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METRICS THAT MATTER - DRIVING APPSEC SUCCESS WITH DATA-DRIVEN INSIGHTS

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Agenda

What we'll cover today



1. Why Metrics Matter

The origins of measurement in management and the pitfalls of poorly chosen metrics



2. Understanding AppSec Metrics

Categories, types and examples of meaningful appsec metrics



3. Gathering the Right Data

Explore data sources from SAST, DAST, Pentest, Bug Bounties, etc.



4. Calculating Risks & Scorecards

How to combine multiple metrics into one weighted score



5. Visualising Metrics

Make metrics useful for engineers and executives with effective dashboards



6. Driving Program Impact

Make metrics actionable - training, coverage, and more



Who am i?

Short introduction

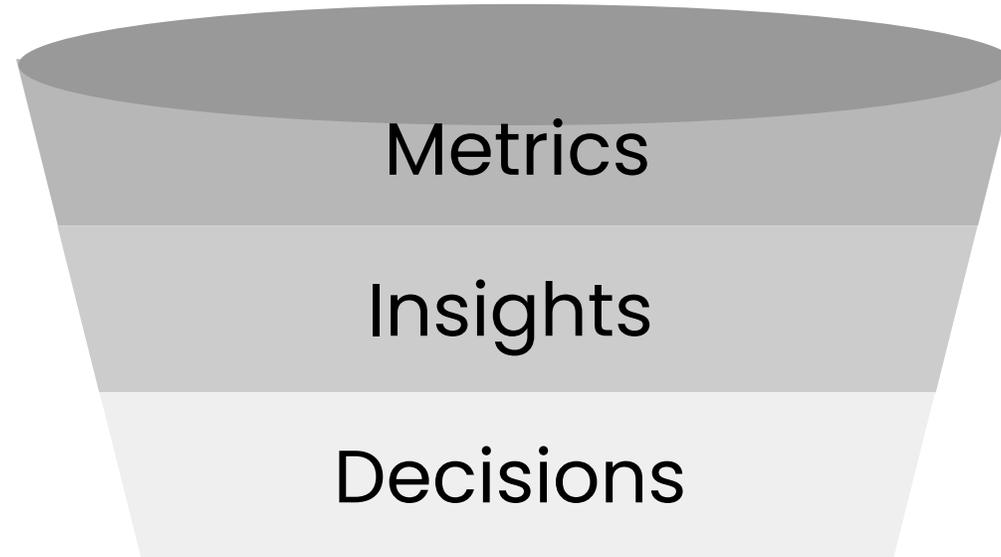
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Why Metrics Matter

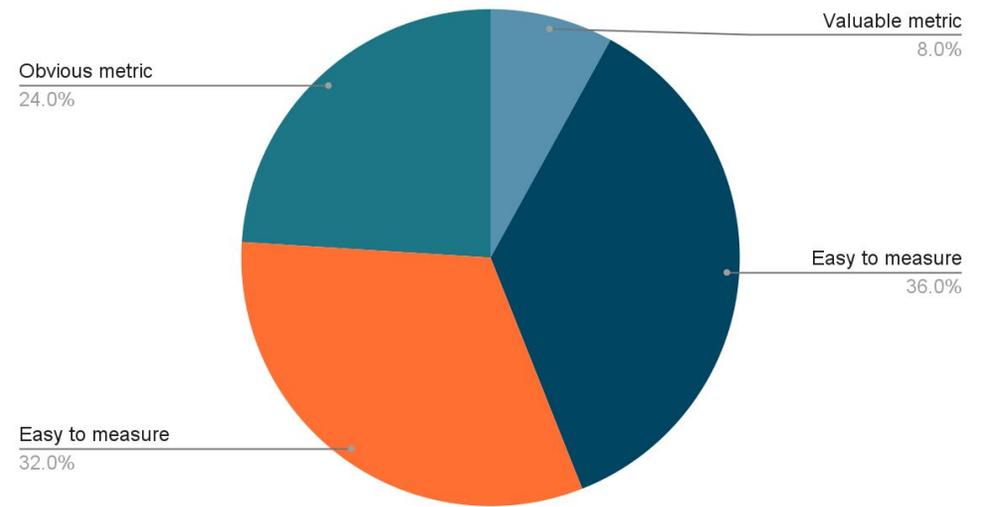
Linking metrics to security impact





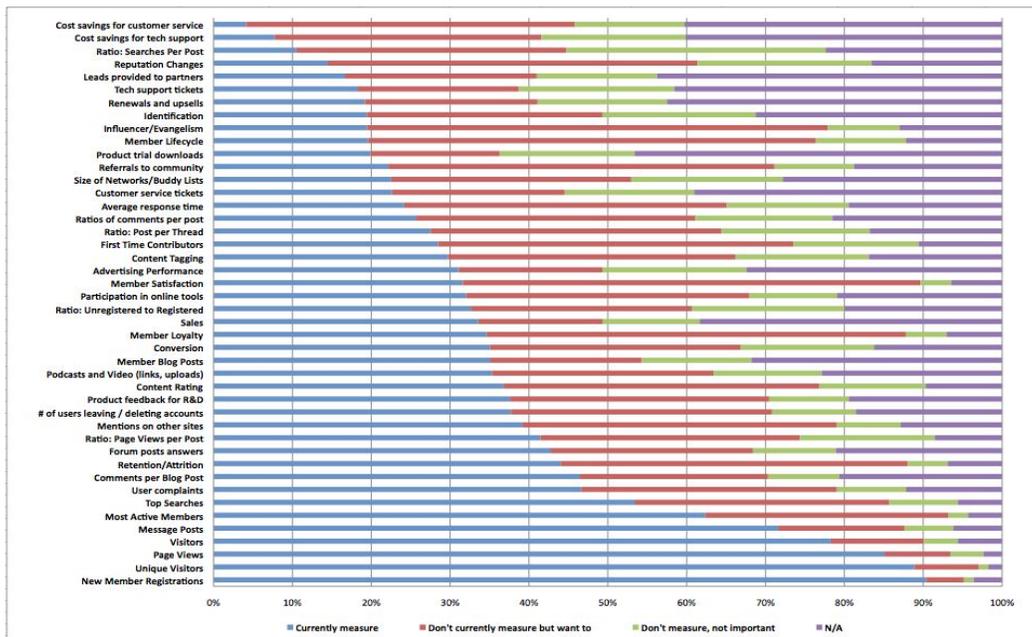
"Civil Engineering Training" by Georgia National Guard CC BY 2.0.

Types of metrics



The metric maze

When more data means less clarity



Goodhart's Law
“When a measure becomes a target, it ceases to be a good measure”

*Standards give us direction –
not rules, but reason*



"Wollongong Breakwater Lighthouse" by Bernard Spragg CC0 1.0

Types of AppSec Metrics



Operational vs. Technical / Quantitative vs. Qualitative

Operational

Technical

Quantitative

Count or percentage of applications have completed penetration testing in the last "n" months

Count or percentage of vulnerabilities by weakness

Qualitative

Maturity of AppSec practices

Security Culture in Engineering Teams

What Makes a Good Metric?

Five traits for trust and actionability

 **Relevant**

 **Timely**

 **Measurable**

 **Repeatable**

 **Understandable**

From Trait to Metric

Examples of metrics that get it right

MTTD/MTTR

**Open
Vulnerabilities per
Severity**

**Number of
vulnerable
dependencies**

**Average bug
bounty reports per
day**

**Number of security
trained engineers**

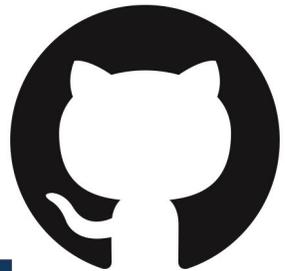


Sources of truth

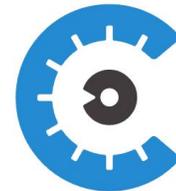
Where your data comes from



INTIGRITY



hackerone



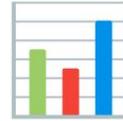
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Practical Metrics



Choosing the right signals

Data Source

LMS

Ticketing System

Scanners

Engagements

Metric

Training Coverage

MTTR

MTTD

Last Date

What it tells us

Developer Maturity

Responsiveness

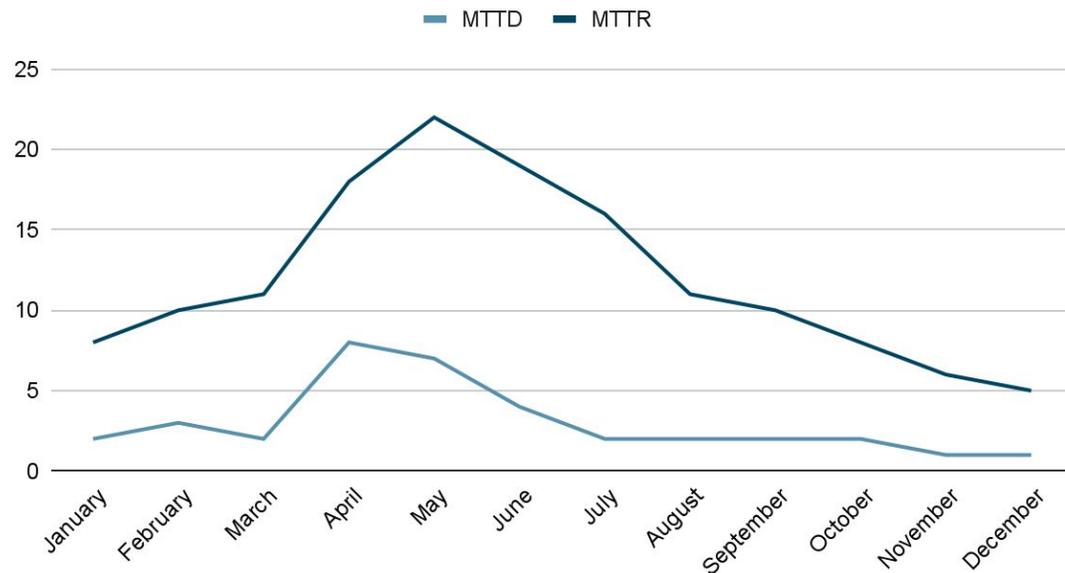
Detection capabilities

How to plan engagements

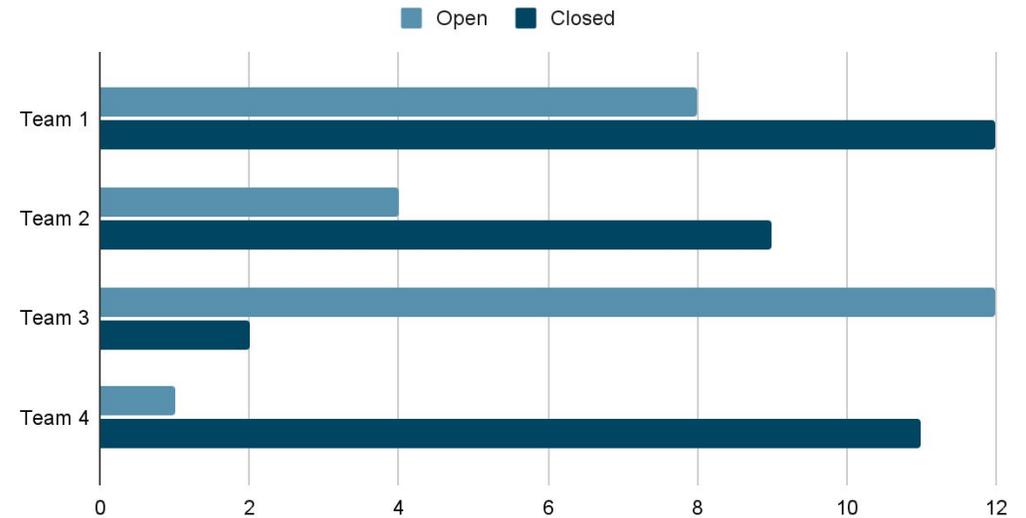
Getting Insights from Details

Interpreting MTTR/MTTD & open vs. closed trends

Mean Time To Detection / Mean Time To Resolution

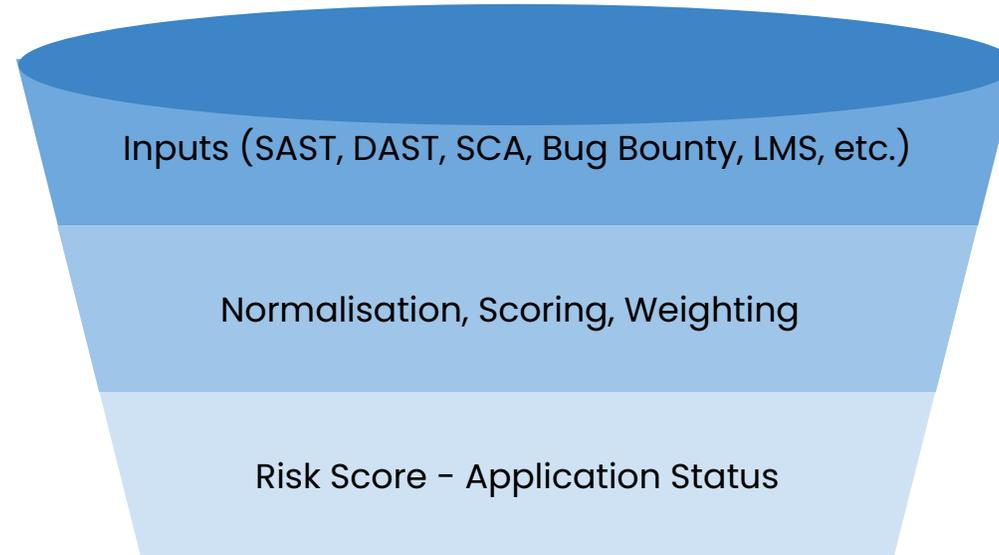


Open/Closed SAST findings



From Metrics to Meaning

Building the insight pipeline



Normalizing the Noise

Turning messy raw signal into comparable signal

Raw Metric

SAST findings

SCA findings

DAST findings

Days to Close High Bug Bounty Issues

Recency of Last Pentest

Count of Open Critical Vulns

Normalized Metric

SAST Severity Score per KLoC

SCA Severity Score per KLoC

DAST Severity Score per scanned URL

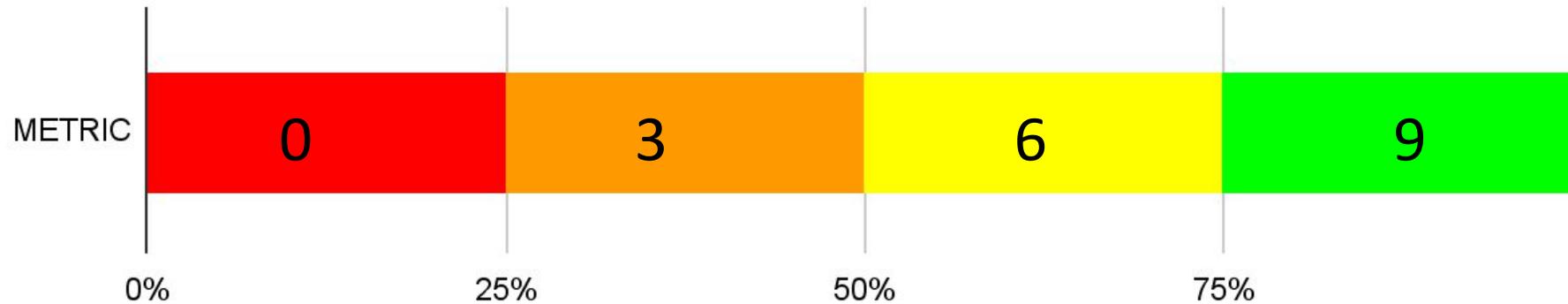
Bug Bounty Responsiveness

Pentest Freshness Score

Open Critical Risk Density

Normalizing the Noise 1 2 3 4

Defining bands



```
# Normalisation helpers
def score_sast(row):
    raw_score = (100 * row["critical"] + 50 * row["high"] +
                10 * row["medium"] + row["low"]) / max(row["loc"], 1000) * 1000
    if raw_score <= 10:
        return 9
    elif 10 < raw_score <= 25:
        return 6
    elif 25 < raw_score <= 50:
        return 3
    return 0
```

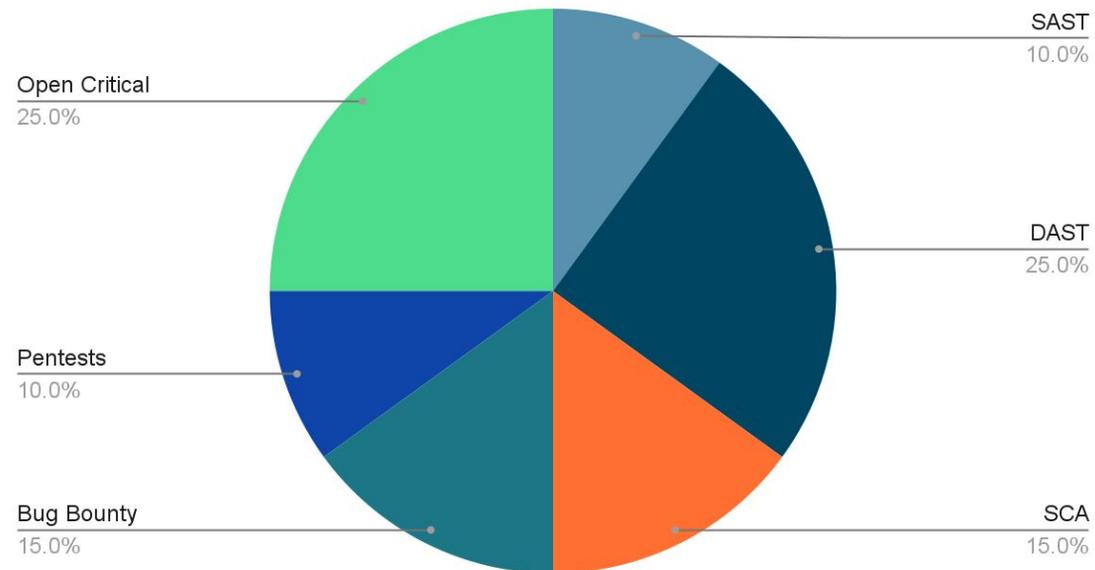
```
def score_bug_bounty(row):
    raw_score = row["mtrr_high"]

    # Normalized Bug Bounty score
    if raw_score <= 7:
        return 9
    elif 7 < raw_score <= 14:
        return 6
    elif 14 < raw_score <= 30:
        return 3
    elif raw_score > 30:
        return 0
```

Scoring Strategy Overview

Weighting what matters most

Weighting



Risk Score Formula



Combining normalized scores into a unified rating

$$\text{Risk} = \sum(\text{Normalized}_i \times \text{Weight}_i)$$

Scorecard Example



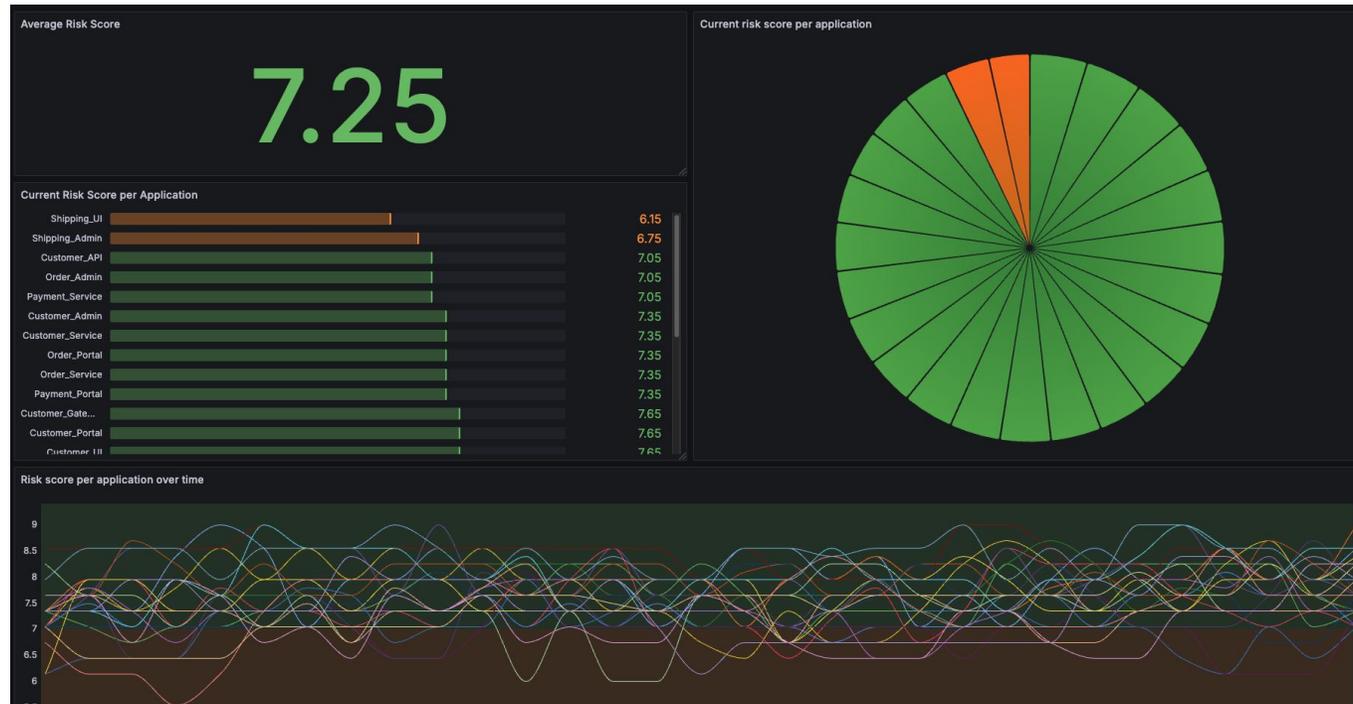
Sample risk calculation for one application

Source	Metric	Normalised score	Weight factor	Score	Status
SAST	SAST Severity Score per KLoC	6	10%	0.6	● Amber
SCA	SCA Severity Score per KLoC	3	15%	0.45	● Red
DAST	DAST Severity Score per scanned URL	6	25%	1.5	● Amber
Bug Bounty	Bug Bounty Responsiveness	9	15%	1.35	● Green
Penetration Test	Pentest Freshness Score	0	10%	0	● Red
Any Source	Open Critical Risk Density	6	25%	1.5	● Amber
Total				5.4	● Amber

Visualizing Risk



What could it look like?



Mapping Metrics to Actions



CWE mapping to targeted training

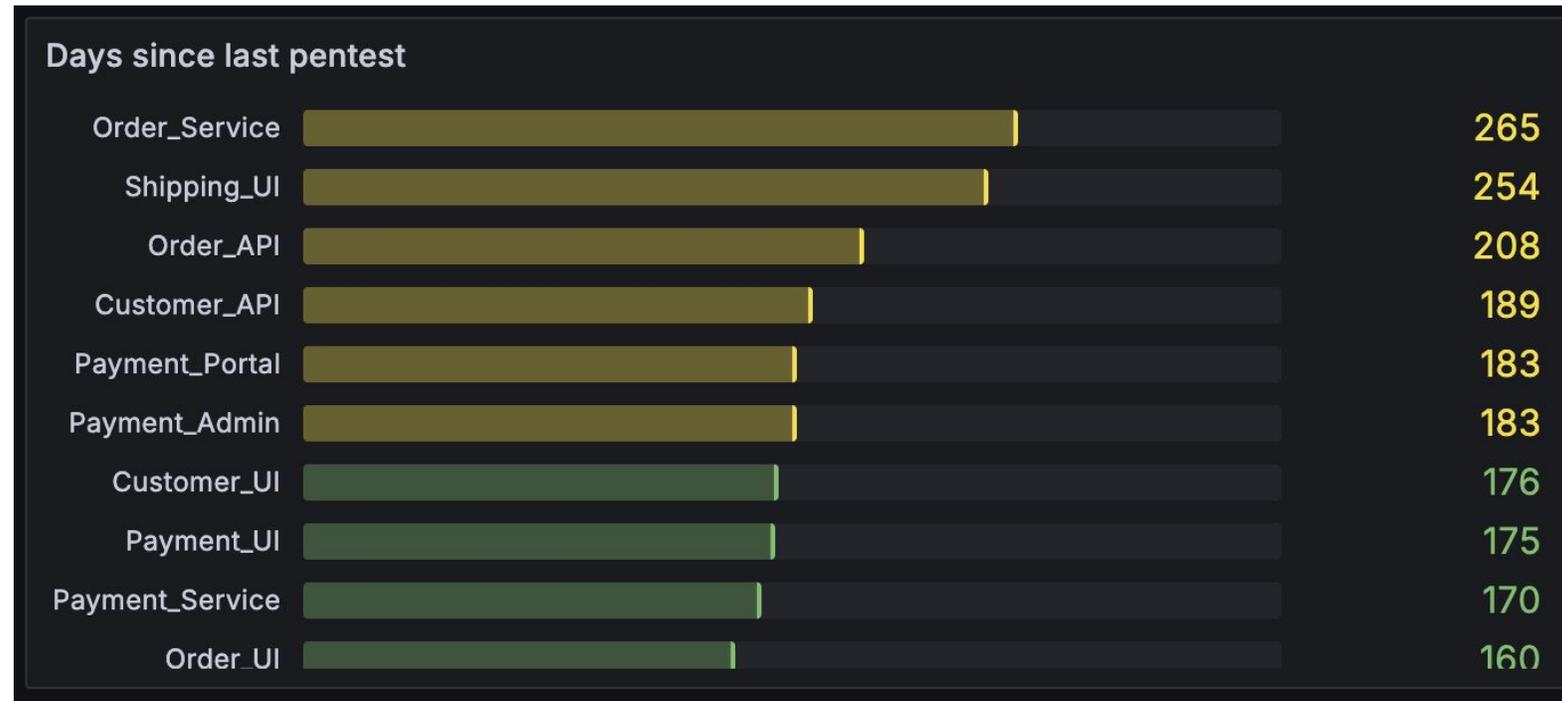
Top 10 Vulnerabilities per CWE



Mapping Metrics to Actions

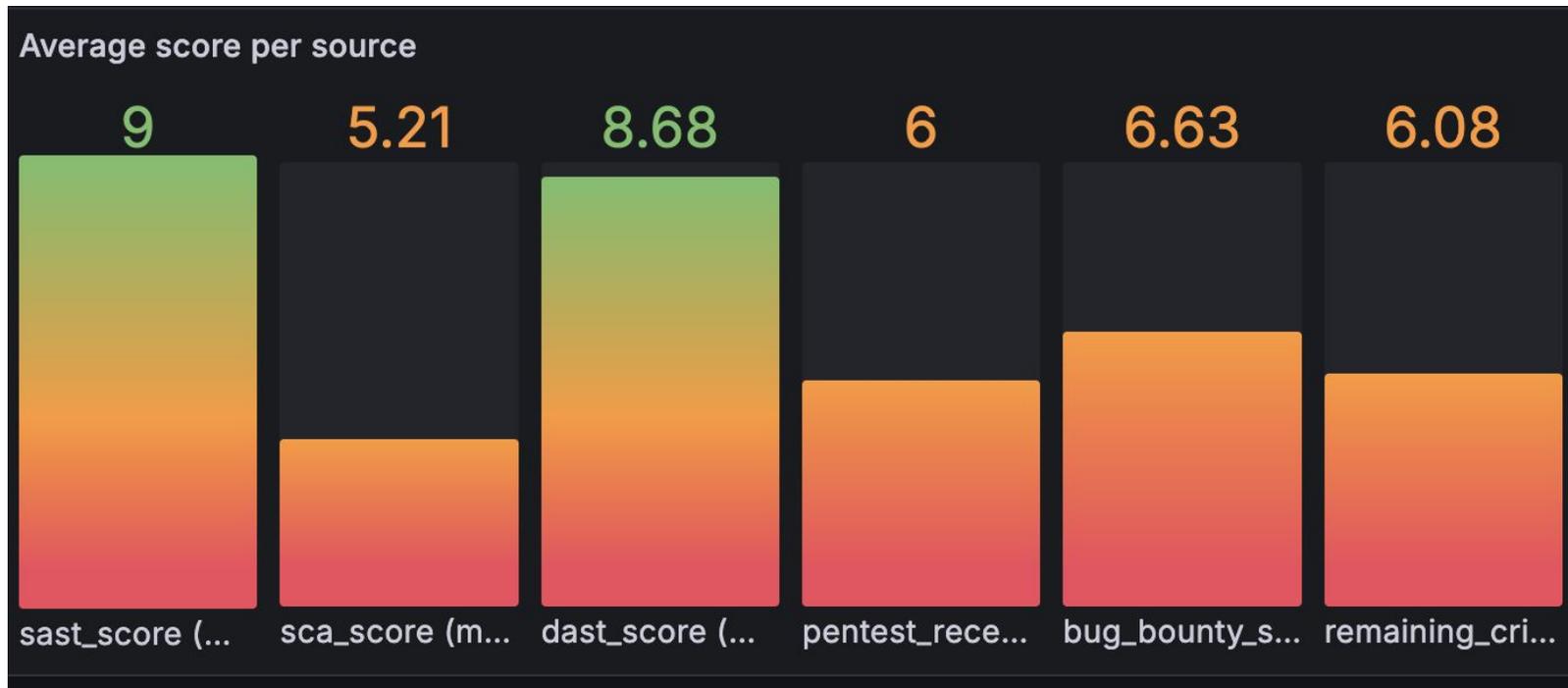


Help in planning



Mapping Metrics to Actions

Signal per score



Metrics as Dialogue

Start conversations, not end them

“Metrics should start conversations, not end them”

Recap & Key Takeaways

What to remember

Measure with purpose

Normalize intelligently

**Make metrics
actionable**

Thank You ✨

Stay in touch!



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THANK YOU!