model Theft from Model Registry

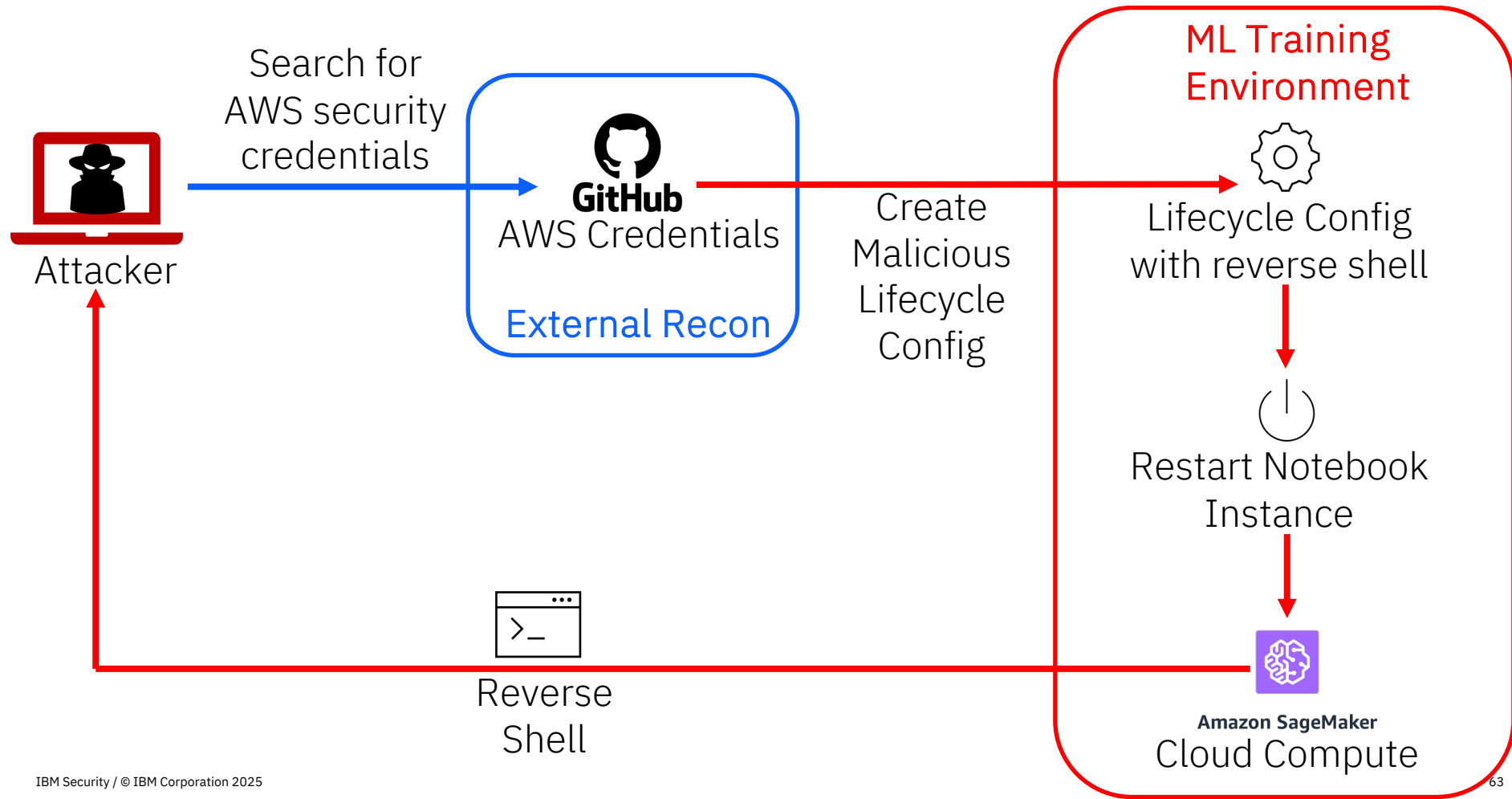
#2: SageMaker - Lateral Movement from SCM System to Cloud Compute

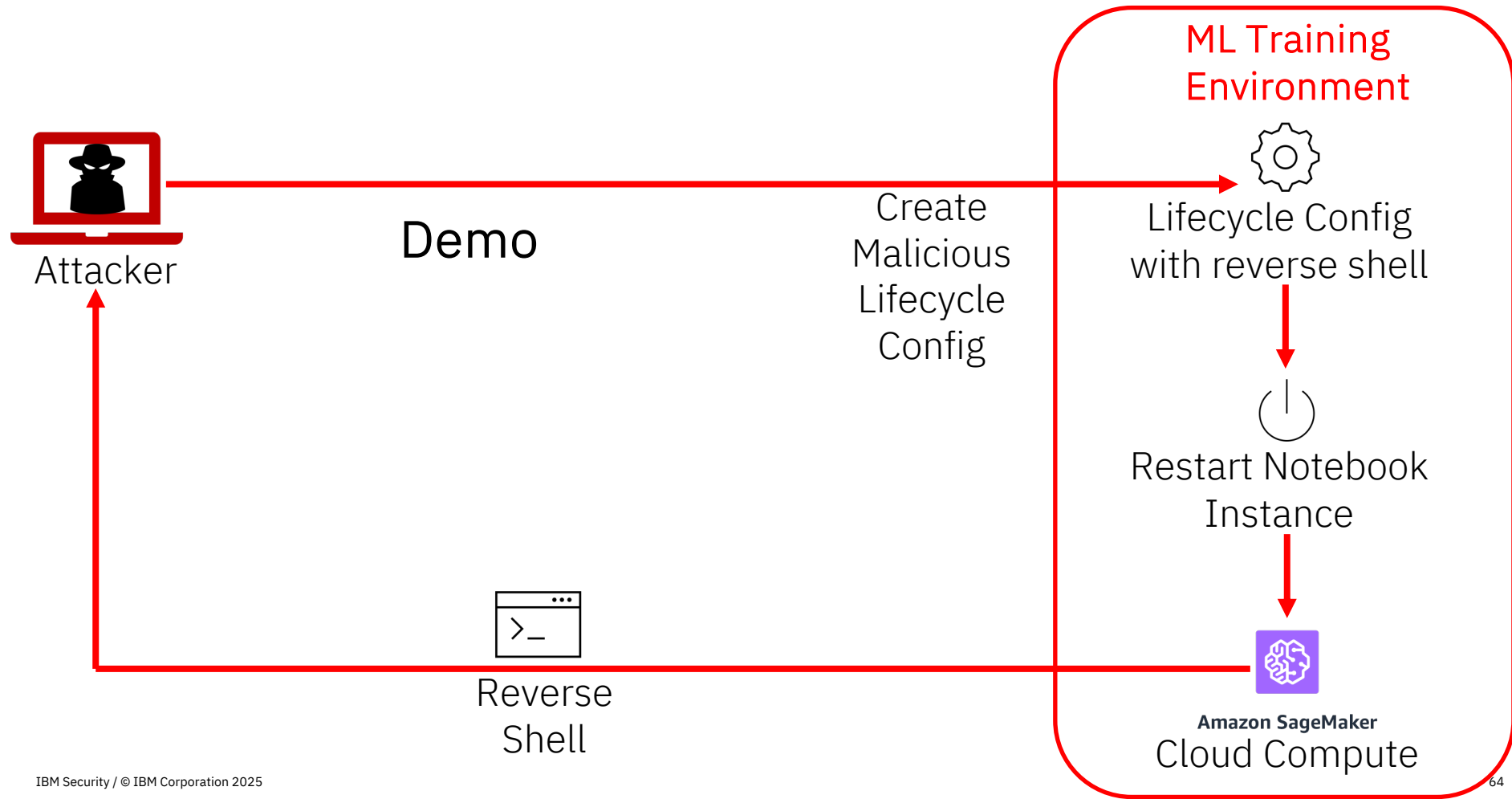
#3: SageMaker - Lateral Movement to Cloud Compute using Malicious Lifecycle Configuration

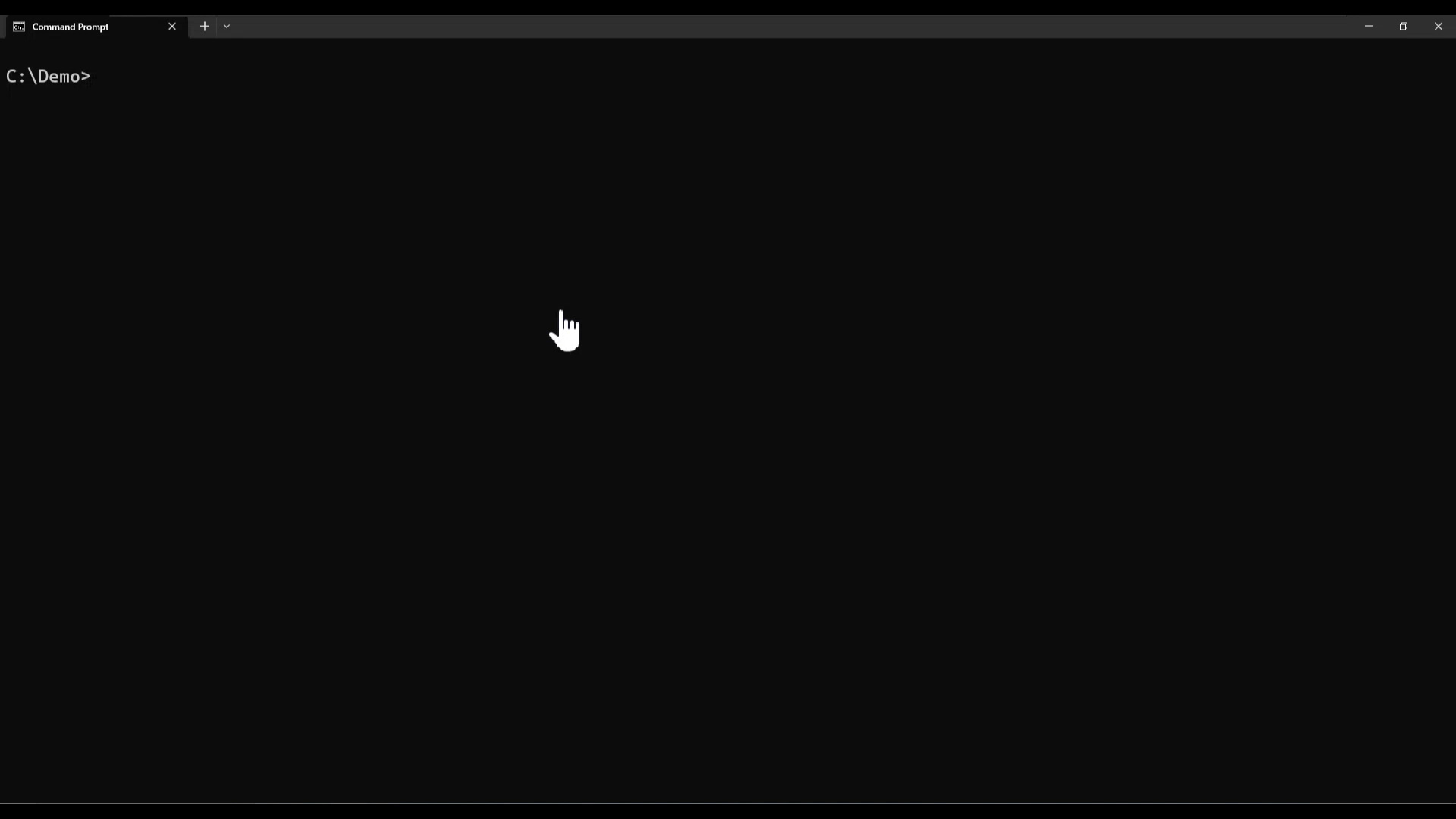
#4: SageMaker - Model Theft from Model Registry

#5: SageMaker - Model Poisoning to gain Code Execution

#6: Azure ML - Model Poisoning to gain Code Execution







Demos: Attack Scenarios

#1: MLFlow - Initial Access and Model Theft from Model Registry

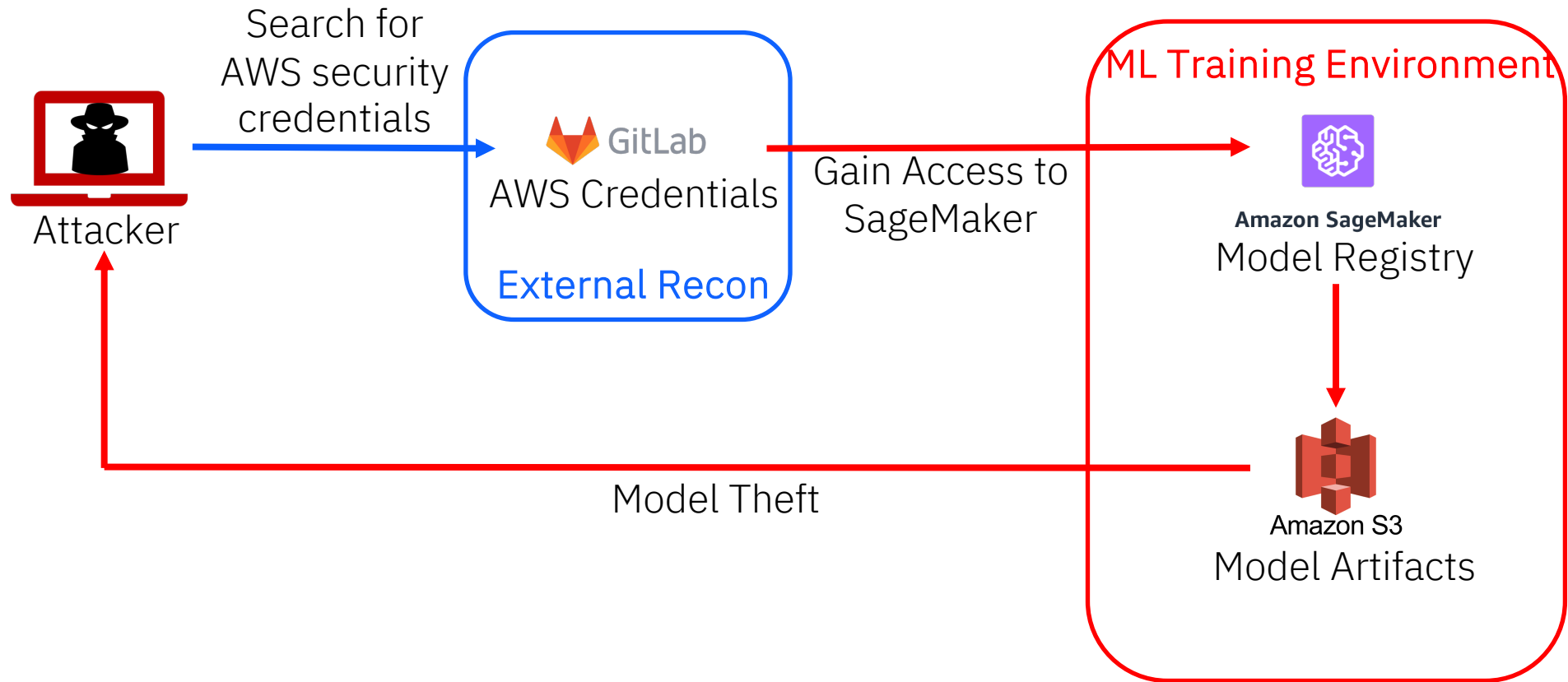
#2: SageMaker - Lateral Movement from SCM System to Cloud Compute

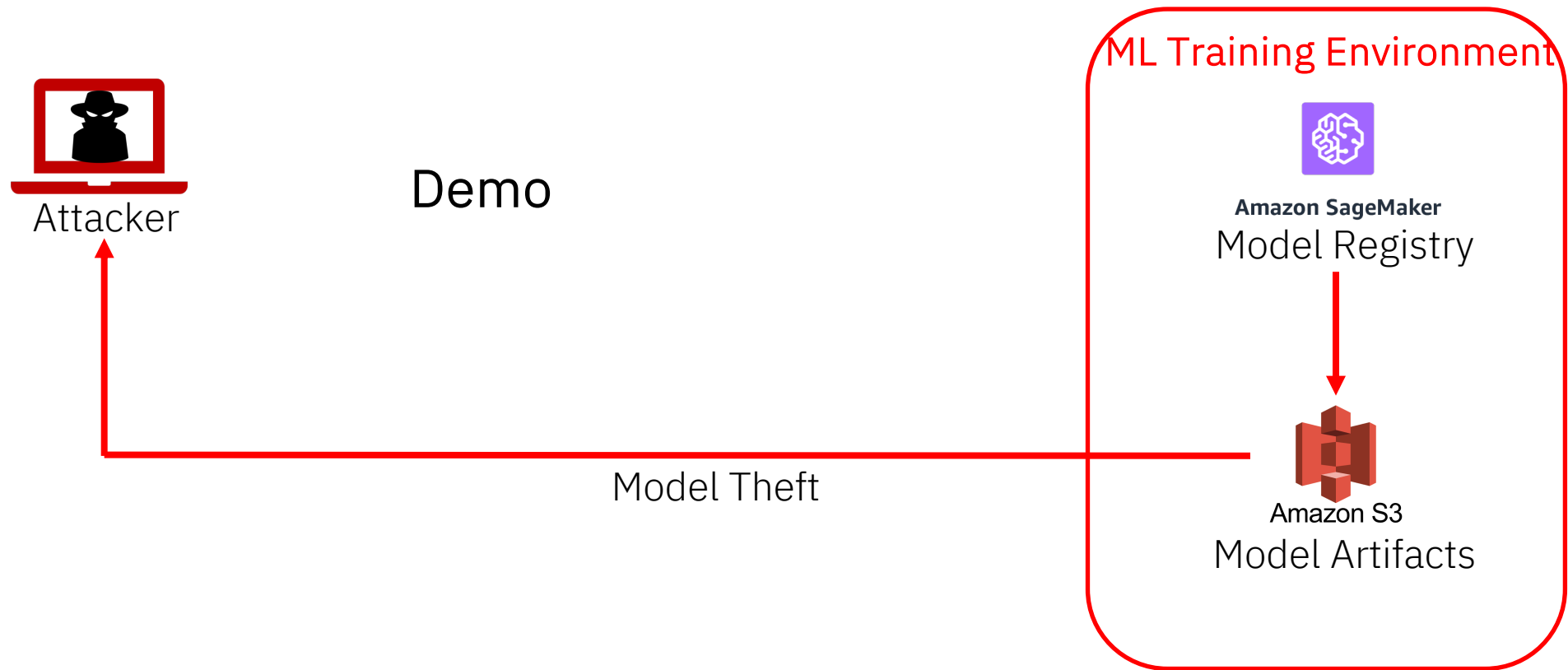
#3: SageMaker - Lateral Movement to Cloud Compute using Malicious Lifecycle Configuration

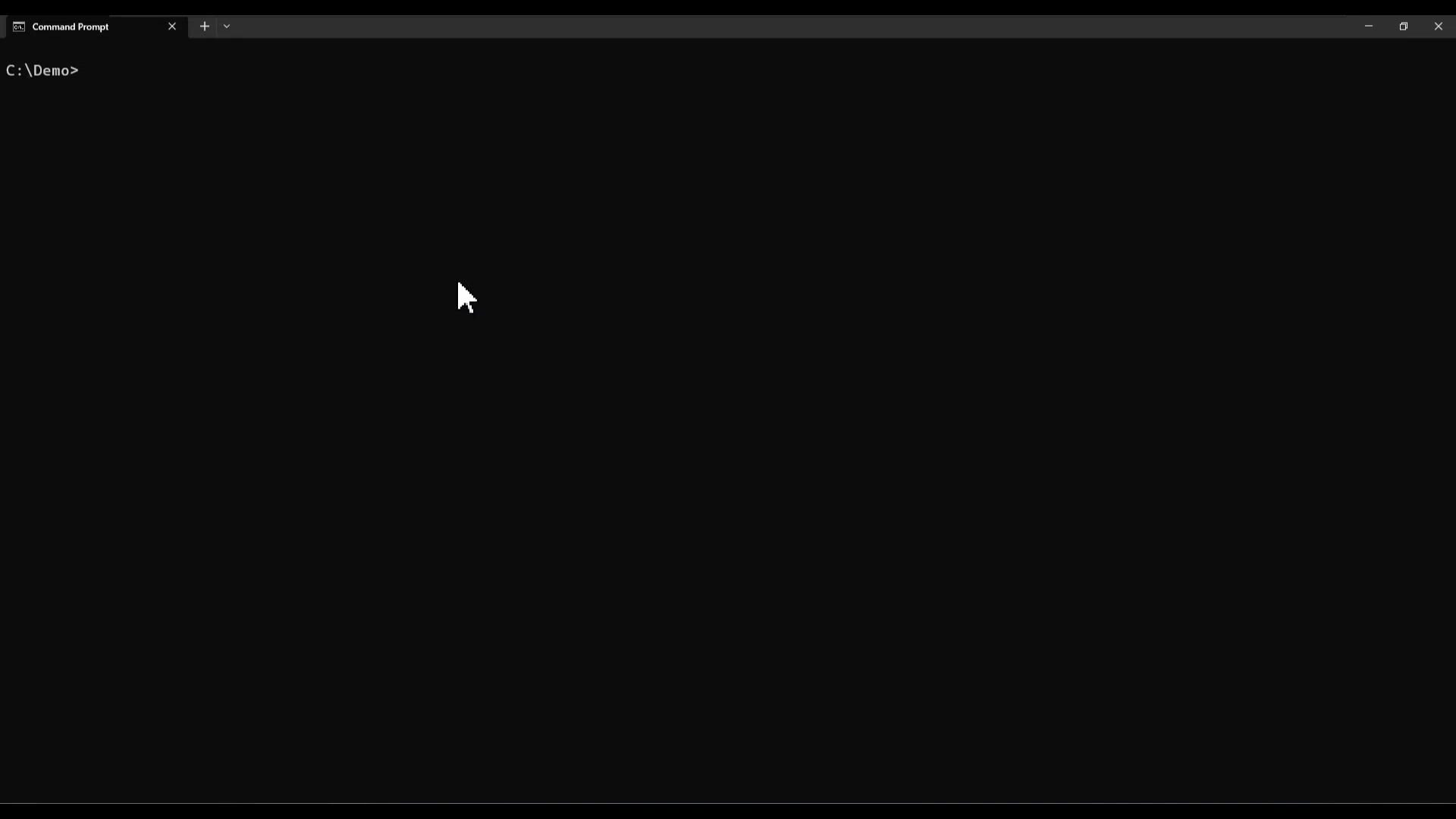
#4: SageMaker - Model Theft from Model Registry

#5: SageMaker - Model Poisoning to gain Code Execution

#6: Azure ML - Model Poisoning to gain Code Execution







Demos: Attack Scenarios

#1: MLFlow - Initial Access and Model Theft from Model Registry

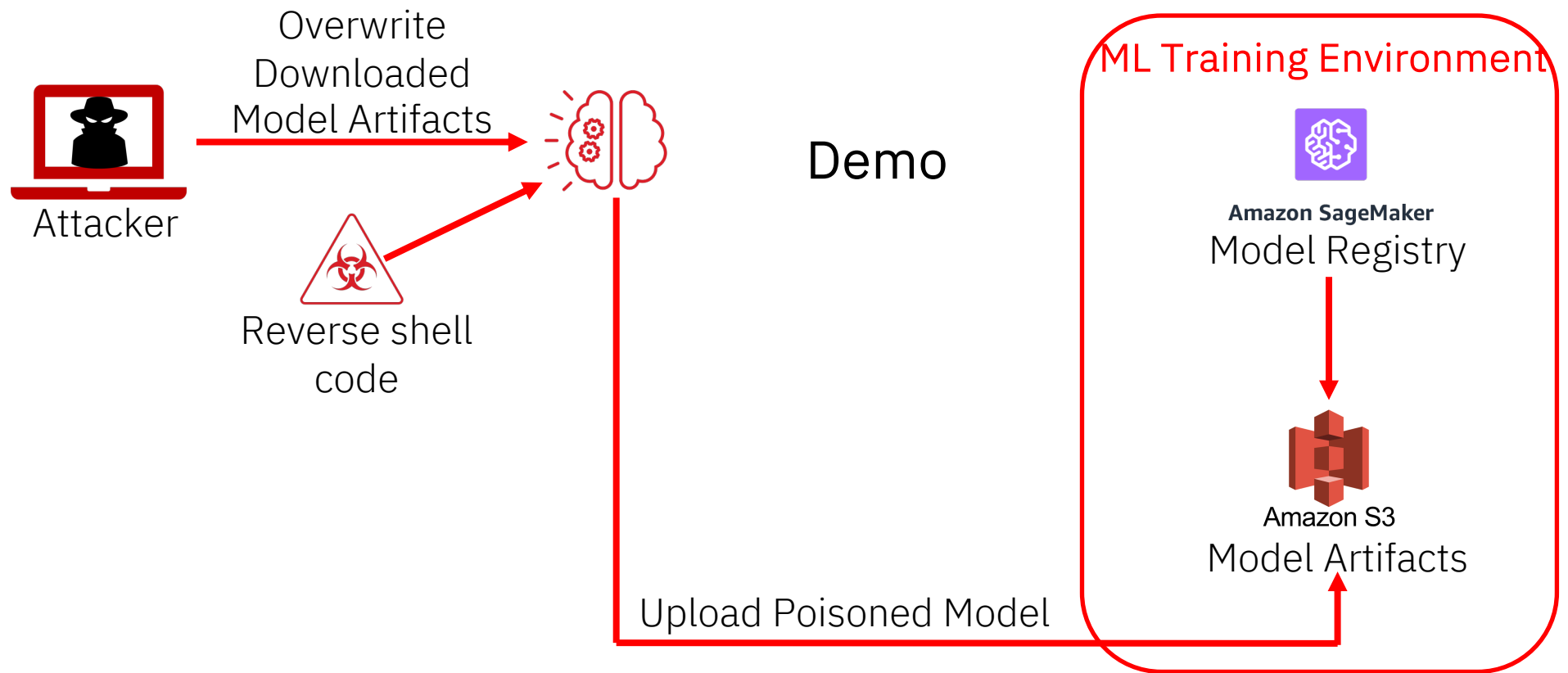
#2: SageMaker - Lateral Movement from SCM System to Cloud Compute

#3: SageMaker - Lateral Movement to Cloud Compute using Malicious Lifecycle Configuration

#4: SageMaker - Model Theft from Model Registry

#5: SageMaker - Model Poisoning to gain Code Execution

#6: Azure ML - Model Poisoning to gain Code Execution





C:\Demo>



**Showing model
artifacts of model
we previously
downloaded**

Demos: Attack Scenarios

#1: MLFlow - Initial Access and Model Theft from Model Registry

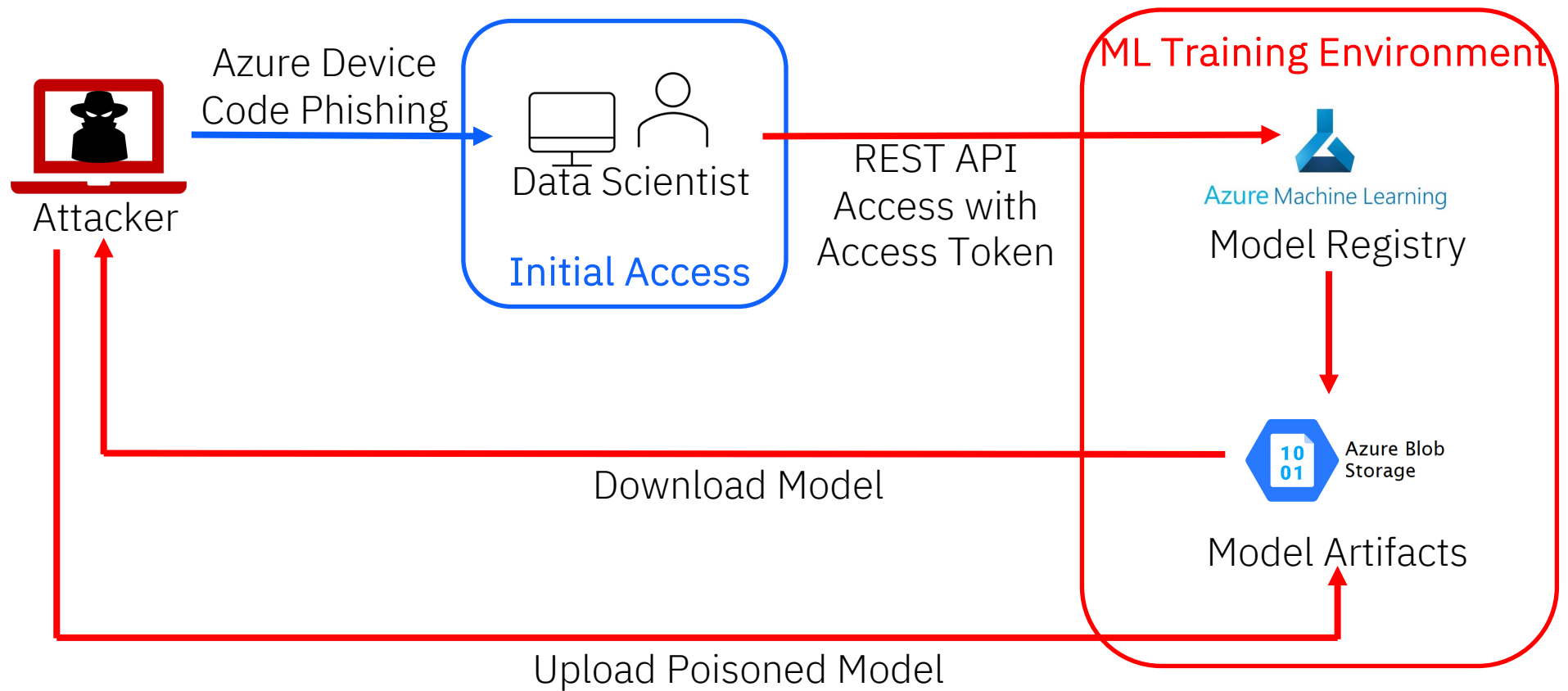
#2: SageMaker - Lateral Movement from SCM System to Cloud Compute

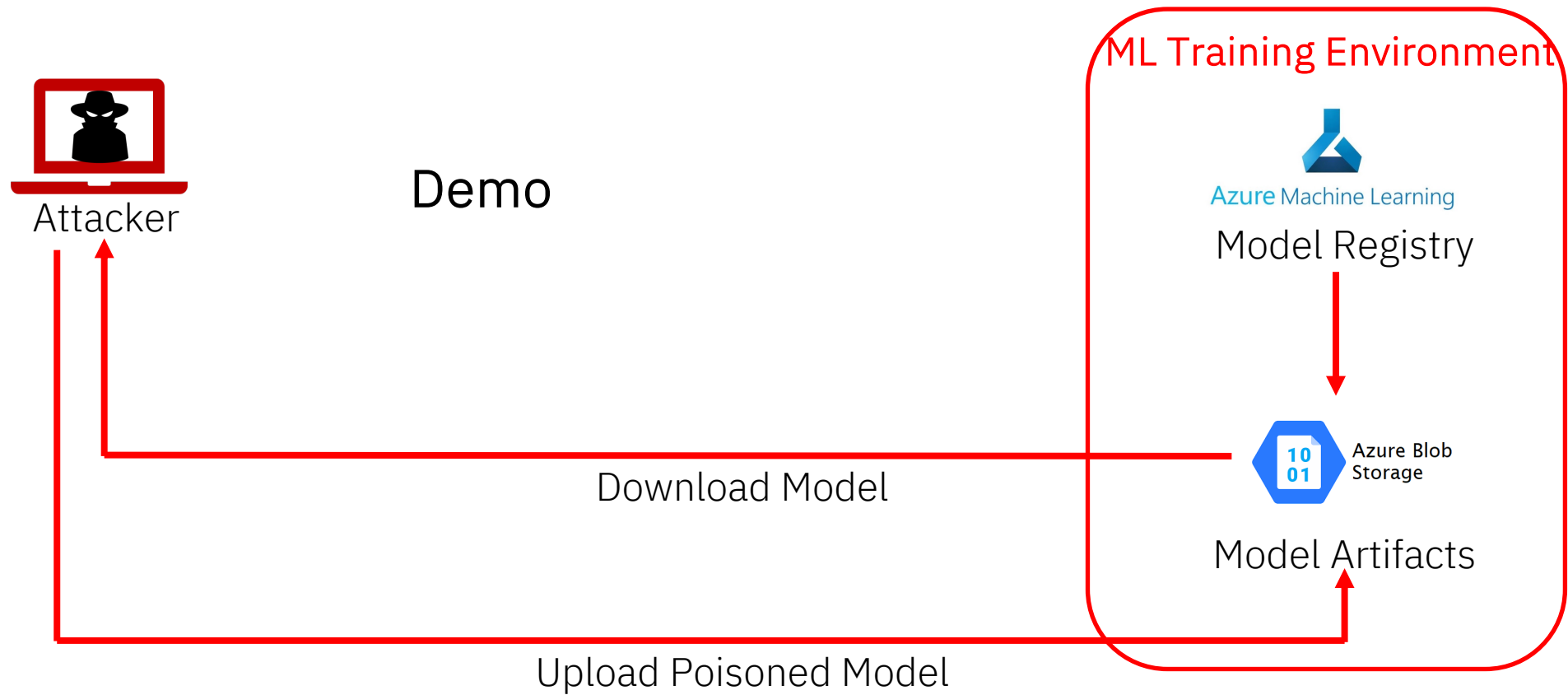
#3: SageMaker - Lateral Movement to Cloud Compute using Malicious Lifecycle Configuration

#4: SageMaker - Model Theft from Model Registry

#5: SageMaker - Model Poisoning to gain Code Execution

#6: Azure ML - Model Poisoning to gain Code Execution



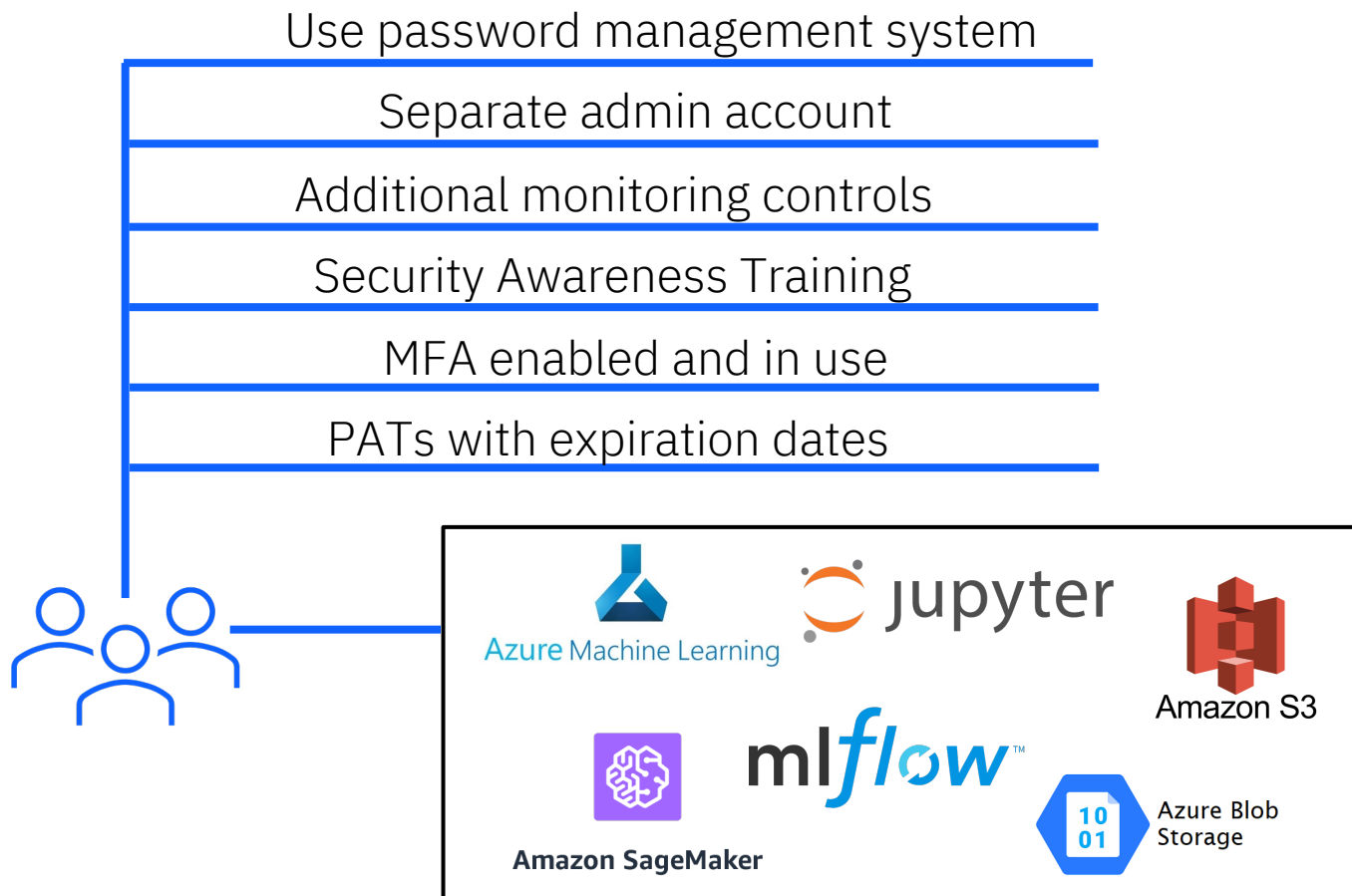


```
C:\Users\hawk\Desktop\Demo.txt - Notepad++
File Edit Search View Encoding Language Settings Tools Macro Run Plugins Window ?
[Icons]
Demo.txt [poison_model.py]
1 ===Azure ML===
2
3 MLOKit.exe check /platform:azureml /credential:%CRED%
4
5 MLOKit.exe list-projects /platform:azureml /subscription-id:47c5aaab-dbda-44ca-802e-00
6
7 MLOKit.exe list-models /platform:azureml /credential:%CRED% /subscription-id:47c5aaab-
8
9 MLOKit.exe download-model /platform:azureml /credential:%CRED% /subscription-id:47c5aa
10
11 MLOKit.exe poison-model /platform:azureml /credential:%CRED% /subscription-id:47c5aaab
12
13
14
15
16
17
18
19
20
21
22
23
Normal text file length: 886 lines: 28 Ln: 7 Col: 12 Pos: 206 Windows (CR LF) UTF-8 INS
```


Protecting ML Training Environments



Users



Notebook Environments

Password protect notebook

IP address restrictions

Limits to kernel execution times

Use virtual environment

Run as non-root account

No cleartext credentials/secrets



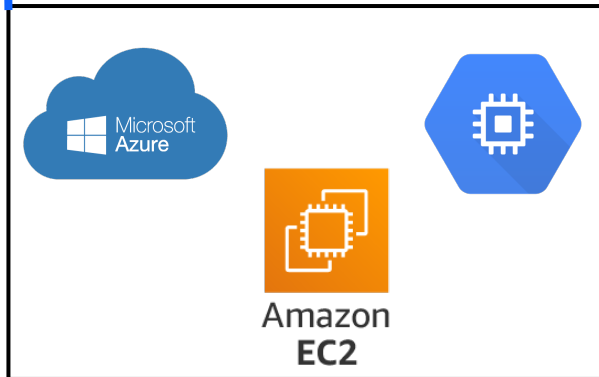
Cloud Compute

Enable auto-shutdown and auto-start schedule

Delete compute if no longer needed

Disable unneeded services

Configure role-based access



Model Artifact Storage and Registry

Cleanup/delete old model artifacts

Restrict access to backend storage

IP-address restrictions

Enable logging and apply detection rules

Implement model integrity verification



Detection Guidance – Summary

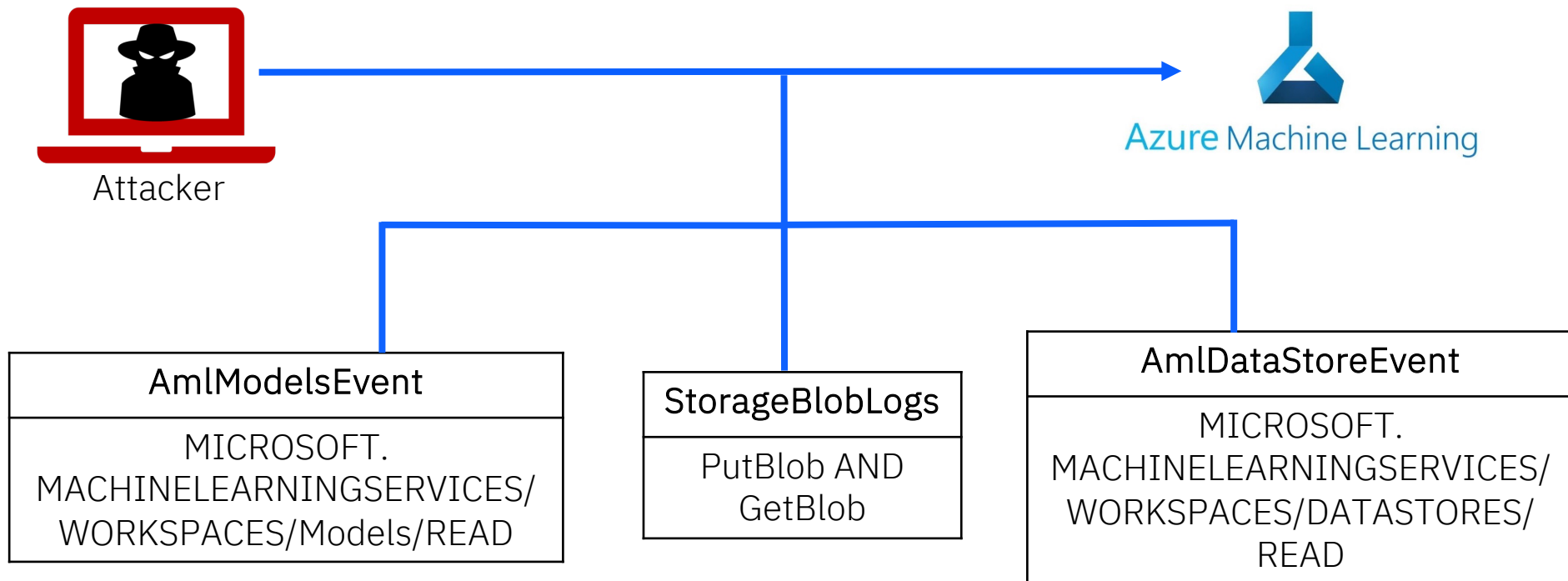
Azure ML Detections 
<https://github.com/h4wkst3r/KQL-Queries>

Dataset Poisoning
Dataset Recon
Dataset Theft
Model Poisoning
Model Recon
Model Theft

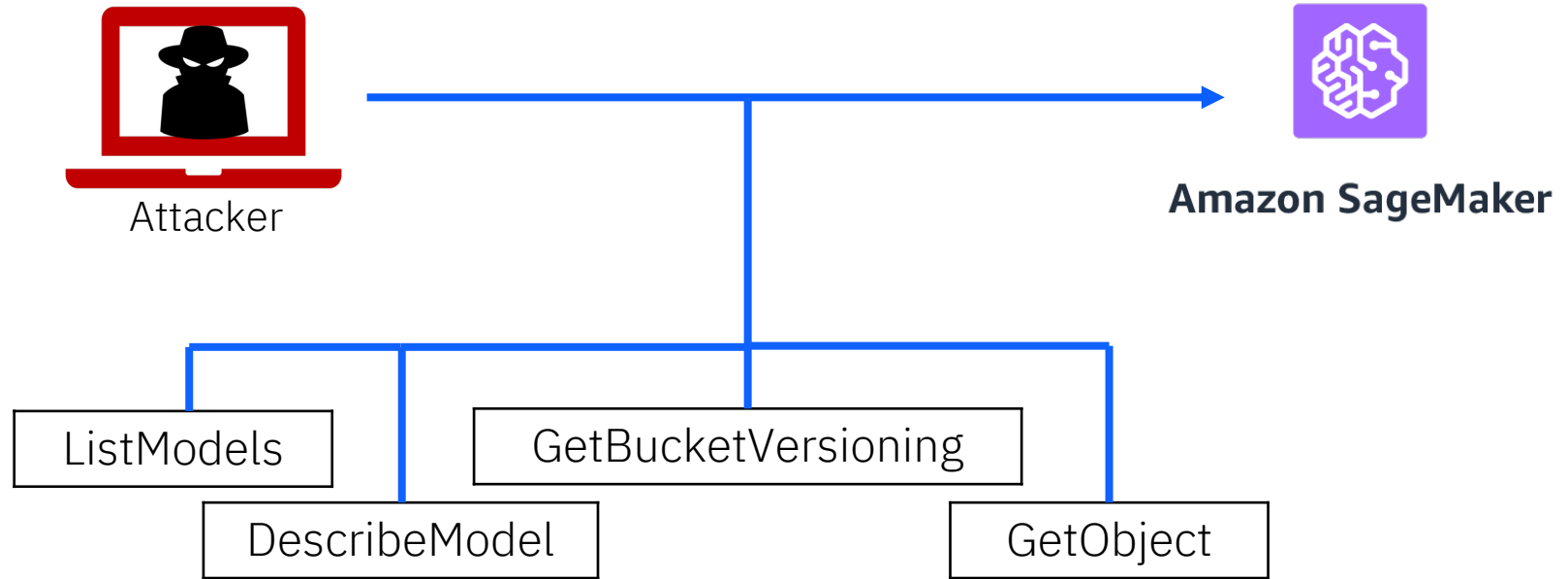
SageMaker Detections 
<https://github.com/h4wkst3r/CloudTrail-Queries>

Model Poisoning
Model Theft
Malicious Lifecycle Config

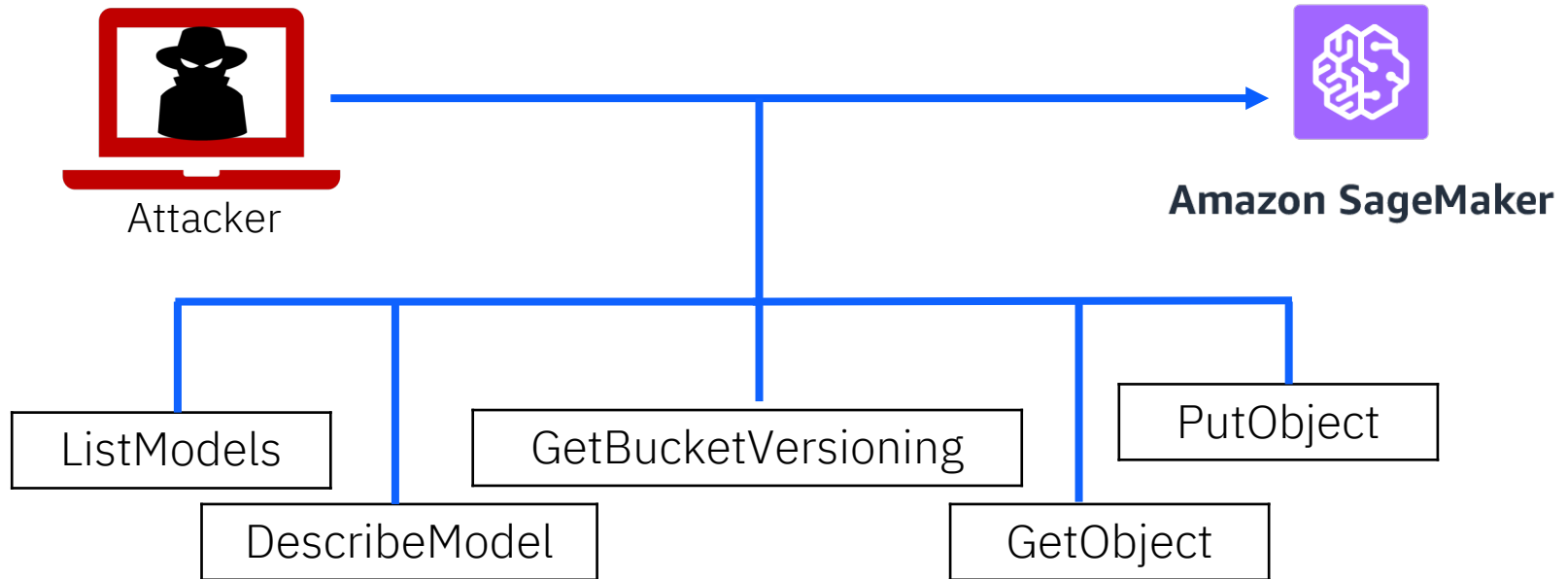
Detection Guidance – Azure ML Model Poisoning



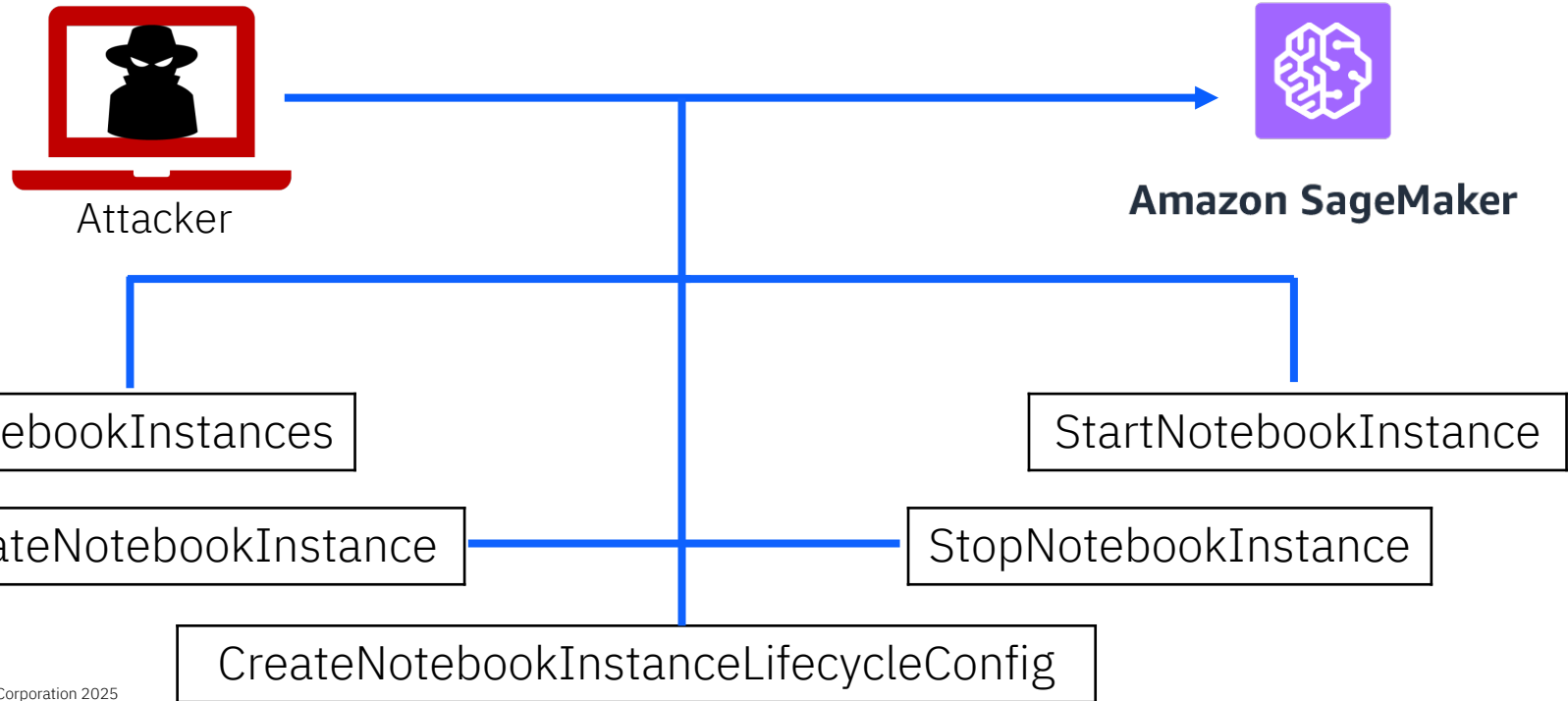
Detection Guidance – SageMaker Model Theft



Detection Guidance – SageMaker Model Poisoning



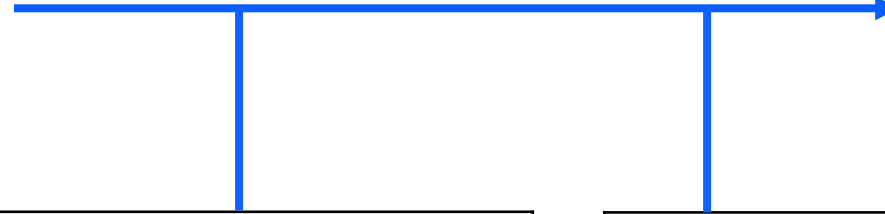
Detection Guidance – SageMaker Malicious Lifecycle Configuration



Detection Guidance – MLFlow



Attacker



mlflow™

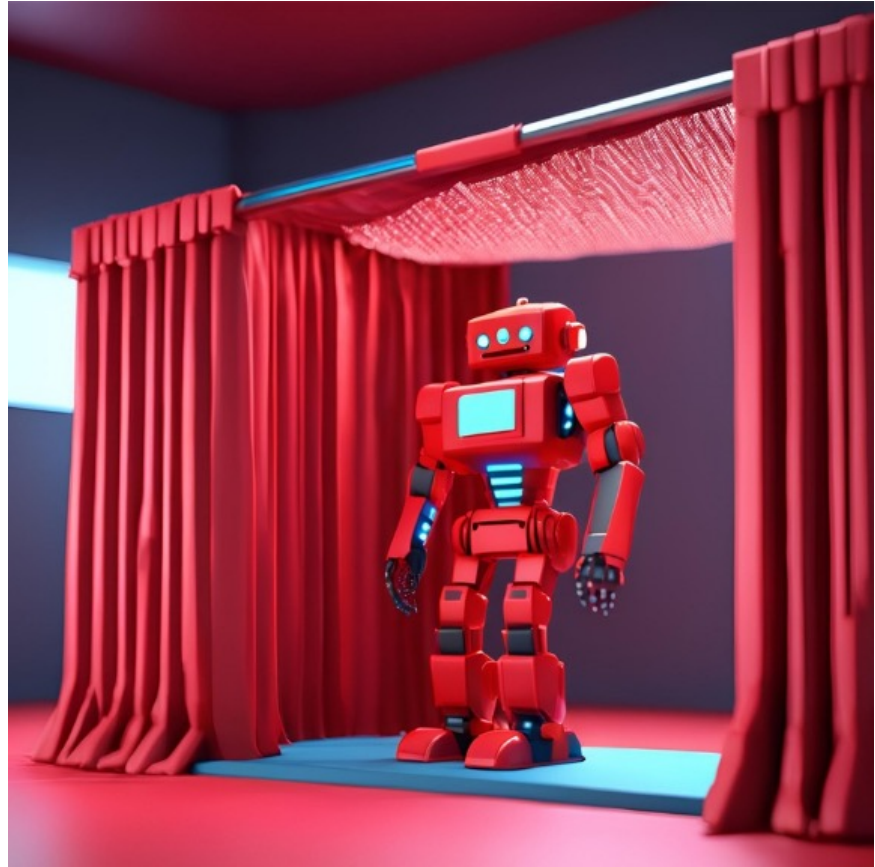
/api/2.0/mlflow/model-versions/search

/get-artifact

```
grep -i /api/2.0/mlflow/model-versions/search access.log
b/2025:20:59:42 -0500] "GET /api/2.0/mlflow/model-versions/search HTTP/1.1" 200 904 "-" "MLOKit-e977ac02118a3cb2
b/2025:20:59:42 -0500] "GET /api/2.0/mlflow/model-versions/search HTTP/1.1" 200 904 "-" "MLOKit-e977ac02118a3cb2
```

```
grep -i /get-artifact access.log
b/2025:21:05:59 -0500] "GET /get-artifact?path=ElasticNet/code/data.py&run_id=f0b129a5e11c4bb6b1c0459a2f5ae7f6 HTTP/1.1" 200 0 "-" "MLOKit-e977ac02118a3cb2"
b/2025:21:06:00 -0500] "GET /get-artifact?path=ElasticNet/code/params.py&run_id=f0b129a5e11c4bb6b1c0459a2f5ae7f6 HTTP/1.1" 200 0 "-" "MLOKit-e977ac02118a3cb2"
b/2025:21:06:01 -0500] "GET /get-artifact?path=ElasticNet/code/train.py&run_id=f0b129a5e11c4bb6b1c0459a2f5ae7f6 HTTP/1.1" 200 0 "-" "MLOKit-e977ac02118a3cb2"
b/2025:21:06:01 -0500] "GET /get-artifact?path=ElasticNet/code/utils.py&run_id=f0b129a5e11c4bb6b1c0459a2f5ae7f6 HTTP/1.1" 200 0 "-" "MLOKit-e977ac02118a3cb2"
b/2025:21:06:02 -0500] "GET /get-artifact?path=ElasticNet/conda.yaml&run_id=f0b129a5e11c4bb6b1c0459a2f5ae7f6 HTTP/1.1" 200 0 "-" "MLOKit-e977ac02118a3cb2"
b/2025:21:06:02 -0500] "GET /get-artifact?path=ElasticNet/input_example.json&run_id=f0b129a5e11c4bb6b1c0459a2f5ae7f6 HTTP/1.1" 200 0 "-" "MLOKit-e977ac02118a3cb2"
b/2025:21:06:03 -0500] "GET /get-artifact?path=ElasticNet/model.pkl&run_id=f0b129a5e11c4bb6b1c0459a2f5ae7f6 HTTP/1.1" 200 0 "-" "MLOKit-e977ac02118a3cb2"
b/2025:21:06:03 -0500] "GET /get-artifact?path=ElasticNet/output_example.json&run_id=f0b129a5e11c4bb6b1c0459a2f5ae7f6 HTTP/1.1" 200 0 "-" "MLOKit-e977ac02118a3cb2"
```

Conclusion



Conclusion

01

ML training environments contain highly sensitive and business critical data

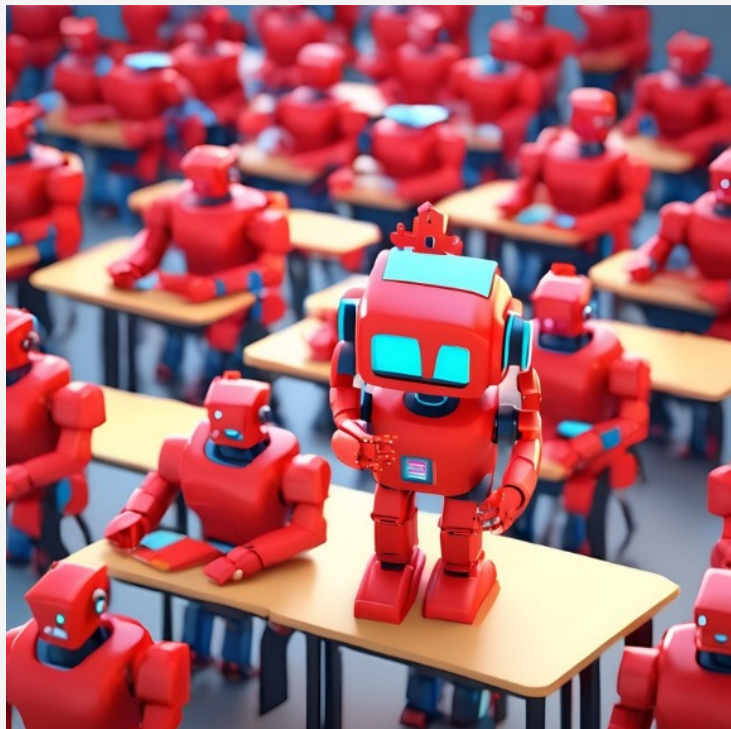
02

We need to understand these systems so we can protect them

03

Unauthorized access to these environments could be significant

Questions?



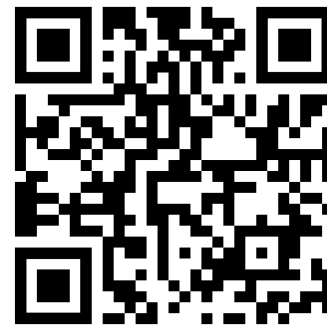
Brett Hawkins

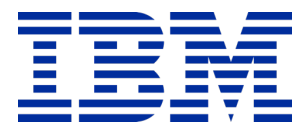
@h4wkst3r  

Blog Post -



MLOKit Tool -





References

- <https://twitter.com/retBandit>
- <https://twitter.com/h4wkst3r>
- <https://www.ibm.com/downloads/documents/us-en/11630e2cbc302316>
- <https://github.com/xforced/MLOKit>
- <https://twitter.com/azarzaror>
- <https://web.archive.org/web/20241214000321/https://www.panoptica.app/blog/protect-your-environment-when-working-with-amazon-sagemaker>
- <https://www.ibm.com/topics/mlops>
- <https://neptune.ai/blog/mlops-tools-platforms-landscape>
- <https://www.databricks.com/glossary/jupyter-notebook>
- <https://www.ibm.com/docs/en/z-devops-guide?topic=applications-source-code-management>
- <https://azure.microsoft.com/en-us/resources/cloud-computing-dictionary/what-is-cloud-computing>
- <https://mlflow.org/docs/latest/tracking/artifacts-stores.html>

References

- <https://wandb.ai/site/articles/what-is-an-ML-model-registry/>
- <https://learn.microsoft.com/en-us/azure/devops/user-guide/what-is-azure-devops>
- <https://about.gitlab.com/>
- <https://github.com/>
- <https://aws.amazon.com/sagemaker/>
- <https://learn.microsoft.com/en-us/azure/machine-learning/overview-what-is-azure-machine-learning>
- <https://mlflow.org/>
- <https://docs.aws.amazon.com/sagemaker/latest/dg/model-registry.html>
- <https://learn.microsoft.com/en-us/azure/machine-learning/how-to-manage-models?view=azureml-api-2&tabs=cli>
- <https://aws.amazon.com/s3/>
- <https://learn.microsoft.com/en-us/azure/storage/blobs/storage-blobs-introduction>
- <https://www.ibm.com/services/adversary-simulation>

References

- <https://mlflow.org/docs/latest/rest-api.html>
- <https://learn.microsoft.com/en-us/azure/devops/organizations/accounts/use-personal-access-tokens-to-authenticate?view=azure-devops&tabs=Windows>
- <https://mlflow.org/docs/latest/projects.html>
- <https://github.com/xforcered/ADOKit>
- <https://aws.amazon.com/blogs/machine-learning/build-an-end-to-end-mlops-pipeline-using-amazon-sagemaker-pipelines-github-and-github-actions/>
- <https://docs.aws.amazon.com/IAM/latest/UserGuide/security-creds.html>
- <https://docs.aws.amazon.com/sagemaker/latest/dg/security-iam-awsmanpol.html>
- <https://docs.aws.amazon.com/sagemaker/latest/dg/nbi.html>
- <https://docs.aws.amazon.com/sagemaker/latest/dg/notebook-lifecycle-config.html>
- <https://docs.aws.amazon.com/AmazonS3/latest/userguide/security-iam-awsmanpol.html>
- <https://jfrog.com/blog/from-mlops-to-mloops-exposing-the-attack-surface-of-machine-learning-platforms/>

References

- <https://docs.python.org/3/library/pickle.html>
- <https://gist.githubusercontent.com/h4wkst3r/2c30a3d39e20b7cd8606211ba3132d85/raw/e4e253b052f5b916134409fcb61b91c49b91d912/CreatePickle.py>
- <https://github.com/coldwaterq/MaliciousPickles>
- <https://github.com/moohax/Charcuterie>
- <https://github.com/trailofbits/fickling>
- <https://github.com/hiddenlayerai/HiddenPickle>
- <https://docs.aws.amazon.com/sagemaker/latest/dg/how-it-works-deployment.html>
- <https://www.microsoft.com/en-us/security/blog/2025/02/13/storm-2372-conducts-device-code-phishing-campaign/>
- <https://learn.microsoft.com/en-us/entra/identity-platform/access-tokens>
- <https://learn.microsoft.com/en-us/rest/api/azureml/>
- <https://learn.microsoft.com/en-us/azure/machine-learning/concept-workspace?view=azureml-api-2>
- <https://docs.gunicorn.org/en/stable/settings.html>

References

- <https://learn.microsoft.com/en-us/microsoft-365/enterprise/subscriptions-licenses-accounts-and-tenants-for-microsoft-cloud-offerings?view=o365-worldwide>
- <https://learn.microsoft.com/en-us/azure/machine-learning/concept-data?view=azureml-api-2>
- <https://learn.microsoft.com/en-us/azure/machine-learning/concept-endpoints?view=azureml-api-2>
- <https://blog.gitguardian.com/how-to-handle-secrets-in-jupyter-notebooks/>
- <https://medium.com/@techlatest.net/security-best-practices-for-ai-ml-in-jupyter-notebooks-a-blog-post-on-the-security-best-practices-c0e0659cfccb>
- <https://github.com/h4wkst3r/KQL-Queries>
- <https://github.com/h4wkst3r/CloudTrail-Queries>
- <http://www.ibm.com/downloads/documents/us-en/11630e2cbc302316>
- [https://github.com/aws-samples/aws-customer-playbook-framework/blob/main/docs/Responding to SageMaker.md](https://github.com/aws-samples/aws-customer-playbook-framework/blob/main/docs/Responding%20to%20SageMaker.md)
- <https://aws.amazon.com/blogs/machine-learning/securing-mlflow-in-aws-fine-grained-access-control-with-aws-native-services/>

References

- <https://learn.microsoft.com/en-us/azure/storage/blobs/monitor-blob-storage?tabs=azure-portal>
- <https://learn.microsoft.com/en-us/azure/machine-learning/monitor-azure-machine-learning?view=azureml-api-2>
- <https://github.com/h4wkst3r/KQL-Queries/blob/main/AzureML/AzureMLModelPoisoning.kql>
- <https://github.com/h4wkst3r/CloudTrail-Queries/blob/main/SageMakerModelTheft.sql>
- <https://github.com/h4wkst3r/CloudTrail-Queries/blob/main/SageMakerModelPoisoning.sql>
- <https://github.com/h4wkst3r/CloudTrail-Queries/blob/main/SageMakerMaliciousLifecycleConfig.sql>
- <https://www.ibm.com/think/x-force/becoming-the-trainer-attacking-ml-training-infrastructure>