Prevent online fraud while ensuring a frictionless customer experience

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Fraud Leads to Friction

SURGE OF FRAUDULENT ACCOUNT REGISTRATIONS

Fake credentials, stolen identities, and bots make creating fake accounts fast and easy

SPIKE IN FRAUD ALERTS INCREASES ACCOUNT REVIEWS

Customer service teams, overwhelmed by alert volumes, struggle to keep up

FRICTION ADDED TO STOP FRAUDULENT REGISTRATIONS

Friction slows and lowers account registrations from legitimate customers

Use Case: Coupon Shopper

A Digital Platform

An online rewards platform that allows users to earn Gift Cards, rewards and coupons for online shopping.

Growing User Base

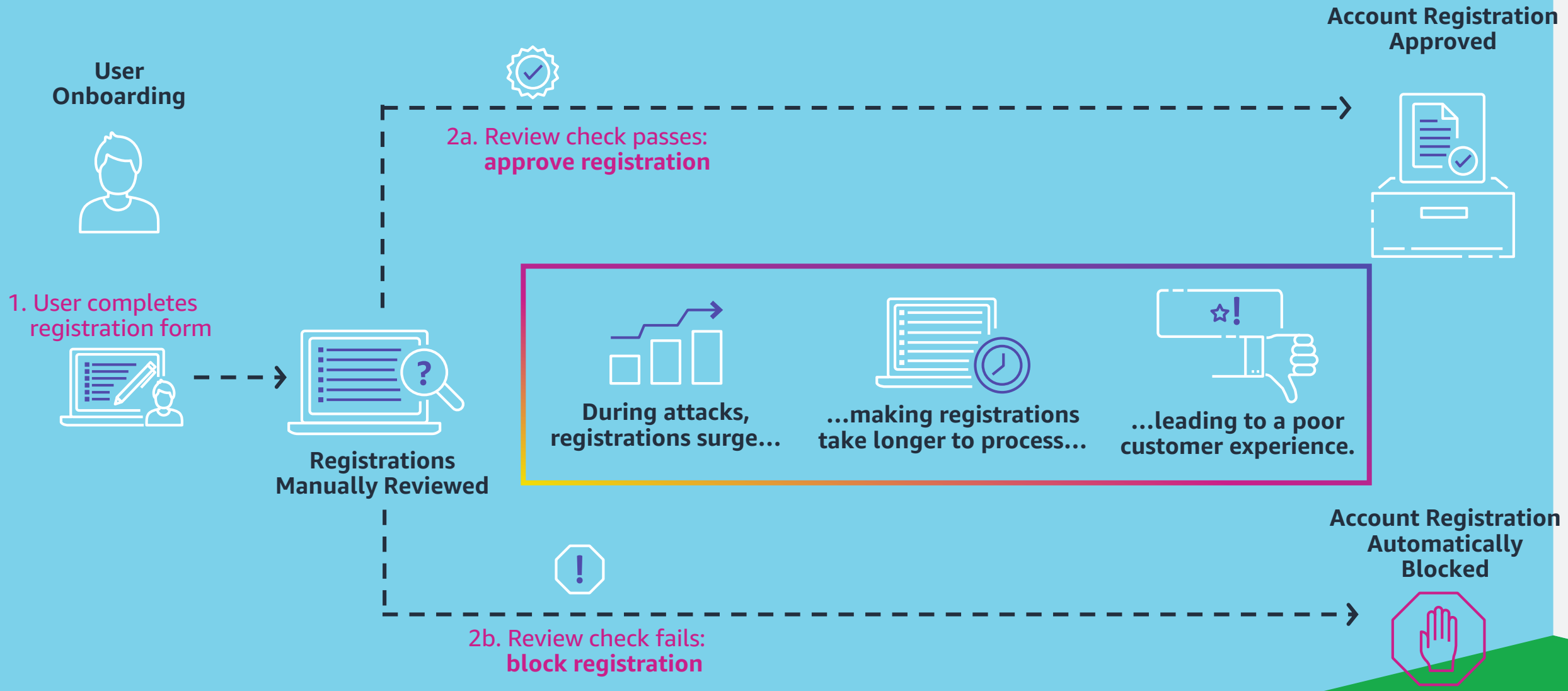
A growing user base with currently ~50k monthly active users. Awards millions of dollars worth of Gift Cards and coupons annually.

Increasing Online Fraud

Increasing fraudulent user registrations from bad actors, automated accounts/bots, who try to game the system to earn Gift Cards and coupons.

Net annual loss of ~$1.5 million

User Onboarding: with manual review

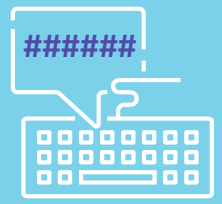


User Onboarding: with automated friction

User Onboarding

1. User completes registration form

Friction Added to Detect Bots and Email Tumbling



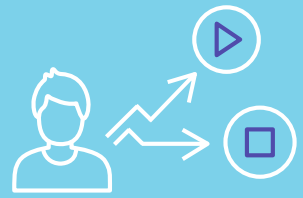
2a. Validate one-time password

2b. Complete CAPTCHA

3a. Friction check passes: **approve registration**

Account Registration Approved

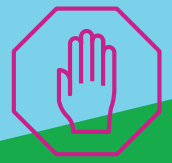
Friction makes the registration process more complex for ALL users…



…leading to the churn of legitimate customers.

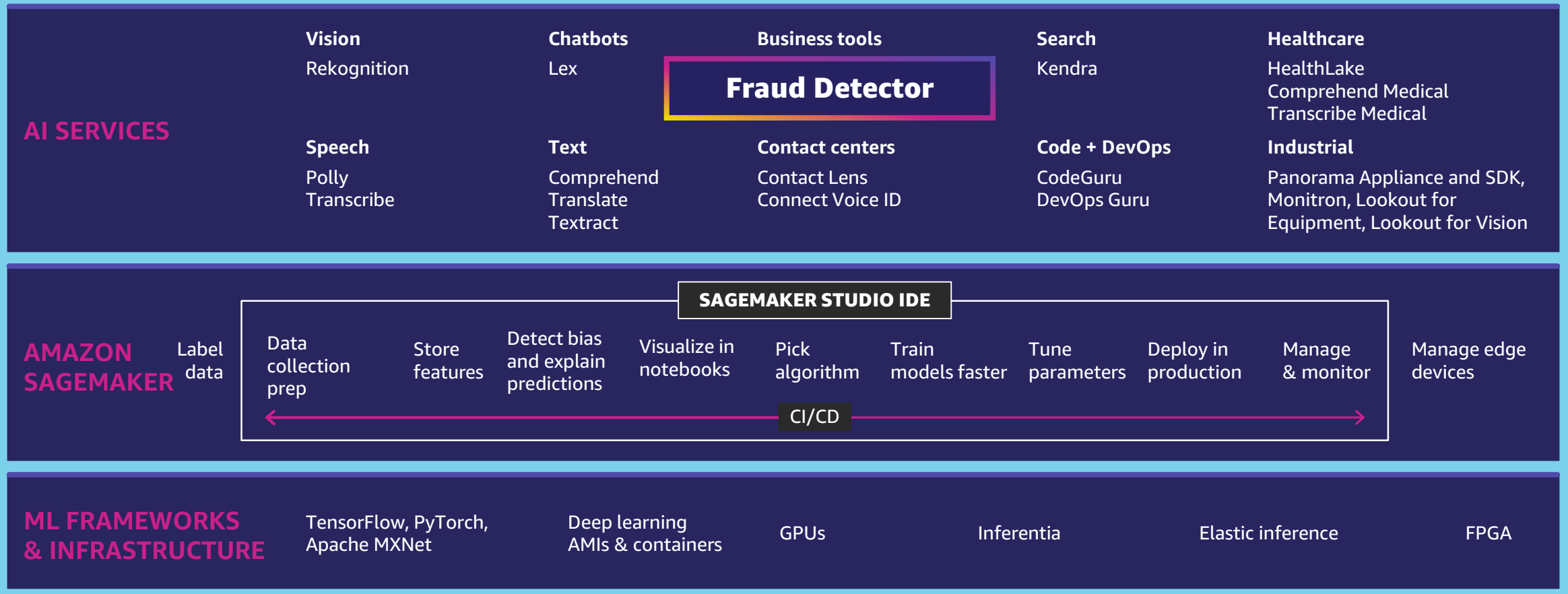
3b. Friction check fails: **block registration**

Account Registration Automatically Blocked



The AWS ML stack

Broadest and most complete set of machine learning capabilities



Amazon Fraud Detector

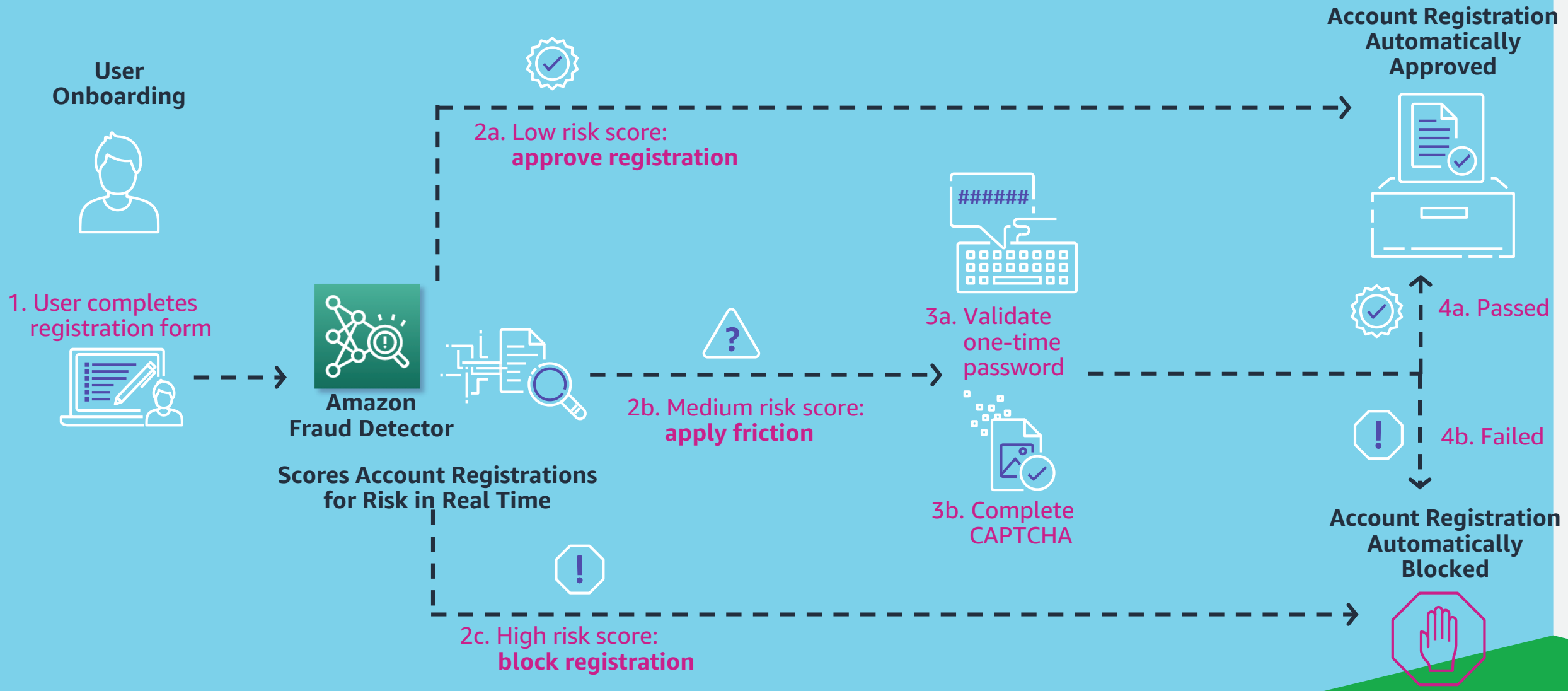
A fully managed service that makes it easy to use machine learning to detect potential online fraud in real-time, at scale

Two Model Types

Online Fraud Insights

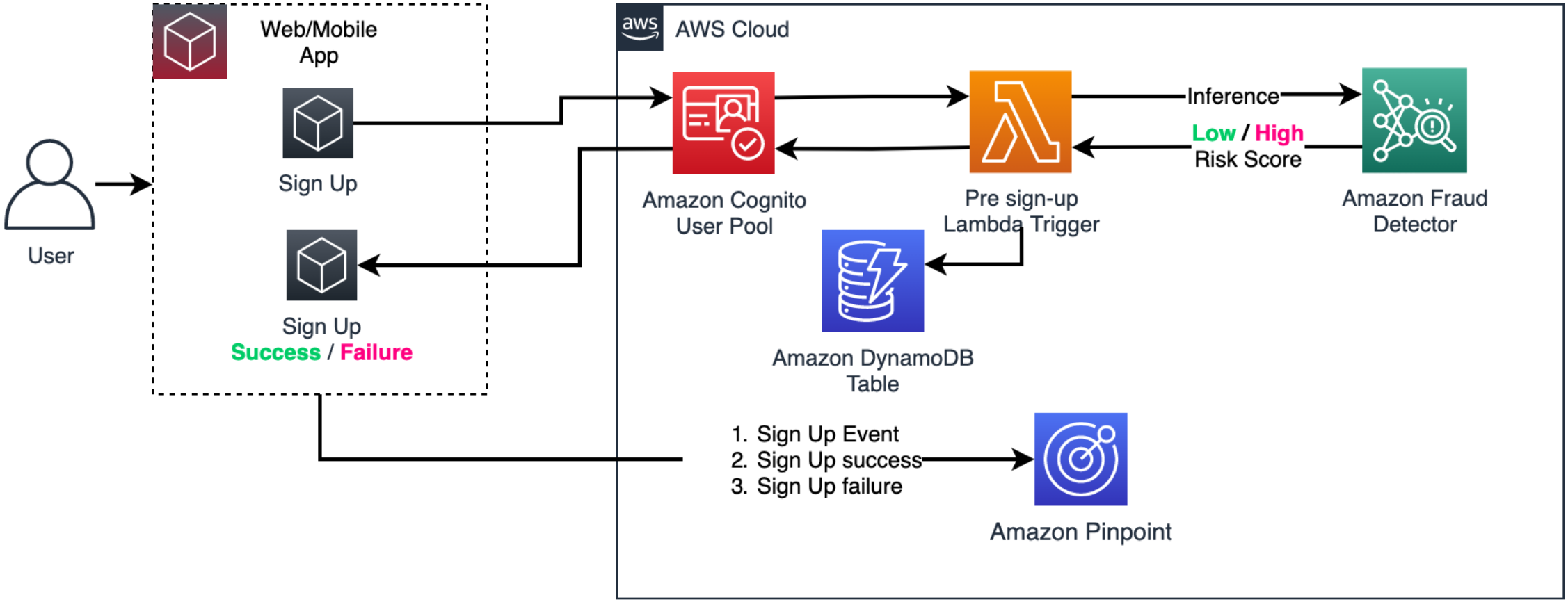
Transaction Fraud Insights

User Onboarding: with Amazon Fraud Detector

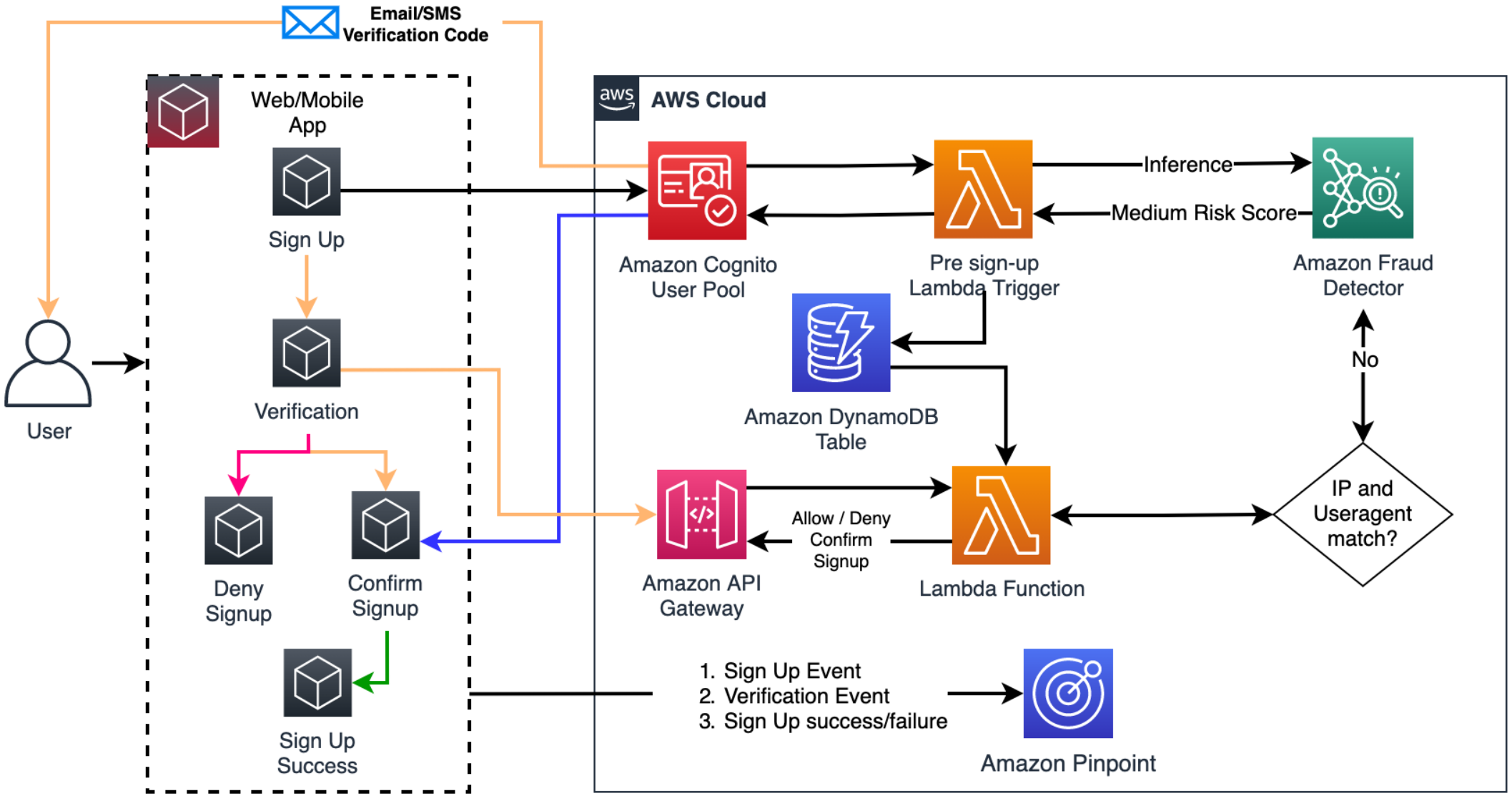


Coupon Shopper Demo

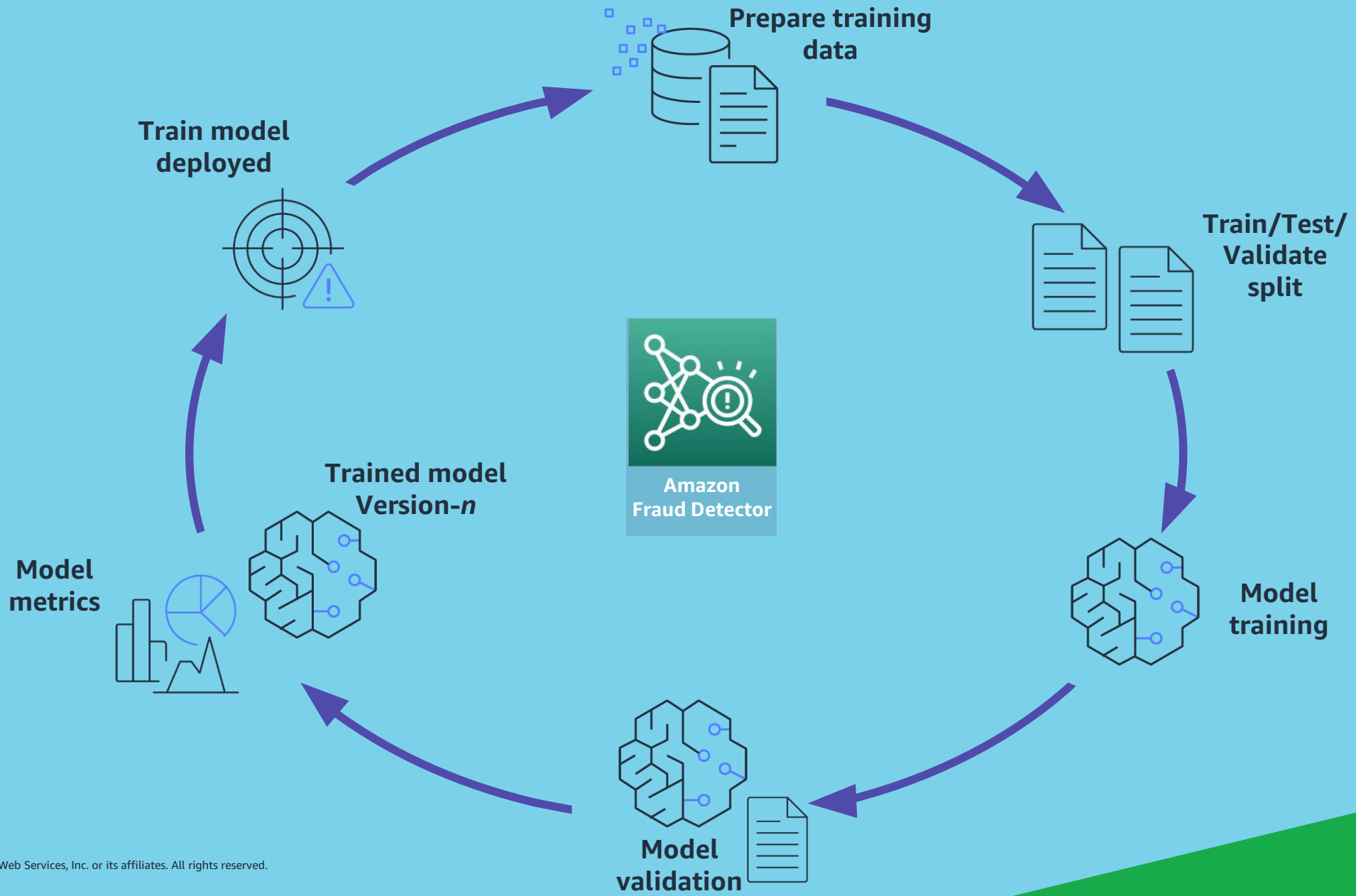
Solution Architecture



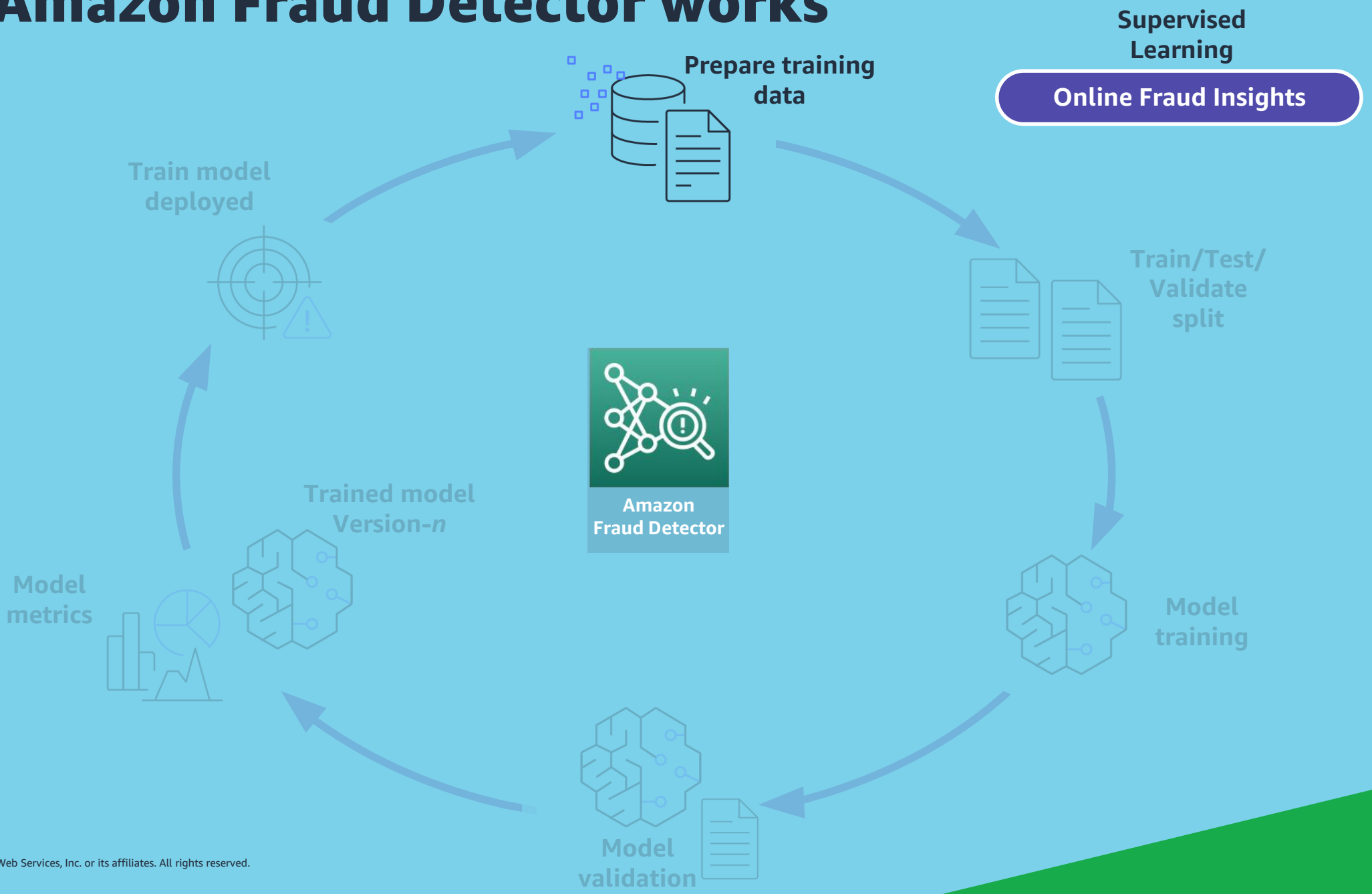
Solution Architecture



How Amazon Fraud Detector works



How Amazon Fraud Detector works



General Training Data Guidelines

At least 2 variables required (max 100)

Metadata **Metadata**

At least
10K total examples
400 fraud examples

EVENT_TIMESTAMP	Variable 1	Variable 2	Variable N	EVENT_LABEL
9/10/2021 11:05	…	…	…	Legit / 0
9/10/2021 19:34	…	…	…	Legit / 0
9/10/2021 20:29	…	…	…	Fraud / 1
…	…	…	…	…

- **Data must be in CSV format – store in Amazon S3 bucket or stream directly.**
- **2 required Event Metadata variables for *Online Fraud Insights*: EVENT_TIMESTAMP and EVENT_LABEL**
- **Recommend at least 3-6 months of data for training the model**
- **Too many NULLs and missing values in a column means it won't be used**

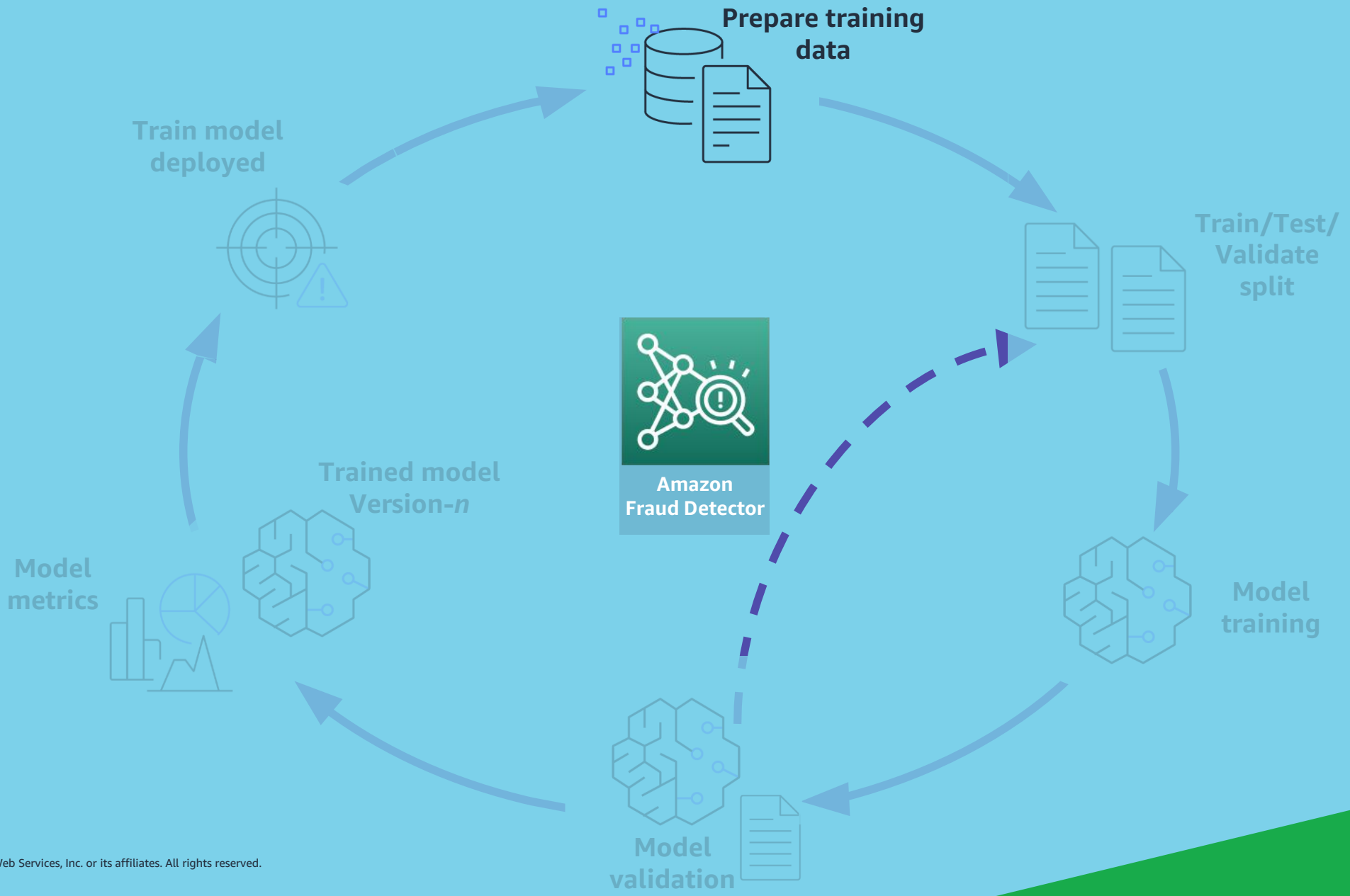
Sample Data

Mandatory variables

6 Additional variables

EVENT_TIMESTAMP	EVENT_LABEL	IP_ADDRESS	EMAIL	PHONE	ADDRESS	STATE	POSTAL
4/29/19 8:34	fraud	192.94.136.134	synth_crossdaniel@gmail.com	(555)715 - 6254	43424 Davis Inlet Apt. 764	NY	32155
4/29/19 8:41	fraud	192.31.192.222	synth_wilsonjustin@yahoo.com	(555)740 - 7222	55403 Jackson Hill	ND	32843
4/29/19 8:43	fraud	192.174.61.17	synth_julie09@hotmail.com	(555)514 - 6454	68642 Leslie Rest Suite 087	NY	34109
4/29/19 8:48	legit	192.88.115.133	synth_max43@hotmail.com	(555)303 - 9960	403 Linda Parkway	KS	33965
4/29/19 8:51	legit	73.60.135.213	synth_billy51@gmail.com	(555)992 - 9688	2566 Henderson Creek Apt. 138	LA	32156
4/29/19 9:28	legit	198.51.101.133	synth_qharris@hotmail.com	(555)672 - 1421	7295 Peterson Ports	ND	33090
4/29/19 9:30	legit	181.20.252.125	synth_john02@yahoo.com	(555)463 - 8749	2404 Huff Pine Apt. 018	MD	34109
…	…	…	…	…	…	…	…
5/2/19 7:36	legit	123.143.40.137	synth_ashley90@hotmail.com	(555)200 - 9565	64680 Solomon Causeway	LA	32678
5/2/19 7:38	legit	192.88.107.202	synth_barrettashley@hotmail.com	(555)562 - 3998	4556 Jordan River Suite 917	AK	32967
5/2/19 7:47	fraud	192.175.40.168	synth_colemariah@gmail.com	(555)491 - 8229	046 Joel Turnpike Suite 872	CA	34205

How Amazon Fraud Detector works



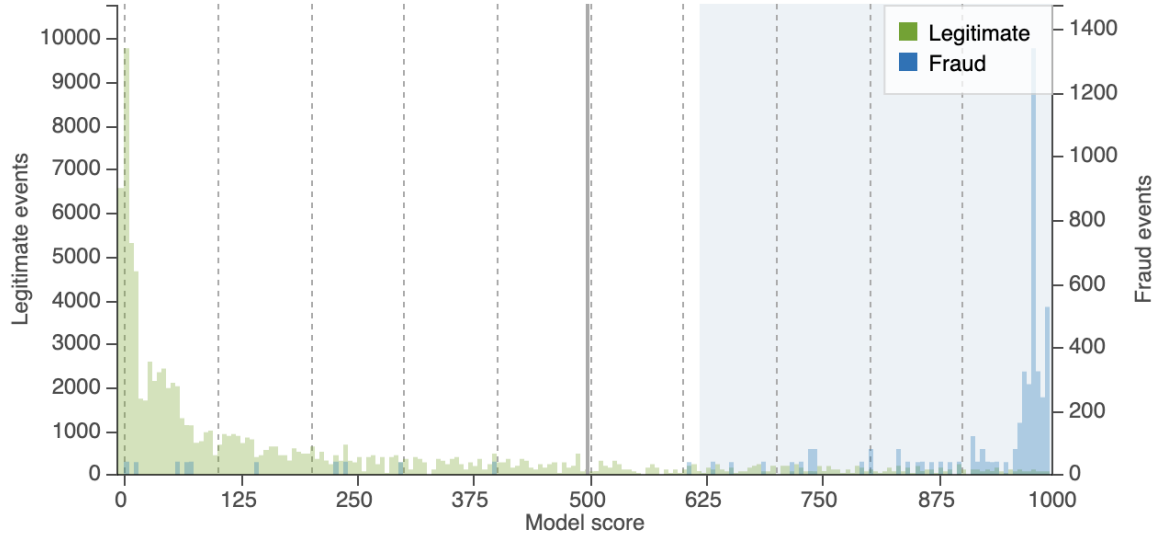
Assessing Model Performance

Model performance AUC 0.95

Charts | Table

Score distribution

By writing a rule using a model score threshold of **500**, you will succeed in catching **91.9%** of all fraudulent events (TPR) while accepting a risk that **12.7%** of legitimate events are incorrectly labeled as fraud (FPR).



Click anywhere on the chart above to select a model threshold score and determine the TPR and FPR.

Confusion matrix

For the selected model score threshold, the confusion matrix represents the expected outcome given 100,000 sample events (95,012 legitimate, 4,988 fraud).

Actual \ Predicted	Fraud	Legitimate	
Fraud	True positive **4582**	False negative **406**	TPR **91.9%**
Legitimate	False positive **12084**	True negative **82928**	FPR **12.7%**

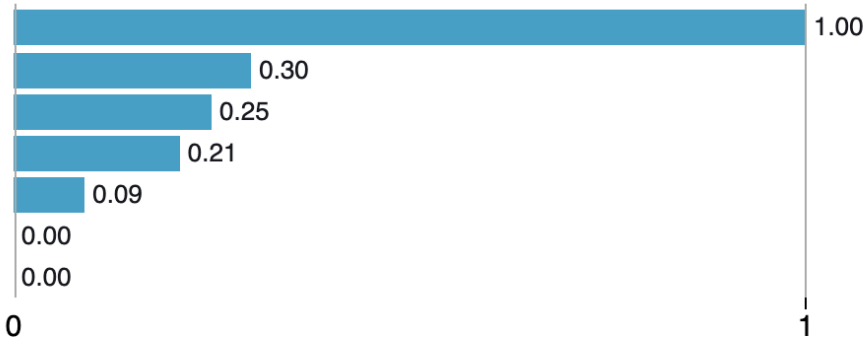
Numbers are based on a sample of 100,000 events.

Assessing Model Performance

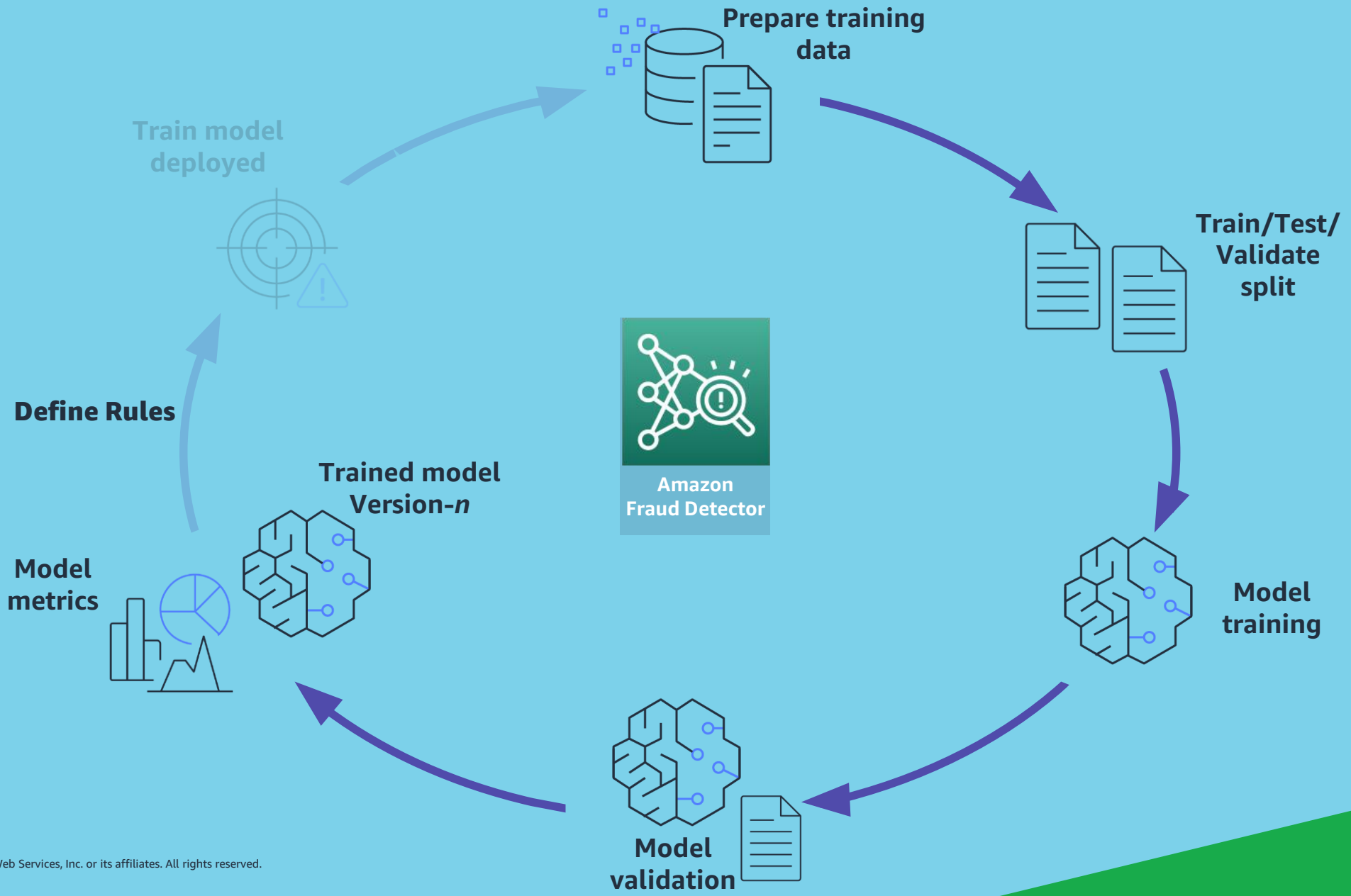
Model variable importance - *new*

Variable Importance gives you an understanding of how each variable is contributing to your model's performance. The chart below lists model input variables in the order of their importance to the model, indicated by the number. A variable with a much higher number relative to the rest could indicate that the model might be overfitting on it, while variables with relatively lowest numbers could just be noise. For details, refer to documentation

Variable name ▲	Variable type ▲	Importance value ▲
billing_state	Billing Address: State or Province	1.00
billing_postal	Billing Addres: Zip Code	0.30
user_agent	User agent	0.25
email_address	Email Address	0.21
ip_address	IP Address	0.09
billing_address	Billing Address: Address Line 1	0.00
phone_number	Phone number	0.00

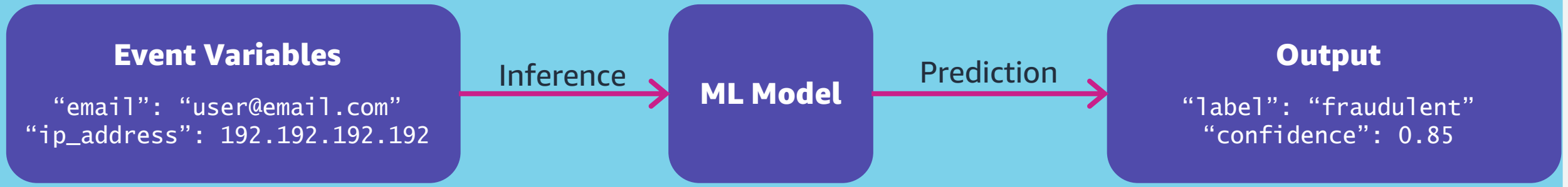


How Amazon Fraud Detector works



Defining Rules

A typical ML model outcome



Confidence score between 0 and 1

Defining Rules

Version details

Model name	Model type	Status	Date created
sample_fraud_detection_model	Online Fraud Insights	Active	Sun, Sep 19, 2021, 10:27 AM PDT

Performance (AUC)	Event type	Output variable	ARN
0.95	sample_registration	sample_fraud_detection_model_insightscore	arn:aws:frauddetector:us-east-2:576090071164:model-version/ONLINE_FRAUD_INSIGHTS/sample_fraud_detection_model/1.0

Model Score	Estimated FPR
975	0.05%
950	1%
900	2%
860	3%
775	5%
700	7%
600	10%

Rule language
`$variable < 100`
`$variable in [5, 10, 25, 100]`
`$variable != "US"`
`$variable == 1000`

Associated rules (3)

The rules listed below have been added to the sample_detector detector

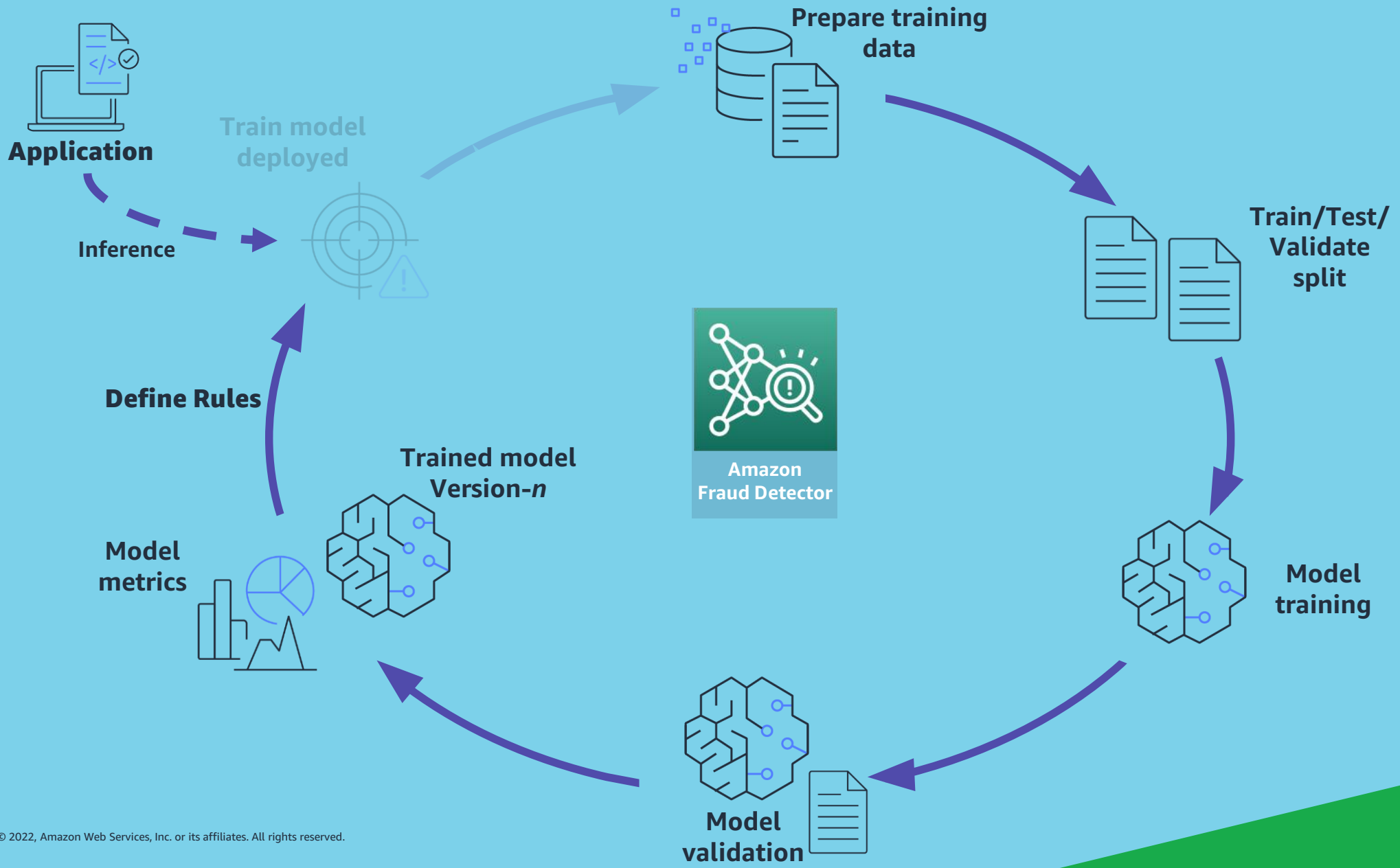
Create rule

Find rules

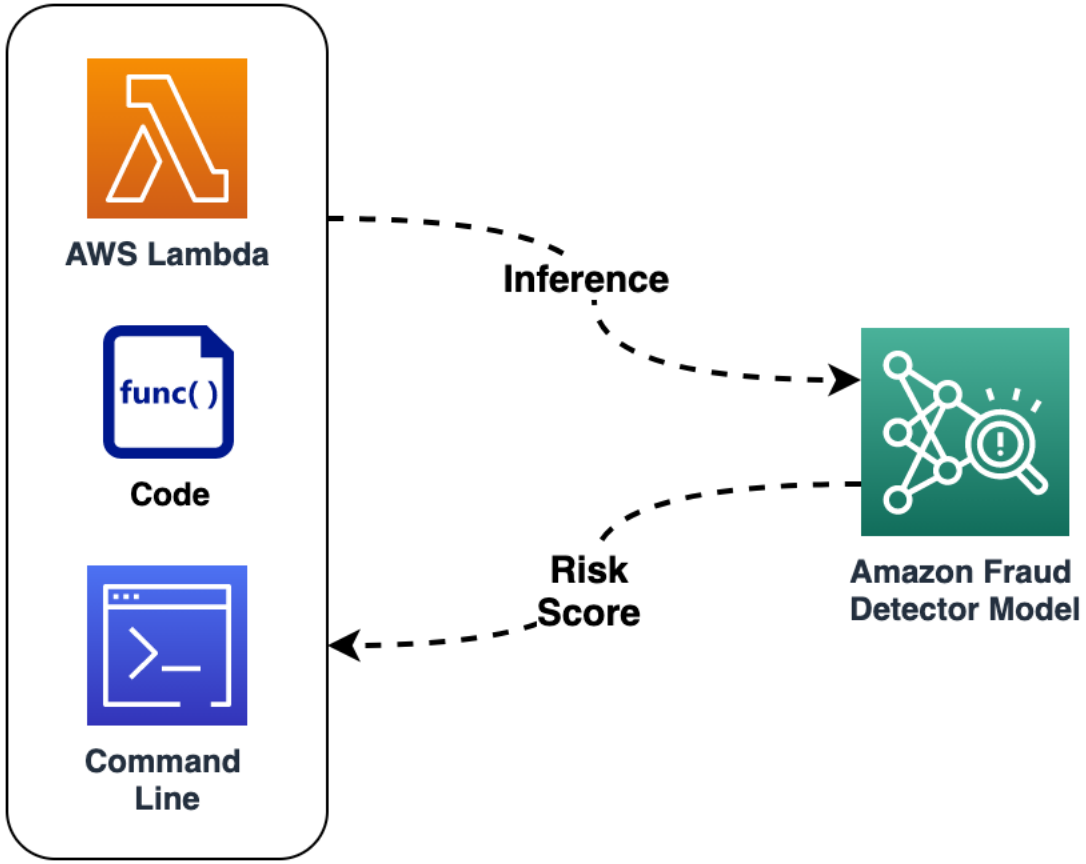
1

Rule name	Version	Description	Expression	Outcomes
high_fraud_risk	2	This rule captures events with a high ML model score	$sample_fraud_detection_model_insightscore > 900	review_customer
low_fraud_risk	1	Low risk of fraud	$sample_fraud_detection_model_insightscore <= 700	approve
medium_fraud_risk	1	Medium Fraud Risk	$sample_fraud_detection_model_insightscore <= 900 and $sample_fraud_detection_model_insightscore > 700	review

How Amazon Fraud Detector works



Generating predictions with a deployed model



Generating predictions with a deployed model

AWS SDK

Python, NodeJS, Java, C++, .Net etc.

GetEventPrediction API

Python example

```python
response = client.get_event_prediction(
    detectorId='sample_detector',
    detectorVersionId='1',
    eventId='20220207',
    eventTypeName='user_registration',
    entities=[
        {
            'entityType': '20220207',
            'entityId': 'customer_entity'
        },
    ],
    eventTimestamp='20220207120903',
    eventVariables={
        'email_address': 'user@email.com',
        'ip_address': '192.192.192.192'
    }
)
```

AWS CLI

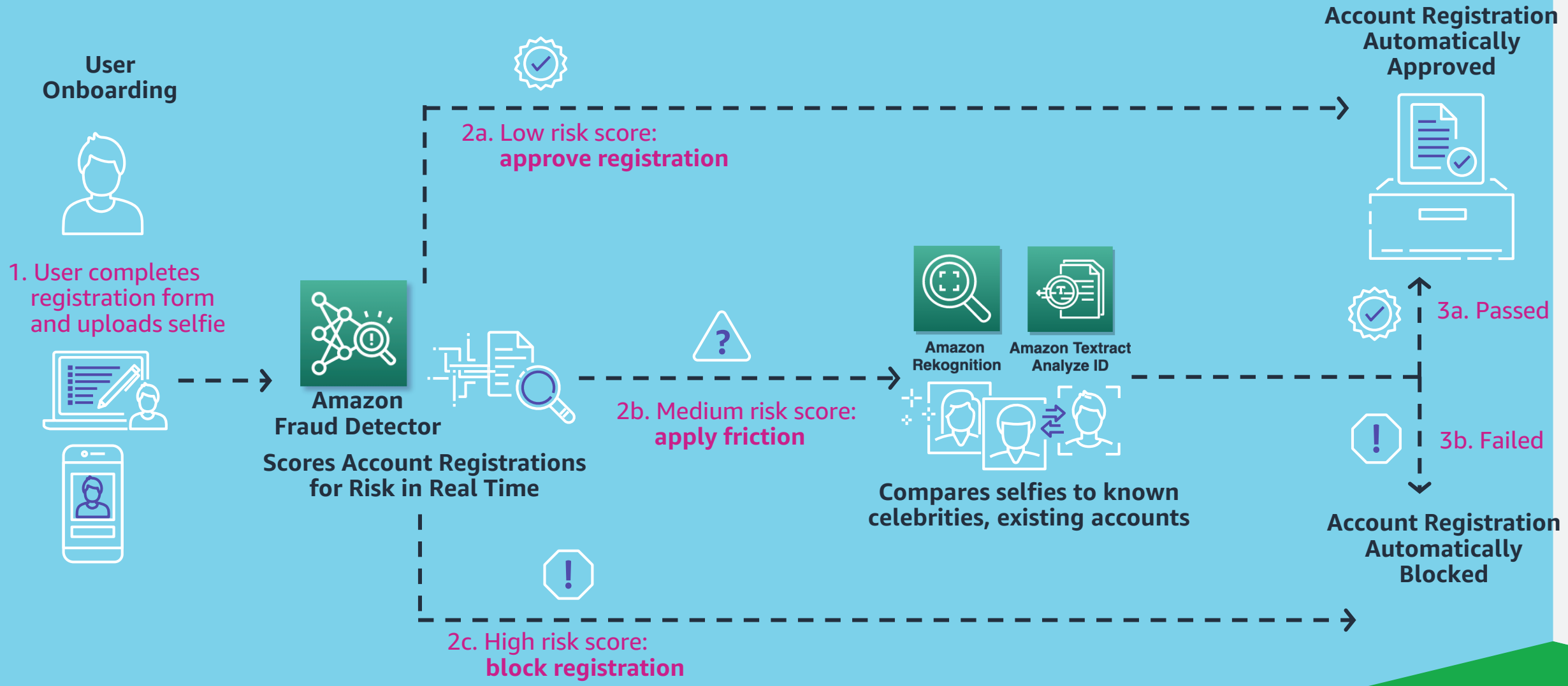
```
aws frauddetector get-event-prediction --detector-id sample_detector \
--detector-version-id 1 \
--event-id 20220207 \
--event-type-name user_registration \
--entities entityType=20220207,entityId=customer_entity
--event-timestamp 20220207120903
--event-variables { \
        "email_address": "user@email.com", \
        "ip_address": "192.192.192.192" \
    }
```

Sample Prediction

Model Score = 984

High Fraud Risk

Fraud Detection
Combined with facial recognition & ID verification



Wuzzon

Wuzzon is an app marketing agency that is committed to aid app owners grow and activate their user acquisition by helping them set up a complete marketing plan, including user acquisition, app store optimization and re-engagement strategies.

By implementing Amazon Fraud Detector into the WuzzTrack system, Wuzzon now has a much more robust and reliable fraud detection solution, which can also detect more novel fraud techniques. The implementation was quick and easy and the results were even better than initially hoped. In some extreme cases there was a **decrease in false positives of up to 43%** (when compared to the previous rule-based solutions), while for other sources, **the true positive rate increased by 11-14%**.

Justin Westerveld

CTO

What Comes Next?

Try It Yourself

Blog:

Prevent fake account sign-ups in real time with AI using Amazon Fraud Detector

https://a.co/4cgDed2

Talk With Us

Deep Dive:

Contact your AWS Account Team to get help with your specific situation

Mention this session!

Learn More

For documentation, code samples and pricing visit:

aws.amazon.com/fraud-detector

Thank you!

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