How Program Managers can exploit DevSecOps Pipeline of Pipelines to Inform Data Driven Decisions

NDIA

October 2023 Bill Nichols Julie Cohen

Software Engineering Institute Carnegie Mellon University Pittsburgh, PA 15213



Carnegie Mellon University Software Engineering Institute



Make Program Decisions!

Your SOCOM commander has learned that an adversary has unexpected capabilities.

You must **reprioritize** your **capabilities**.

Whiz-Bang Software for Sensor Fusion, which was added to the roadmap 18 months ago, is now **top** priority with a need of within the next 6 months.

Two capabilities that were to be the next priority, due in 3 and 9 months, are critical and must be completed as close to the original due dates as feasible.

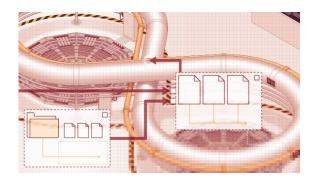
You need to know the following:

- How can the two next-priority capabilities and the new, highest priority capability be delivered without affecting the staffing?
- To deliver the Whiz-Bang capability within six months while maintaining the other two capabilities' schedules, will more teams be required? How many? At what cost?



How do you make informed, defensible decisions?

Agenda



Making defensible decisions with data Automating data and analysis with ACE/PoPs Indicators for Program Managers Observations and Lessons Call to Action

ACE/PoPs Overview

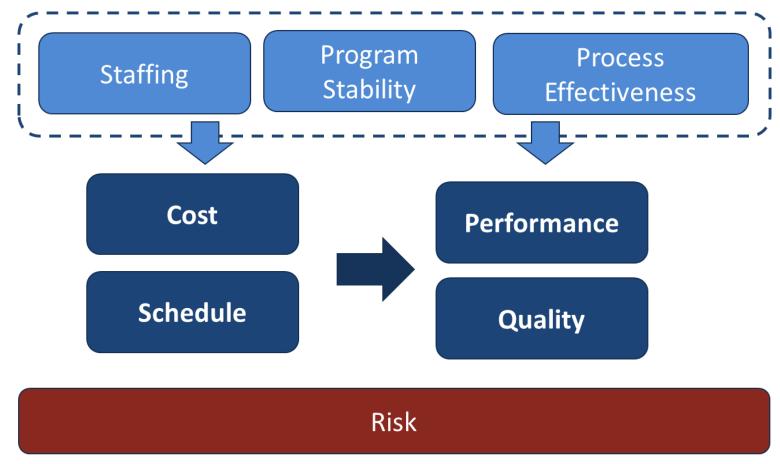
Software Engineering Institute Carnegie Mellon University Pittsburgh, PA 15213



Carnegie Mellon University Software Engineering Institute

Start by getting the information you need

Information needs: What are your targets? What can you manage?



7

Automated Continuous Estimation for **Pipelines of Pipelines**

Automation drives continuous integration and delivery of software, but outpaces program control

Ē

To solve this problem: Automate data collection Model DSO systems with **Monte Carlo**, and provide continuous reporting.

- Determine status
- Project future events
- Provide evidence for corrective actions

Goal: Programs using DSO(DevSecOps) have constant access to information needed to monitor and control schedule and cost commitments.

Status and projection models should be available in real time.

Model pipeline and pipeline-of-pipeline systems.

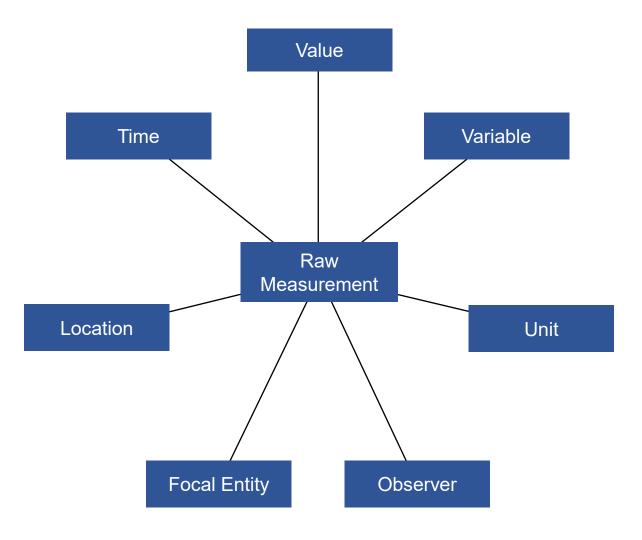
Automate data collection and Program Management Status Reporting for DevSecOps pipelines.

Directly collect data from DevSecOps pipeline tools

- Automate data collection, storage, and reporting
- Correlate data to project outcomes
- Present completion to-date and milestone predictions to Program Management in smart dashboards

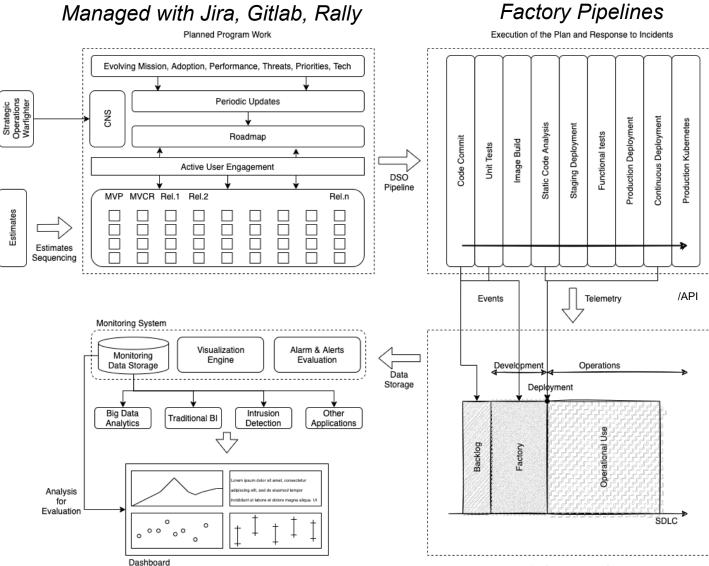
Prototype DSO Measurement

Ę



• Measurement: "A set of observations that reduce uncertainty where the result is expressed as a quantity." -Douglas Hubbard

ACE/PoPs Model: Data Collection Context



Planned work includes the WBS, work packages, work sequencing, and estimates.

Work packages **execute** plan development stages. Tools trigger events (time stamps, package labels).

Data is collected and **transformed** for storage.

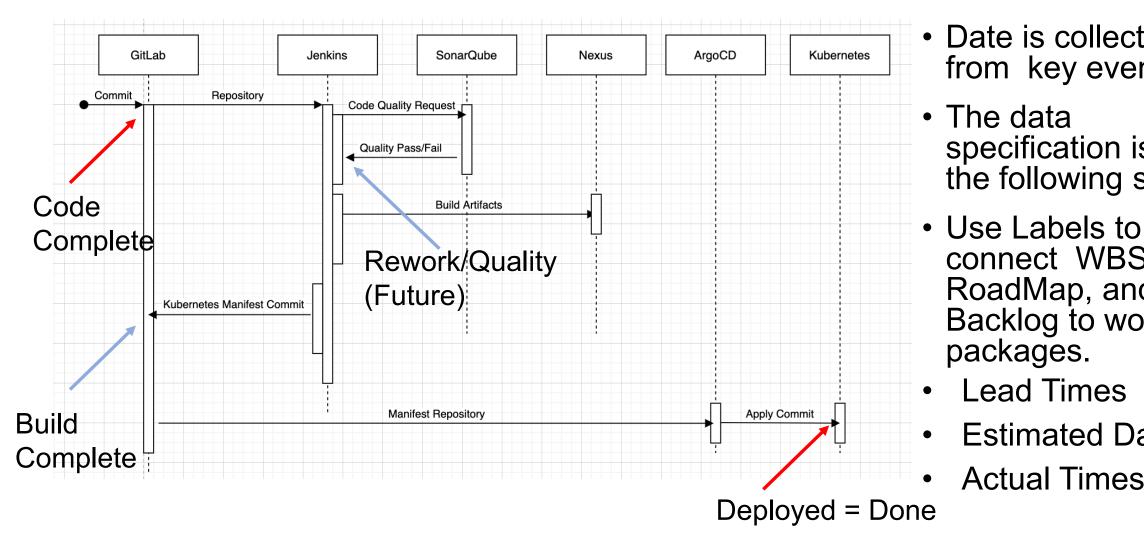
The **warehouse** loads the data and provides the interface for analysis and dashboards.

PSM Context of Data Collection

Track events through the Pipelines

Ę

Extracting metrics https://youtu.be/u96OFTXgr0g



Date is collected

from key events.

specification is on

the following slide.

connect WBS,

RoadMap, and

Lead Times

Actual Times

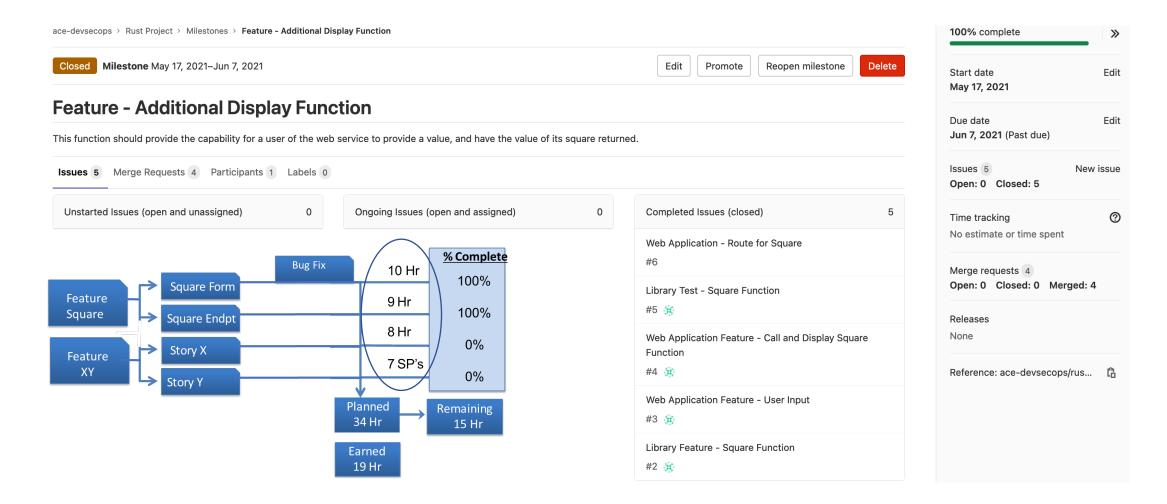
packages.

Backlog to work

Estimated Dates

Track From the Roadmap Through the Pipeline

Work Completion https://youtu.be/X-R1mlZ3sPk



Indicators

Software Engineering Institute Carnegie Mellon University Pittsburgh, PA 15213

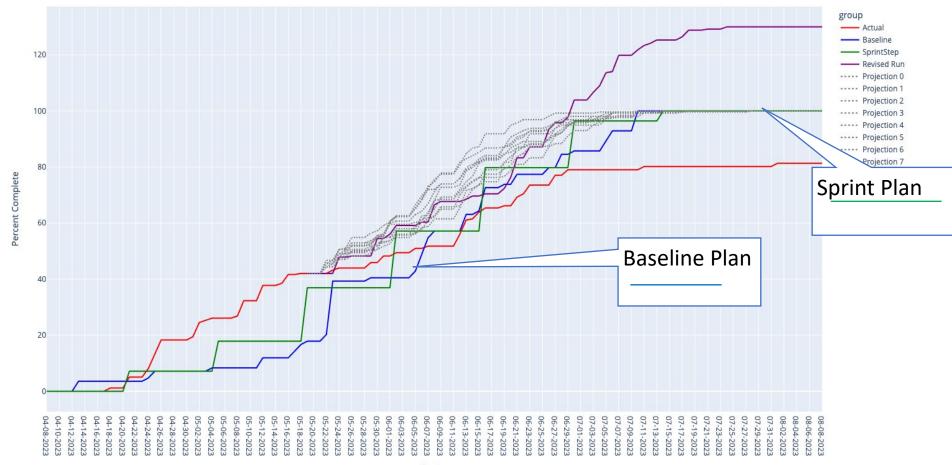


Carnegie Mellon University Software Engineering Institute

[DISTRIBUTION STATEMENT A] Approved for public release and unlimited distribution.

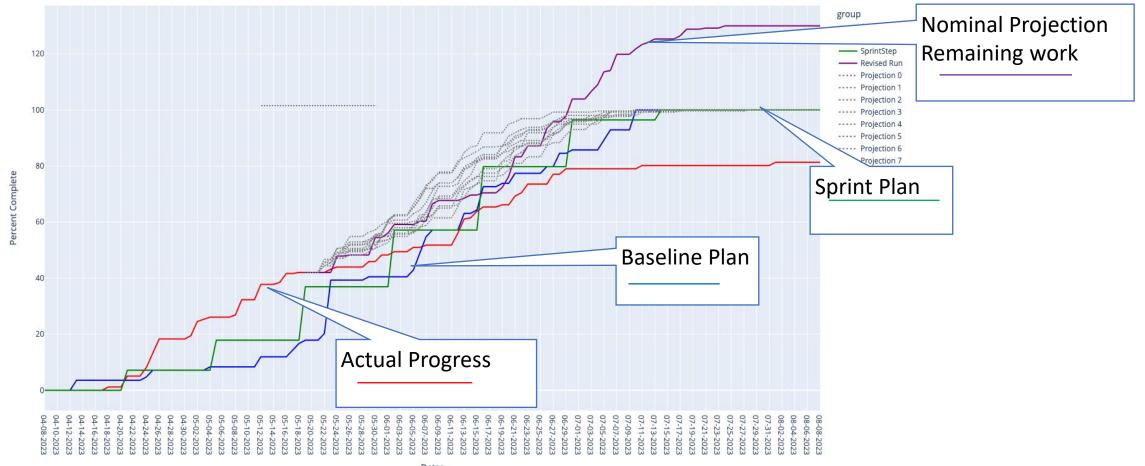
Analysis: What is% Completion, When will we Finish?

Ē



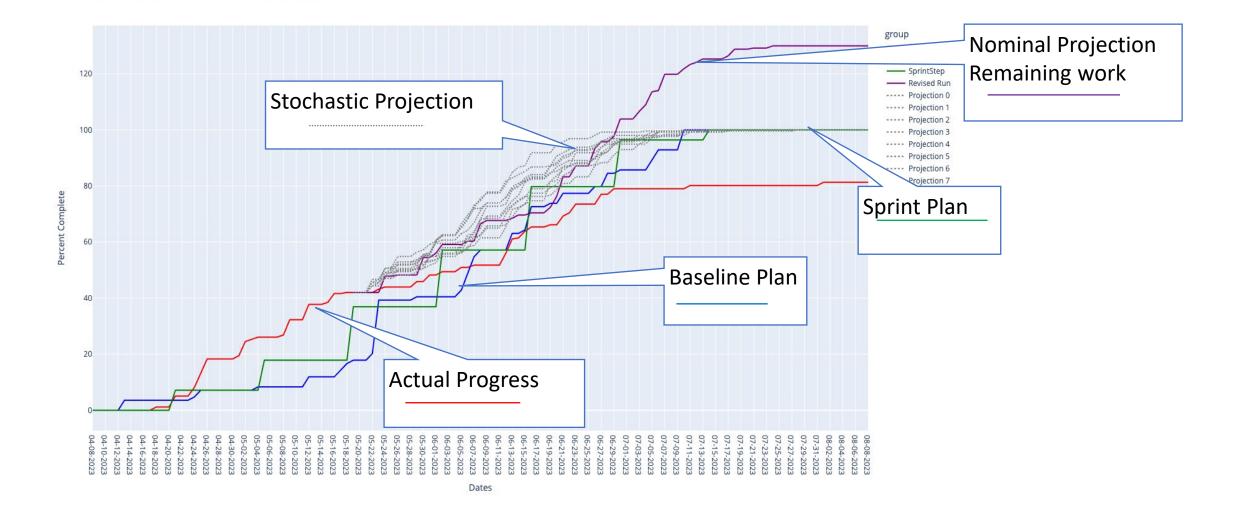
Analysis: What is% Completion, When will we Finish?

Ē



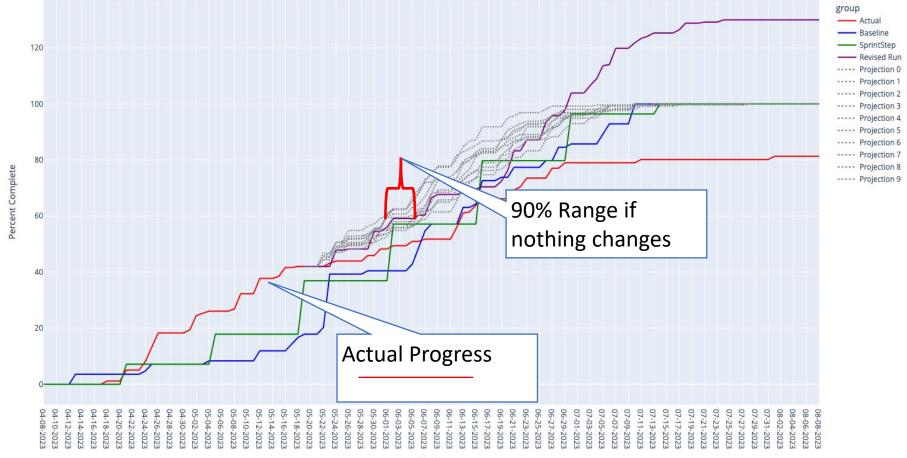
Analysis: What is% Completion, When will we Finish?

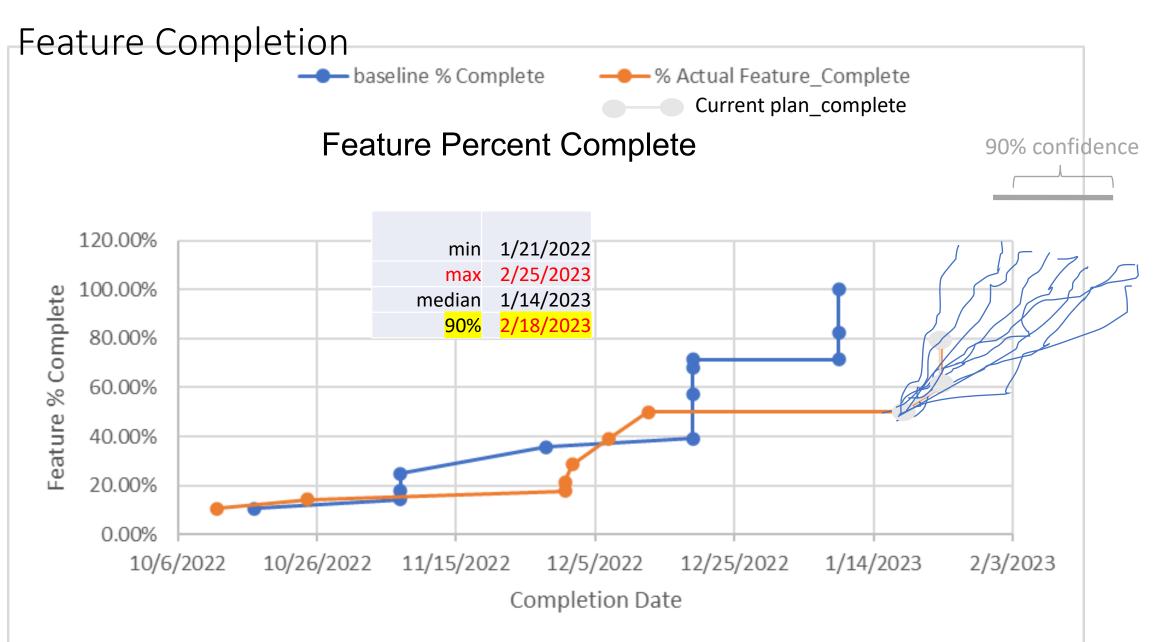
F



Analysis: Did Something Change?

Ę





Ę

Measure Estimation Accuracy and Precision

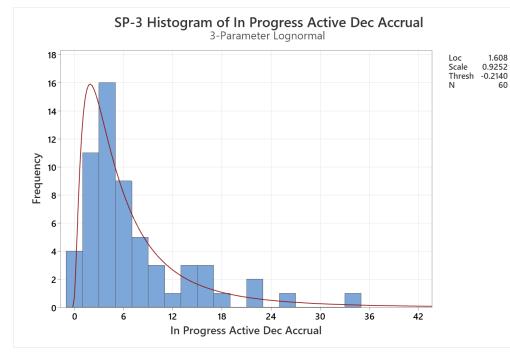
Calibrate for each pipeline.

Determine that work is predictable within a range.

Estimate all work: program planning, road mapping, program increments, sprints.

Measure the work as it passes through the planning, backlog and pipeline.

Compare estimates with actuals.

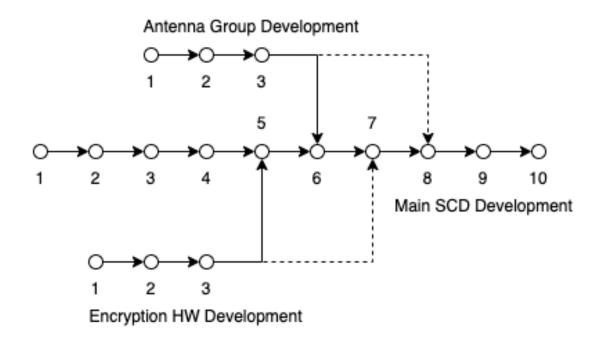


While the estimation is stable, use the ranges to make future predictions.

Statistics Total Cour 60	nt Mean 7.185	SE Mean 0.883	StDev 6.839	
Minimum 0.335	Q1 2.583	Median 4.599	Q3 8.965	Maximum 33.44
Mode 8	IQR 6.382			

Carnegie Mellon University

Pipeline of Pipelines PoPs Workflow Network Example



Model a fictitious device that captures characteristics of a real project dependencies between hardware and software capabilities.

Each pipeline produces **dependencies** used to model schedule, cost, and technical performance risks resulting from production variation, accumulated variance, and rework.

Each pipeline has it's own personality and parameters.

Each category of work item has distinct parameters.

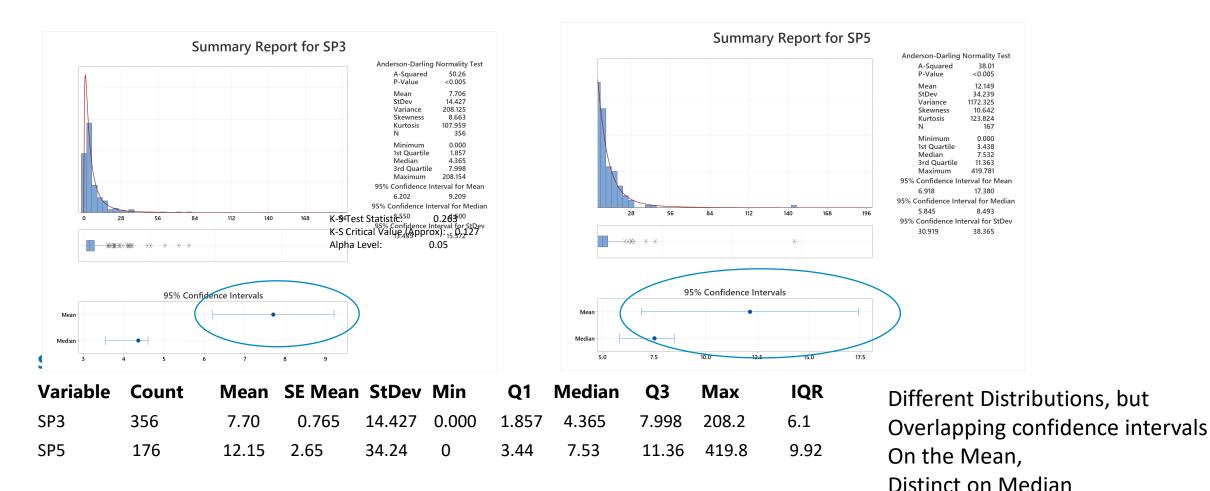
Report dependencies in range of days.

Carnegie Mellon

University

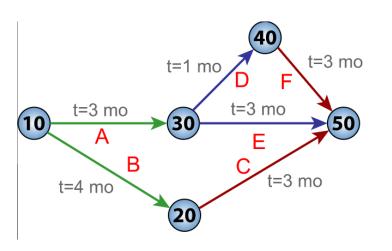
Stories are known unknowns, with predictable distributions

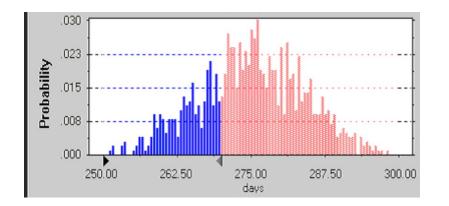
Carnegie Mellon University Software Engineering Institute



Q:Why are some outliers? Q:Can we parameterize the distribution?

Extend Multi-Pipelines with Networks





Approach

- Trace work item through development steps
- Identify blockers and integration points
- Probability of completion date

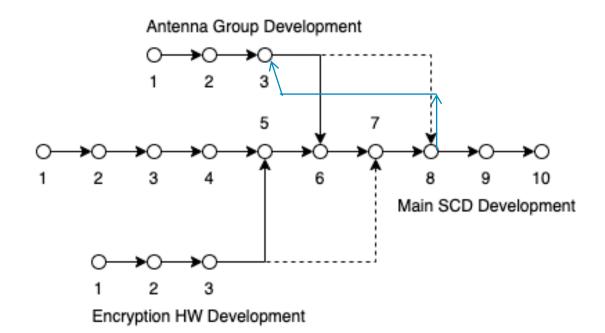
Data from DSO pipeline and other sources

- Product state node structure (capability based WBS, product dependencies, workflow)
- For each Pipeline obtain empirical data for
 - Effort Rate and variation (by skill?)
 - Production Rate and variation by work type
 - Primary work and Rework by activity
 - Defect Rates and fix latencies (build, test)

Carnegie Mellon University Software Engineering Institute

Pipeline of Pipelines PoPs Workflow Network Example

Carnegie Mellon University Software Engineering Institute



Each produce dependencies used to model schedule, cost, and technical performance risks resulting from production variation, accumulated variance, and rework.

Pipelines have backward loops. Sometimes with themselves.

We model likelihood and latency in these loops.

Little's Law assumptions are strongly violated except for **some** linear pipeline segments. Typical Flow Metrics do not accommodate rework, merges, or multiple entry points

Observations and Lessons

Software Engineering Institute Carnegie Mellon University Pittsburgh, PA 15213

Carnegie Mellon University Software Engineering Institute

[DISTRIBUTION STATEMENT A] Approved for public release and unlimited distribution.

Measurement tools are siloed, making it hard to work together.

Different types of work need to be accounted for (product, bugs, research ...). They have different characteristics, estimation accuracy, and outputs.

Averages alone don't support information needs high priority changes or statistical modeling. We need distributions.

We need more specific lead time measures for process steps and baselines for zero rework lead times. (total time until test completes is a candidate quality proxy).

Typical flow metrics don't appear to apply to the pipeline-of-pipelines because of branching and other assumptions violations.

Measuring staff availability remains unresolved.

ACE/PoPs Links for Program Managers

Carnegie Mellon University Software Engineering Institute

Program Managers—The DevSecOps Pipeline Can Provide Actionable Data https://resources.sei.cmu.edu/library/asset-view.cfm?assetid=890538

https://insights.sei.cmu.edu/blog/actionable-data-from-the-devsecops-pipeline/

Why Your Software Cost Estimates Change Over Time and How DevSecOps Data Can Help Reduce Cost Risk

https://insights.sei.cmu.edu/library/why-your-software-cost-estimates-change-over-time-and-howdevsecops-data-can-help-reduce-cost-risk/ https://insights.sei.cmu.edu/library/actionable-data-in-the-devsecops-pipeline/

How to get started Getting Started with ACE/PoPs

https://resources.sei.cmu.edu/library/asset-view.cfm?assetid=890665

Coming Starter packs

Next Steps

Increase number of GQIM indicator Templates Improve projection precision Experience Package

Reports

- Life Cycle Measurement Gap
- Non-Functional Requirement
- Capability Based Estimation Literature Review

Prepare for a Transition, search for transition partners

Stay out of the **SWAMP** (SoftWare Analysis Metrics Pool)

Carnegie Mellon University Software Engineering Institute

Automation requires precise definitions.

Every measurement has precise meaning in a known context. Every metric supports an information need.

Disciplined work decomposition (WBS) connected to product deliverables Categorize and estimate work items Automation support for workflow (e.g. Jira/GitLab) and technical implementation (DevSecOps) Consistent workflow with start and finish Minimize humans in the loop for data collection

Call to Action

Would you benefit from continuous updates to status and projections?

Are you using DevSecOps tool chains, issue trackers, and workflow management?

Can you share process data and discuss results?

Will you participate in our quarterly research review?

We can help!

- Share our Program Management Measurement White Paper
- Specify information, data, and displays for your program management
- Recommend approaches and tools to get started
- Evaluate your results for effectiveness

Carnegie

Iniversity

ACE/PoPs Team

Mellon





Brent Clausner

DevOps Engineer

William Nichols Principal Engineer Telephone: +1 412.268.1727

Email: info@sei.cmu.edu

Luiz Antunes **DevOps Engineer**



Julie Cohen Senior Engineer

ACE/PoPs Team

Carnegie Mellon University Software Engineering Institute



Brigid O'Hearn Senior Engineer



Chris Miller Senior Researcher



Anandi Hira Senior Researcher



Michael Bandor Senior Engineer