

## Key Facts

### Use Cases

- ▶ Measure coverage of existing test sets.
- ▶ Verify coverage of constructed Covering Arrays.
- ▶ Qualitative comparison of test sets with identical coverage with additional distribution and distance analysis.

### Architecture

- ▶ Rust implementation:
  - ⇒ Native Executable (Linux, Windows, macOS)
  - ⇒ WebAssembly
- ▶ Modular input parsing.
- ▶ Machine- or human-readable output.

### Implementation Highlights of CAMetrics

- ▶ Significantly fast based on experiments:
  - ⇒ Multi-threaded
  - ⇒ Multiple algorithms
- ▶ Low memory usage.
- ▶ Web UI and command line interface.
- ▶ Sophisticated constraint support.

### Future improvements

- ▶ HPC and cluster implementations.
- ▶ Faster constraint processing.
- ▶  $\alpha$ -balance.

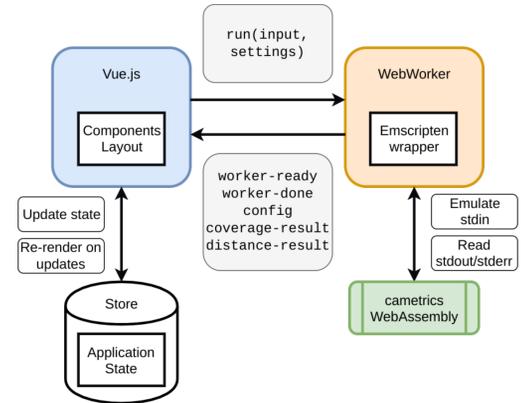


Figure 1: Web UI architecture

## Coverage

- ▶ Simple  $t$ -way combination coverage: How many  $t$ -selections of parameters are fully covered?
- ▶ Simple  $(t + 1)$ -way coverage:  $t + \frac{\text{Covered } (t+1)\text{-tuples}}{\text{Total } (t+1)\text{-tuples}}$

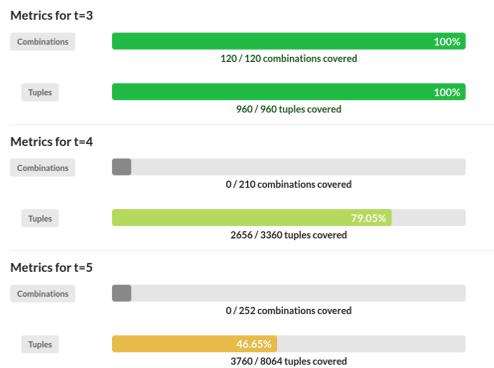


Figure 2: Simple  $t$ -way combination and tuple coverage



Figure 3: Coverage Map showing  $t$ -tuple coverage per parameter selection

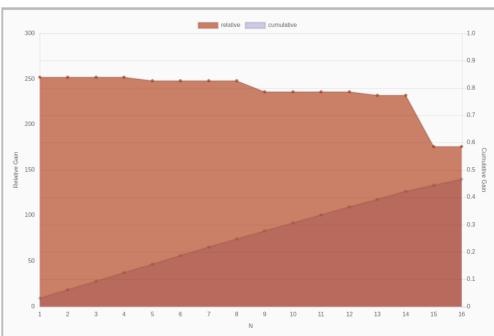


Figure 4: Per-test and cumulative coverage gain.

The coverage gain describes how many previously uncovered tuples are covered by a single test.

## Select between Two Algorithms

Input:  $N \times k$  array, alphabet size  $v$ , strength  $t$

cametrics-fast

- ▶ Great general-purpose performance
- ▶  $\mathcal{O}\left(\binom{k}{t} v^t\right)$  memory
- ▶  $\mathcal{O}\left(\binom{k}{t} N\right)$  time

cametrics-light

- ▶ Prevents memory explosion, great for binary/small arrays
- ▶  $\mathcal{O}\left(\binom{k}{t}\right)$  memory
- ▶  $\mathcal{O}\left(\binom{k}{t} N v^t\right)$  time

## Distance Metrics

### Inter-test distance: baseline for success

- ▶ (Generalized) Hamming Distance: How many parameter values differ between tests?
- ▶ Total Cartesian/Euclidian Distance of test  $T$  to array  $A$ :

$$CD(A, T) = \sum_{n=1}^{|A|} \sqrt{\sum_{l=1}^k (A_{nl} - T_l)^2}$$

### Balance is everything!

- ▶ *Balanced* array: Each parameter value (or  $t$ -tuple) appears roughly the same number of times.

### Modified $\chi^2$ Distance:

- ▶ How close to ideal distribution of parameter values?

Ideal distribution:  $D_{ij} = \frac{1}{v_j} |j \in \{1, \dots, k\}, j \in V_i$

Actual distribution:

$$D_{ij} = \frac{1}{N} \sum_{n=1}^N \begin{cases} 1 & \text{if } A_{ni} = V_{ij} \\ 0 & \text{otherwise.} \end{cases}$$

$$\text{Modified } \chi^2 \text{ Distance: } \left( \sum_{j=1}^k \frac{v_j - 1}{v_j + 1} \right) - \frac{\sum_{j=1}^k \sum_{i=1}^N \frac{(D_{ij} - D_{ij}^{\text{ideal}})^2}{D_{ij}^{\text{ideal}}}}{2}$$

960 tuples to cover in total, 120 max coverage gain per test

snip  
[1, 1, 1, 1, 0, 0, 0, 0, 0] i = 10 H\_min = 4 H\_max = 8 H\_sum = 48 CD\_sum = 20.6717 X^2 = 3.31313131 848 covered (88.33%), 40 gain  
[0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1] i = 11 H\_min = 4 H\_max = 6 H\_sum = 52 CD\_sum = 22.7422 X^2 = 3.31262940 880 covered (91.67%), 32 gain  
[1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1] i = 12 H\_min = 4 H\_max = 8 H\_sum = 60 CD\_sum = 25.5707 X^2 = 3.33333333 912 covered (95.00%), 32 gain  
[0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1] i = 13 H\_min = 4 H\_max = 6 H\_sum = 60 CD\_sum = 26.7875 X^2 = 3.31851852 928 covered (96.67%), 16 gain  
[1, 1, 0, 0, 0, 0, 1, 1, 1, 0, 1] i = 14 H\_min = 4 H\_max = 8 H\_sum = 68 CD\_sum = 29.6160 X^2 = 3.32307692 944 covered (98.33%), 16 gain  
[0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1] i = 15 H\_min = 2 H\_max = 10 H\_sum = 72 CD\_sum = 31.3640 X^2 = 3.32220986 952 covered (99.17%), 8 gain  
[1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1] i = 16 H\_min = 2 H\_max = 10 H\_sum = 80 CD\_sum = 34.1924 X^2 = 3.33333333 960 covered (100.00%), 8 gain  
Optimal X^2 value: 3.33333333333333

Figure 5: Distance measurements: Minimum/Maximum/Total Hamming Distance, Total Cartesian Distance,  $\chi^2$ , and coverage