# Using Developmental T&E to Inform Operational T&E Decision-Based Analysis

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- The decision to implement a new system is often based on a comparison to a benchmark
- When multiple options are available, we can use DT&E to weed out infeasible options
- Use two-stage statistical methods to decide how to allocate effort in OT&E
  - First stage: represents DT&E
  - Second stage: represents OT&E

#### Choosing the best system configuration



## Confidence Intervals



- Confidence intervals represent the uncertainty in the mean performance of a system based on *n* samples
- Often assume normality in the data
- The half-width should be small enough to ensure that the variation in the mean estimate is acceptable choose  $\delta$  as this acceptable half-width.

## Fixed Sampling Rules – Choosing the sample size



- If a variance estimate is available, can calculate ahead of time how many samples should be taken to obtain a confidence interval with a half-width smaller than δ.
- Challenges:
  - hard to choose  $\delta$
  - *n* might be large
  - Variance estimate might not be available, but can be estimated if samples available

## First stage screening process

- Let benchmark system that defines minimum performance be  $\mu$ .
- Determine the probability of a system having performance better than the benchmark using p-values from first stage (DT&E)

$$p_i = F_{t_{n-1}} \left( \frac{\overline{X}_i - \mu}{\hat{\sigma}_i / \sqrt{n}} \right)$$

• Eliminate the systems with small p-values

$$p_i \le \alpha$$

• Test remaining systems in the second stage.

#### Benchmarked sample size calculation

Instead of using the fixed-sample rule:

$$n \ge \left(\frac{z_\alpha \sigma}{\delta}\right)^2$$

Calculate sample sizes needed to distinguish from the benchmark using first stage information:



Difference in first stage system *i* from benchmark system  $\mu$ 

Systems with performance close to the benchmark will require more samples.

## Choosing the best system



#### Sensors tracking moving target



## System configurations

System Configuration	Coverage Width (each sensor, degrees )
Lynx (single) benchmark	0.72
Dual20	0.20
Dual30	0.30
Dual35	0.35
Dual37	0.37
Dual38	0.38
Dual39	0.39
Dual40	0.40
Dual50	0.50

Dual sensor system is **feasible** if it has a higher probability of detection than the single Lynx.

A dual sensor system is **optimal** if it has the smallest single sensor coverage area out of all feasible systems (assuming smaller coverage is lower cost).

## Sensor configuration example



## First stage results – 30 replications for each config

System Configuration	First Stage Mean	p- value	Number of Samples	Proportion of Samples for Second Stage
Lynx (single) benchmark	0.1457		35	9%
Dual20	0.0408	0		
Dual30	0.0892	0		
Dual35	0.1155	0		
Dual37	0.1273	0		
Dual38	0.1407	0.17	220	58%
Dual39	0.1543	0.96	65	17%
Dual40	0.1542	0.97	57	15%
Dual50	0.2235	1	2	1%

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System	First	p-
Configuration	Stage	value
	Mean	
Lynx (single)	0.1437	
benchmark		
Dual38	0.1404	0
Dual39	0.1473	1
Dual40	0.1542	1
Dual50	0.2279	1

## Conclusions and Future Work

- A planned two-stage experiment can potentially save costly OT&E tests by
  - eliminating configurations from DT&E that are likely infeasible
  - re-allocating effort for DT&E
- More testing should be allocated
  - "close to the boundary" of feasibility,
  - to systems with higher variability
  - systems likely to be the optimal/best
- Could directly incorporate cost into second stage allocations