



# Getting The Right Answer with Machine Learning

IEEE-EDPS 2021

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SIEMENS

File Tools Simulation View Help

Simulate Design History Generate report Help

Design Hierarchy Start HSMC-2 x HSMC-0 x PVTMC-0 x

Distributions Statistical summary Impact on outputs Setup Simulation results Simulation status

Show:  Probability density  Normal quantile

Design Hierarchy

- Device
- Type
- cload capacitor
- vbias v
- vdd v
- vin v
- vinb v
- xopamp opamp
- xmxn1 nmos
- xmxn2 nmos
- xmxn3 nmos
- xmxp0 pmos
- xmxp2 pmos

Corners

PVTMC-0

- (bandwidth-min) 50.0\_2.1\_1.1
- (dc\_gain-min) 100.0\_1.9\_1.30
- 0.0\_1.9\_1.20
- 0.0\_1.9\_1.25
- 0.0\_1.9\_1.30
- 0.0\_2.0\_1.20
- 0.0\_2.0\_1.25
- 0.0\_2.0\_1.30
- 0.0\_2.1\_1.20
- 0.0\_2.1\_1.25
- 0.0\_2.1\_1.30
- 100.0\_1.9\_1.20
- 100.0\_1.9\_1.25
- 100.0\_2.0\_1.20
- 100.0\_2.0\_1.25
- 100.0\_2.0\_1.30
- 100.0\_2.1\_1.20
- 100.0\_2.1\_1.25
- 100.0\_2.1\_1.30
- 25.0\_1.9\_1.20
- 25.0\_1.9\_1.25
- 25.0\_1.9\_1.30
- 25.0\_2.0\_1.20
- 25.0\_2.0\_1.25
- 25.0\_2.0\_1.30

Project Files

- data.l
- netlist
- opamp.sp
- solido\_example1.l

Simulation Logs

Probability density of dc\_gain

Normal quantile of dc\_gain

Outputs

Name	Value
<b>Summary</b>	
Circuit	Netlist
Simulator	finesim (hspice-compatible)
No. of run sims	1,299
Brute-force equivalent	45M
No. of Corners	45
Target Sigma1	4.1o
<b>Selected corner</b>	
Corner name	(dc_gain-min) 100.0_1.9_1.30
Accuracy	Verified
Model Set	ttglobalcorner_localmc
Circuit temperature	100.0
vcc	1.9
vref	1.30
-4.1o	16.42
4.1o	20.01
Min	16.53
Max	20.71
Mean	18.22
Std. dev.	435.3m
Skewness	244.0m
Kurtosis	-305.0m
Ordinal class ID	1

Results

- SENS-0 9/9
- HSV-1 error
- SIM-3 1/1
- HSV-0 error
- PVTMC-0 1299/1299
- HSMC-2 4000/4000
- HSMC-1 4000/4000
- FPVT-0 630/630
- HSMC-0 7410/7410
- SPVT-0 308/308

Circuit temperature: All

vcc: All

vref: All

Apply filters

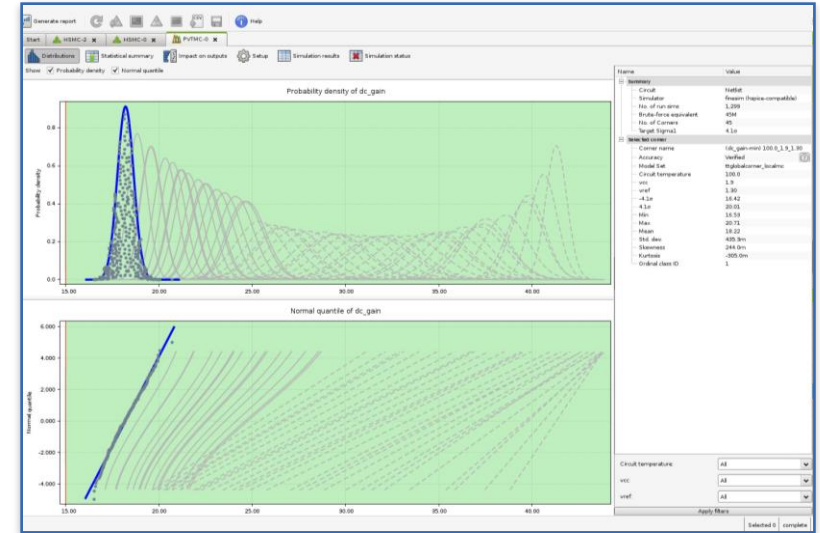
Selected 0 complete 8:02s 100%



# Solido ML Tools and Benefits

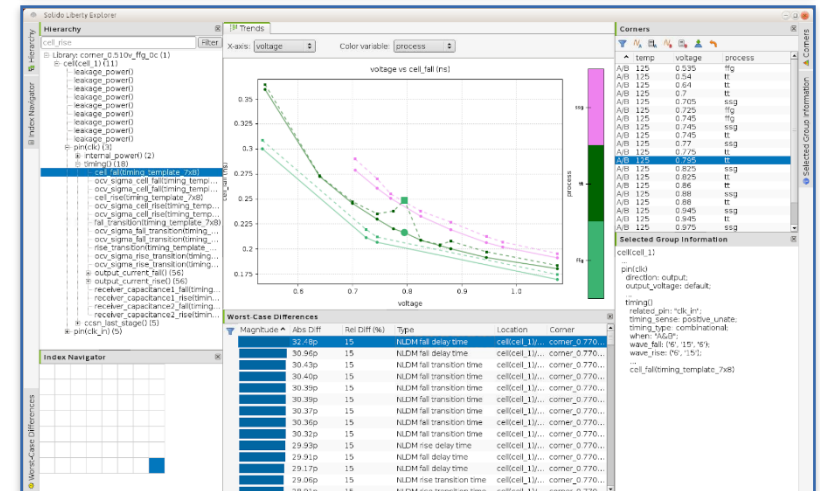
## Solido Variation Designer

- 2-50X faster PVT analysis
- 10X faster 3-sigma Monte Carlo (1 PVT)
- 100X faster 3-sigma Monte Carlo (many PVTs)
- >1KX faster 4.5-sigma Monte Carlo
- >1MX 6-sigma Monte Carlo



## Solido Characterization Suite

- Generates Liberty models at new PVTs 1000X faster
- Reduces overall timing characterization schedule by 50%
- Finds new classes of errors in Liberty data
- Reduces Liberty debugging time from weeks to hours



## Great ML Requires More Than Just Speedups...

**Verifiability:** Can you tell if the ML model is right?

**Accuracy:** How close to perfect is the ML model?

**Generality:** Does the ML approach work on everything?

**Robustness:** Can I bet my next design schedule on it?

**Usability:** Does it “just work” for my team?

# ML CAD Tool Capability Levels

## **Level 0: No ML**

Gives the right answer by running all cases. Often too compute-intensive for production use.

## **Level 1: Partially reliable ML**

Big speedups. Sometimes right, sometimes wrong, and can't tell. Cool demo! Shows promise. Not useful, yet.

## **Level 2: Partially reliable ML with accuracy-aware self-verification**

Can tell when it's right or wrong based on accuracy criteria. Tool can't solve it automatically. Need a backup plan.

## **Level 3: Adaptive, accuracy-aware ML with self-verification**

Meets accuracy criteria on most cases automatically. Still the odd bump. Useful, with great support.

## **Level 4: Fully reliable, accuracy-aware ML with self-verification**

Production-ready and dependable. Production hardened and proven. Just works. Use it.

## ML Levelquest and Effort

- **Level 0 to 1:** Identifying and testing ML opportunities (hours-days)
- **Level 1 to 2:** Defining what “accurate” means; making the algorithm self-verifying (weeks-months)
- **Level 2 to 3:** Developing methods for achieving accuracy dynamically and automatically (months-years)
- **Level 3 to 4:** Years of production hardening and addressing corner cases (years-decades)
- Each level is ~an order of magnitude more work

## Level 2: Defining “Right” and Proving It Automatically

### Define what “right” means for a given engineering problem

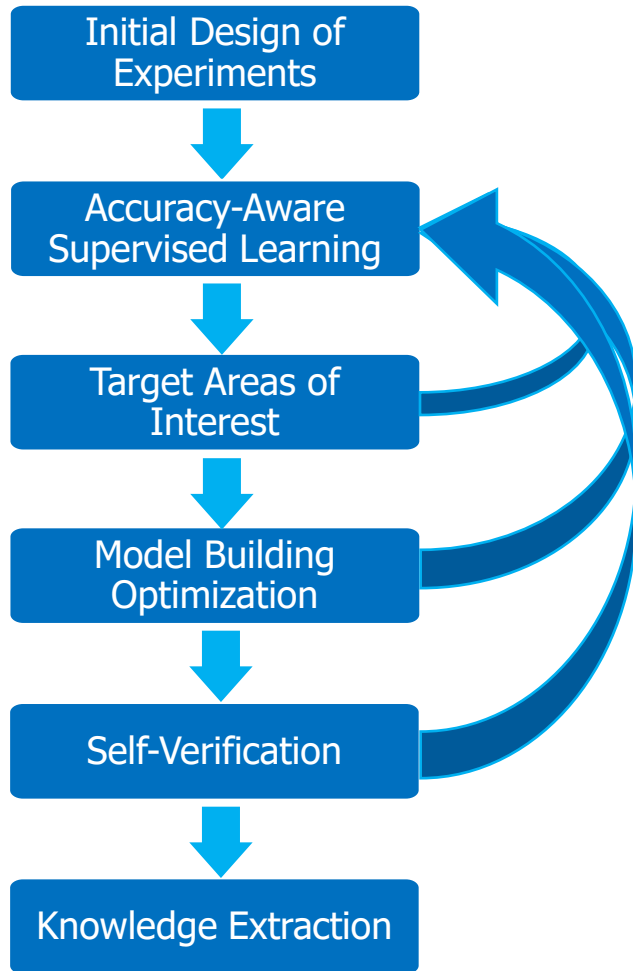
- Tolerances: Target range for a correct answer, absolute, relative, conditional
- Acceptable error rates, if applicable: Statistical affordances
- Biggest acceptable outliers: Min/max
- Regions of interest where higher accuracy is needed
- Regions of disinterest where no answer is even needed
- E.g. +/- 2 ps absolute error, <2% relative error for voltages <0.7, 99.5% of the time; 5ps, 5% tolerable for voltages  $\geq 0.7$
- This is actually hard!

### Designing an algorithm that can tell when ML-generated results are right, automatically

- Automated spot checking with spacefill sampling
- Automated checking in regions of highest interest
- Accuracy-aware modeling and automated checking in regions that have the widest model confidence intervals
- Self-verifying algorithms by design
- How would you prove it manually? Maybe do that?



## Getting to Level 3 With Solido-Style Active Learning



Start by simulating a small amount of the total space, equally spread and covering all main and interaction effects

Build an accuracy-aware ML model that can very quickly predict outcomes for the rest of the space

Use the predicted values to decide which cases are interesting and simulate those; get simulation accuracy where it matters

Make the rest of the predictions adequately accurate; run more simulations in areas of greatest uncertainty

Prove that the areas of interest are covered and that target accuracy is achieved throughout the space

Help with insight and debugging by revealing dominant terms in ML models

## Solido's First Level 3 ML Solution: High-Sigma Monte Carlo (c.2008)

- Challenge: Monte Carlo and SPICE accurate high-sigma analysis
- Order of magnitude more simulations per 0.5 sigma!

Sigma	Simulations
4	~1M
4.5	~10M
5	~100M
5.5	~1B
6	~10B

- Goal: MC and SPICE accuracy in 1000s of simulations (because that is all we have time for in production flows!)

## Solido's First Level 3 ML Solution: High-Sigma Monte Carlo (c.2008)

The screenshot shows the Solido Variation Designer interface for running a High-Sigma Monte Carlo simulation. The main configuration area is titled "Run High-Sigma MC (Beta)".

**Specifications**

- Output specs (2 set): 1M samples generated. Verifies to 4.257 sigma2.

**Stopping criteria**

Stop after any of:

- All failures found. Up to a maximum of 1K failures.
- 5K samples simulated.
- 8.0 hours.

**Variation**

- Using corner Netlist (sa\_offset.sp).
- Global process.
- Local/mismatch process: (all devices).

**Description**

Finds failures and estimates yield quickly, by ordering the samples for the outputs and specs of interest. It generates 1M, 10M or more Monte Carlo samples, but only simulates the ones it predicts to be extreme. Can be 100x+ faster than traditional Monte Carlo.

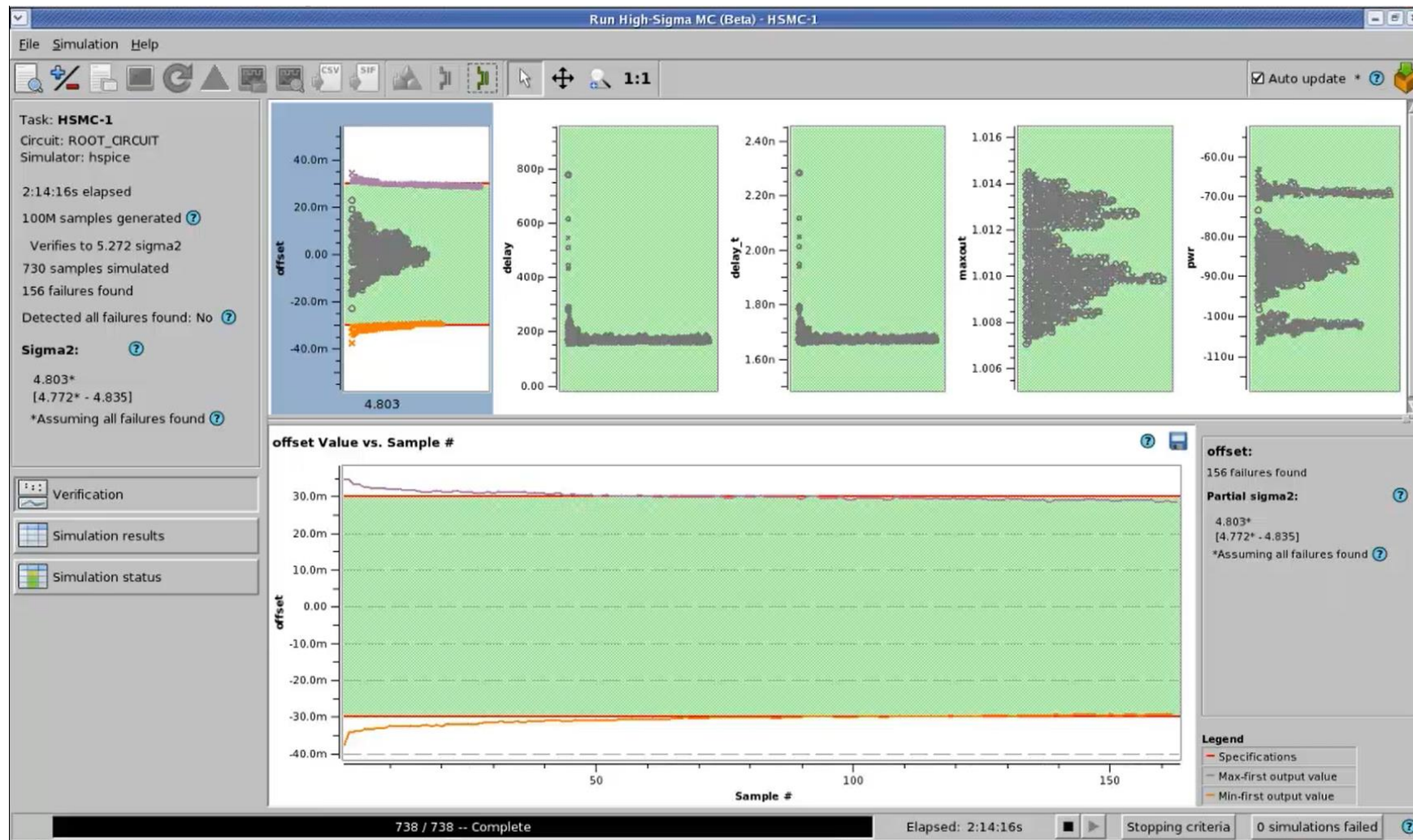
**Simulations (est.)**

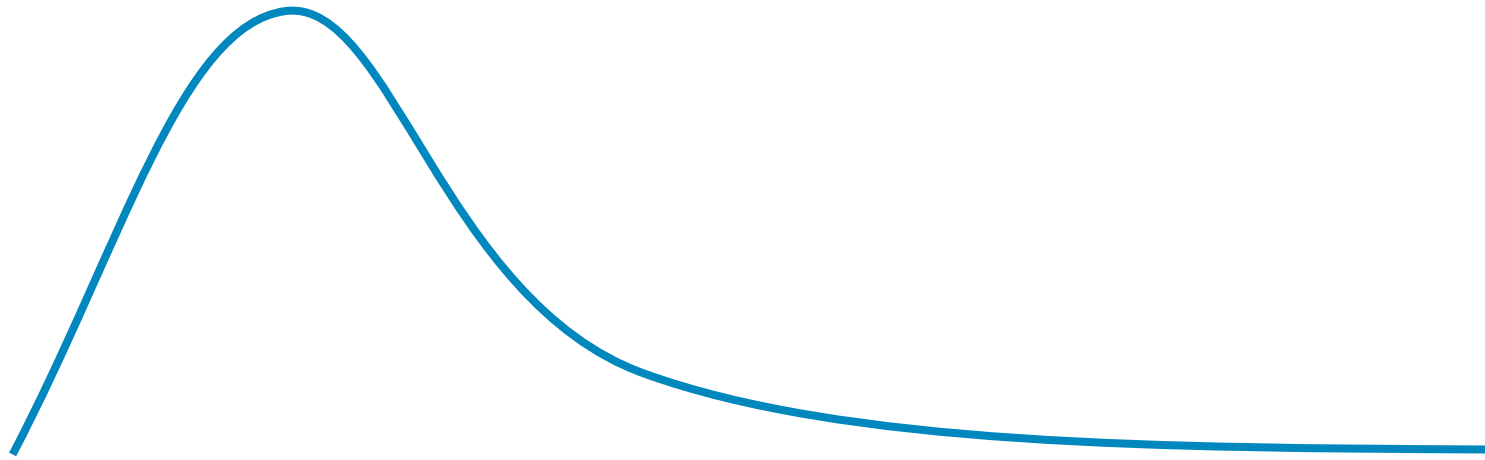
(400 initial samples + (2300 samples per spec x 2 specs)) x 1 corner = **5000 simulations** or **all failures found**

Buttons: Launch, Schedule, Show results.

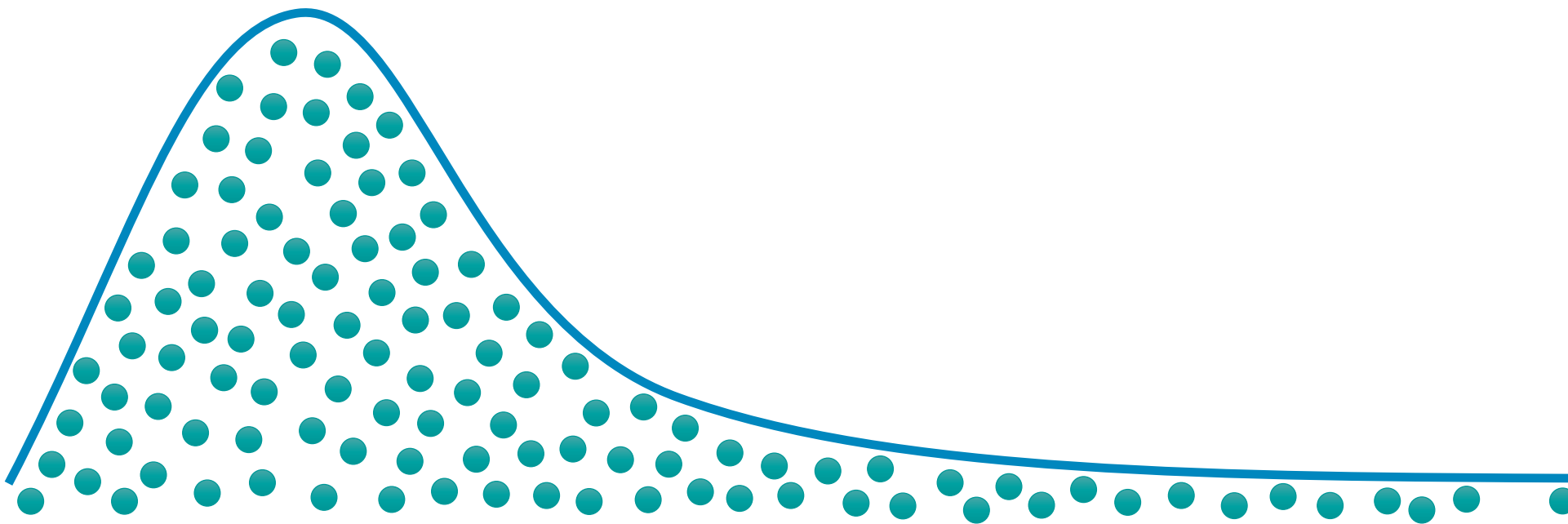
Name	Task	Time	Elapsed	Status	Progress
HSMC-2	Run High-Sigma MC (Beta)	Jan 12 01:51	2:20:06s	complete	10000 / 10000
HSMC-1	Run High-Sigma MC (Beta)	Jan 11 18:44	2:14:17s	complete	738 / 738

# Solido's First Level 3 ML Solution: High-Sigma Monte Carlo (c.2008)



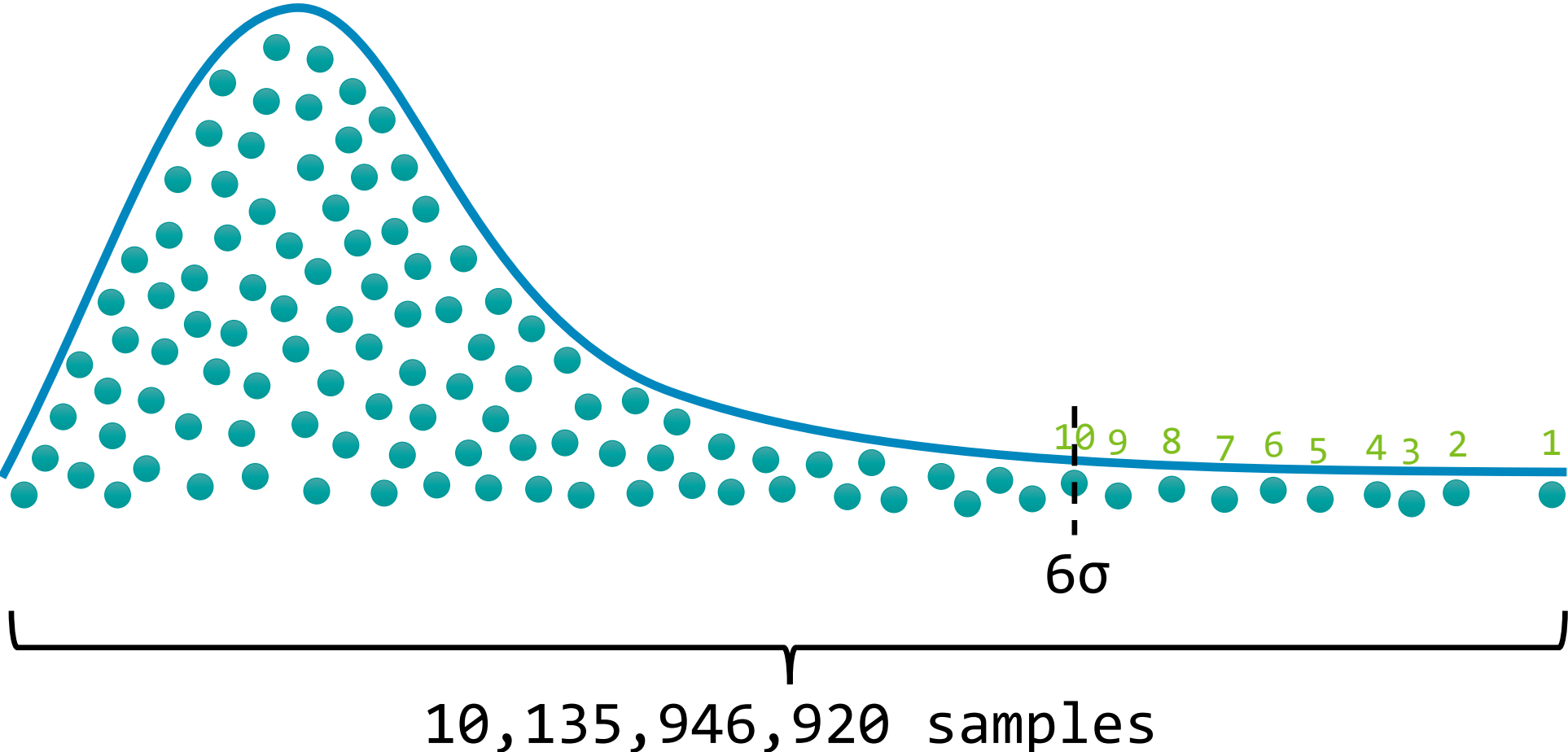


## Brute Force Monte Carlo



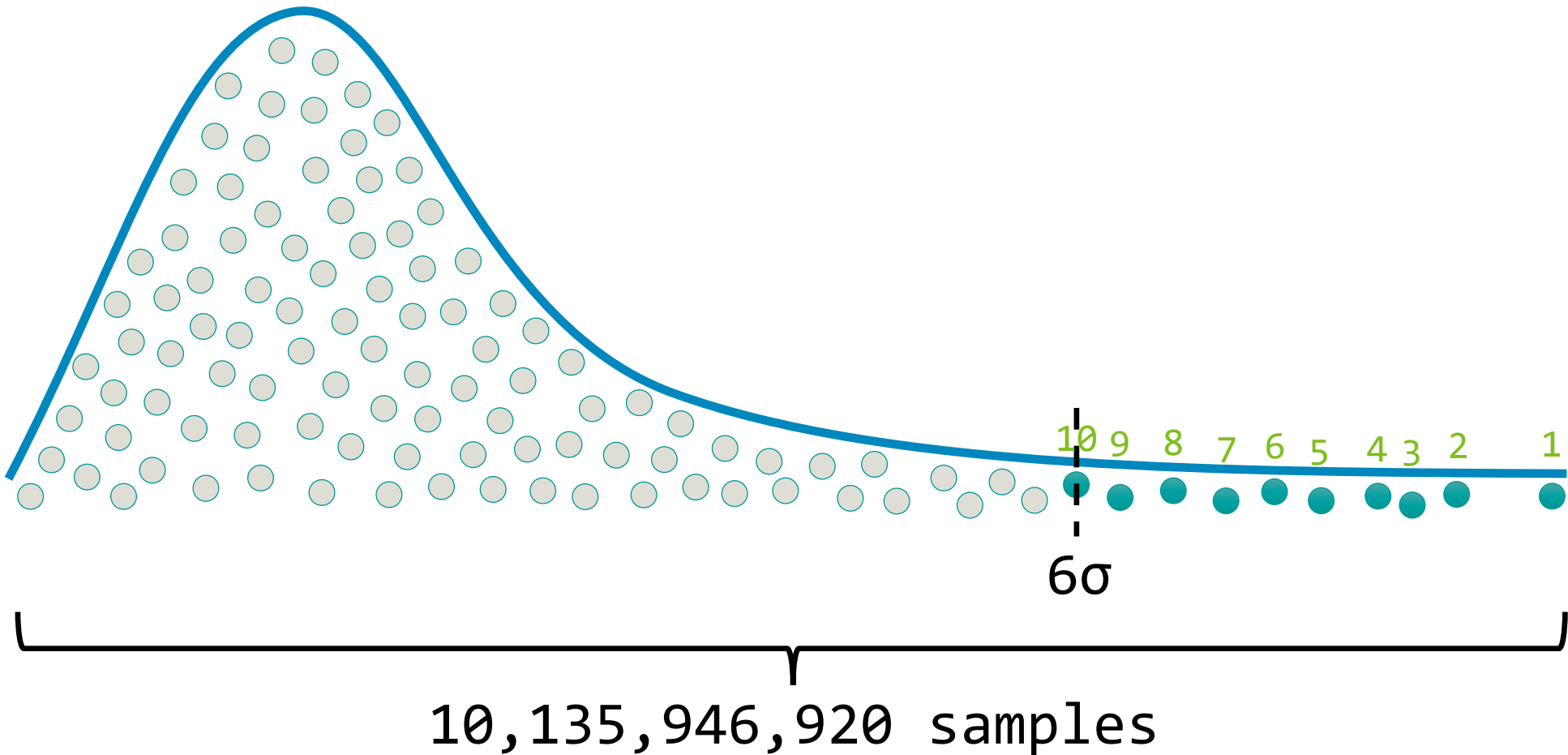
# Brute Force Monte Carlo

$6\sigma = 10$  failures per 10,135,946,920



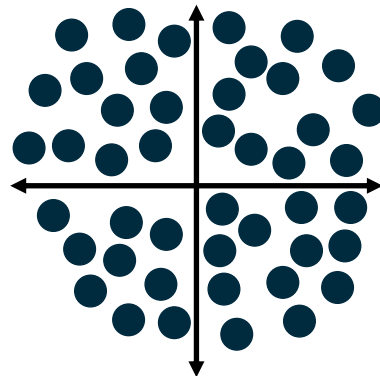
What if...

$6\sigma = 10$  failures per 10,135,946,920



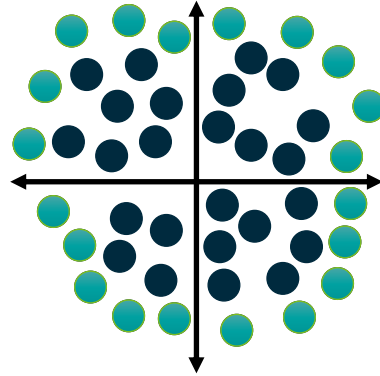


## High-Sigma Monte Carlo



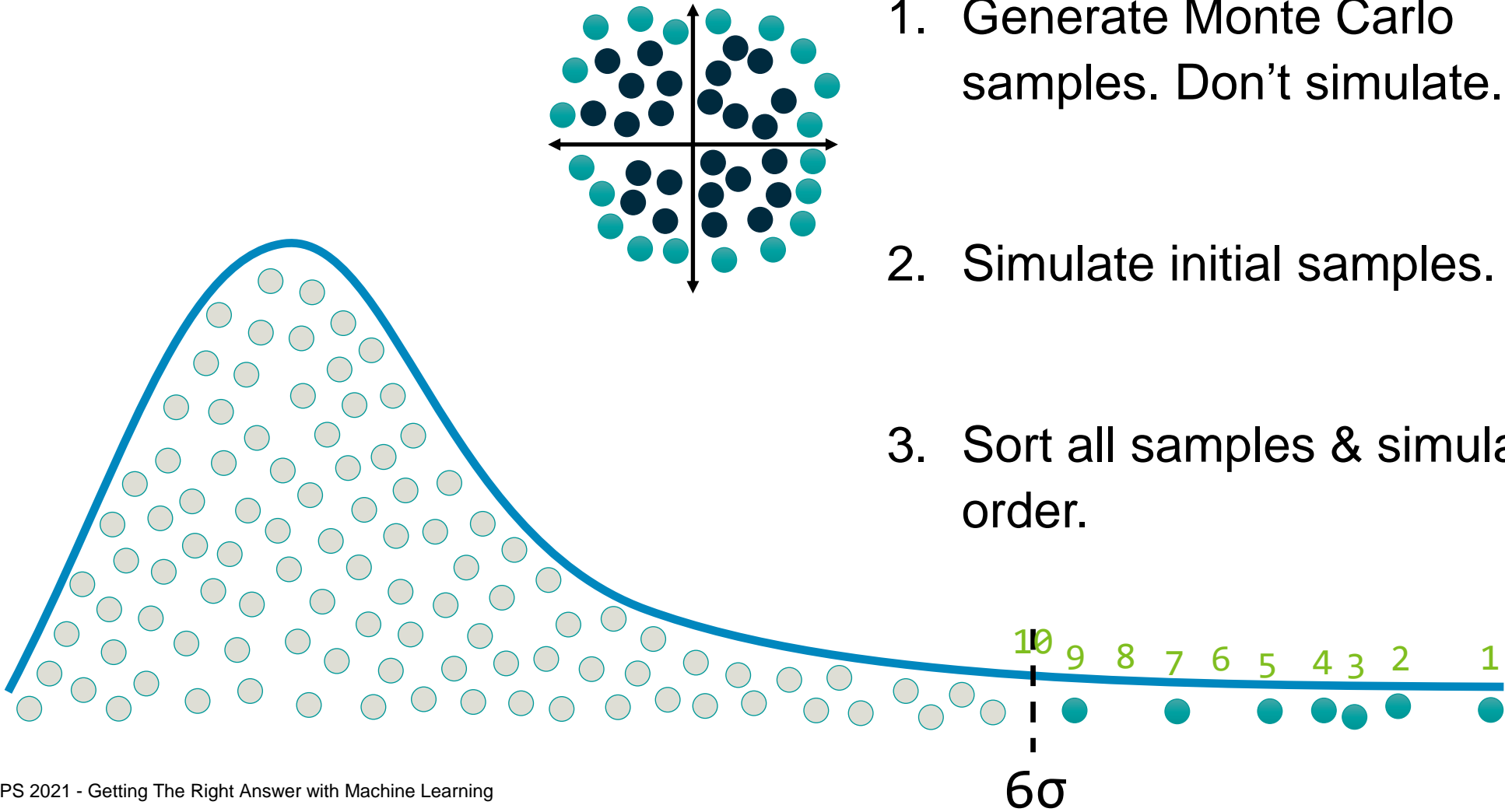
1. Generate Monte Carlo samples. Don't simulate.

## High-Sigma Monte Carlo



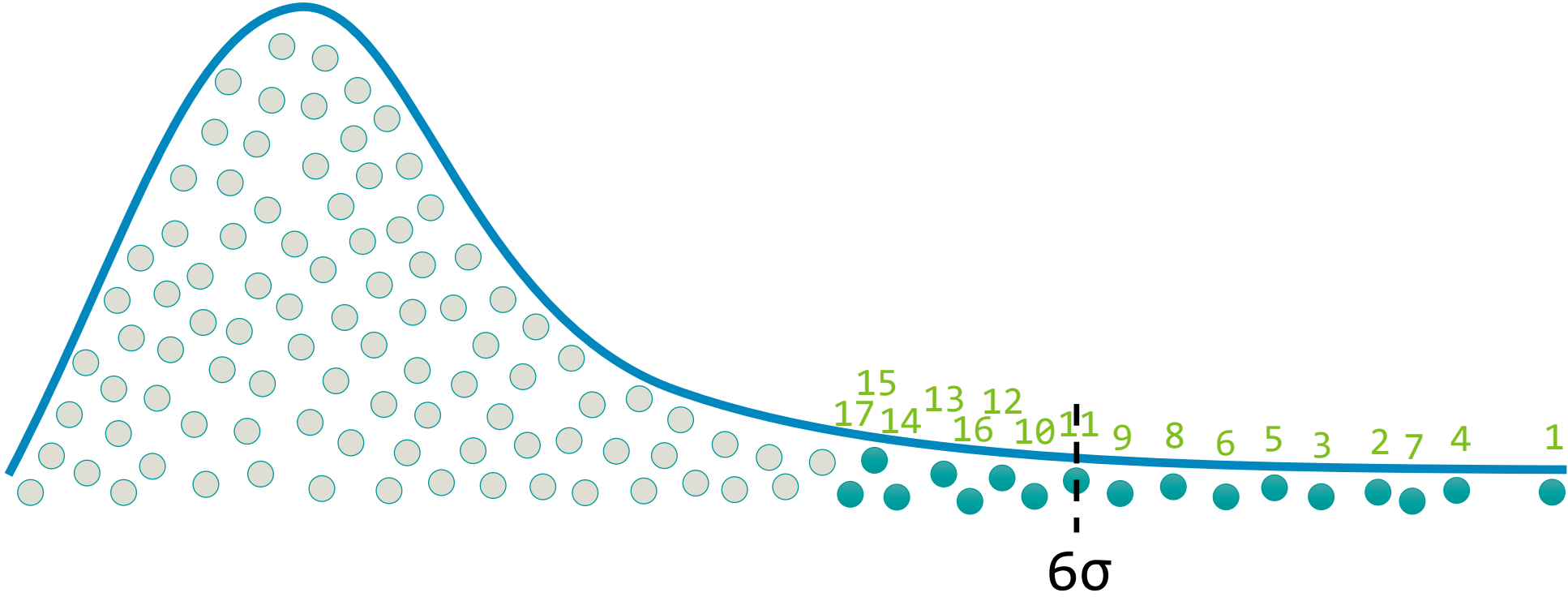
1. Generate Monte Carlo samples. Don't simulate.
2. Simulate initial samples.

# High-Sigma Monte Carlo

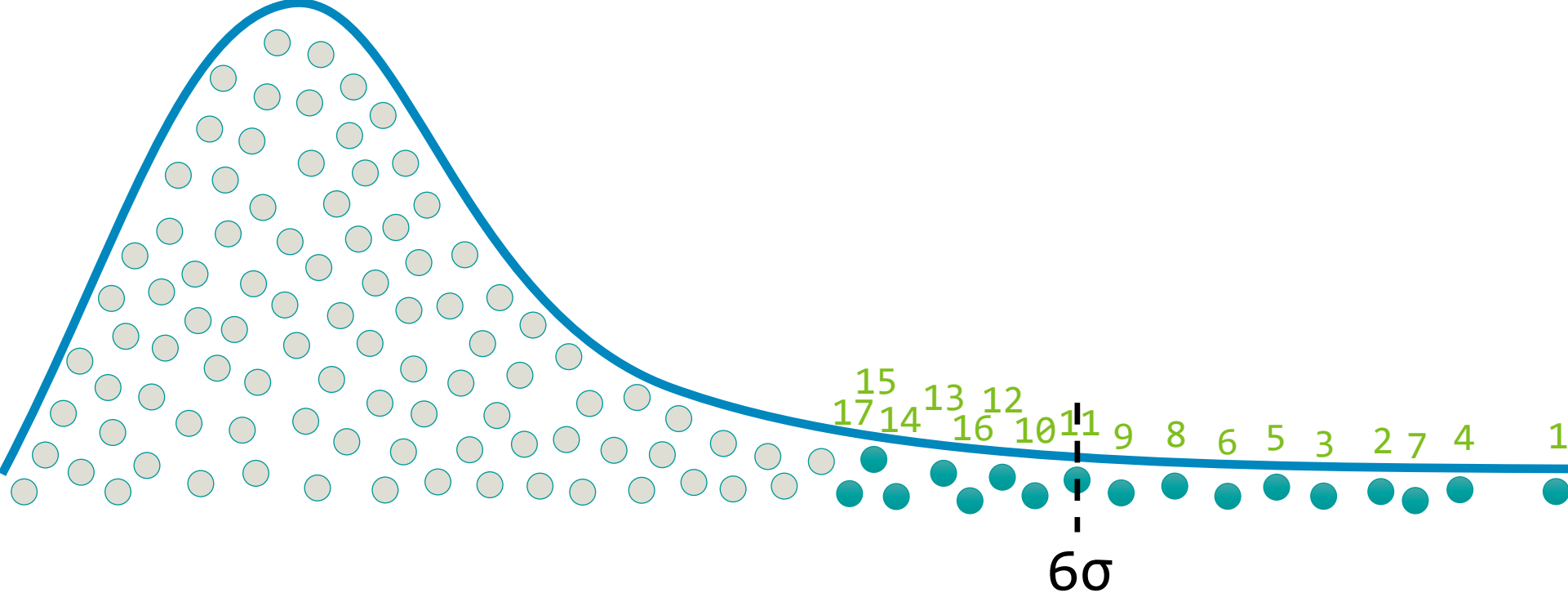


1. Generate Monte Carlo samples. Don't simulate.
2. Simulate initial samples.
3. Sort all samples & simulate in order.

# High-Sigma Monte Carlo applied



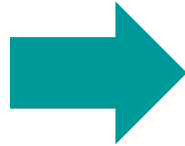
# High-Sigma Monte Carlo applied



## HSMC: Getting From Level 3 to Level 4

### 2008: Invented – level 3

- Capacity: 1M samples
- 1000 process variables
- Continuous outputs only
- 1 failure region
- Self-verification shown in logs
- Manual setup and operation
- Required stable simulation env.



### 2021: Person decades later – nearing level 4

- Capacity: 1T samples
- 100K process variables
- All outputs (e.g. multi-modal, n-ary)
- N failure regions
- Self-verification GUI
- Fully adaptive; no expertise needed
- Handles, recovers from, and corrects sim failures

## Toward Level 4: Finding and Fixing the Corner Cases

- This is the expensive part
- 100s of person years invested to date in going from level 3 to 4

# Level 3 to 4 Example: Accurate ML Modeling for Jagged Waveforms Using Better Descriptive Parameters

Scott Johnston

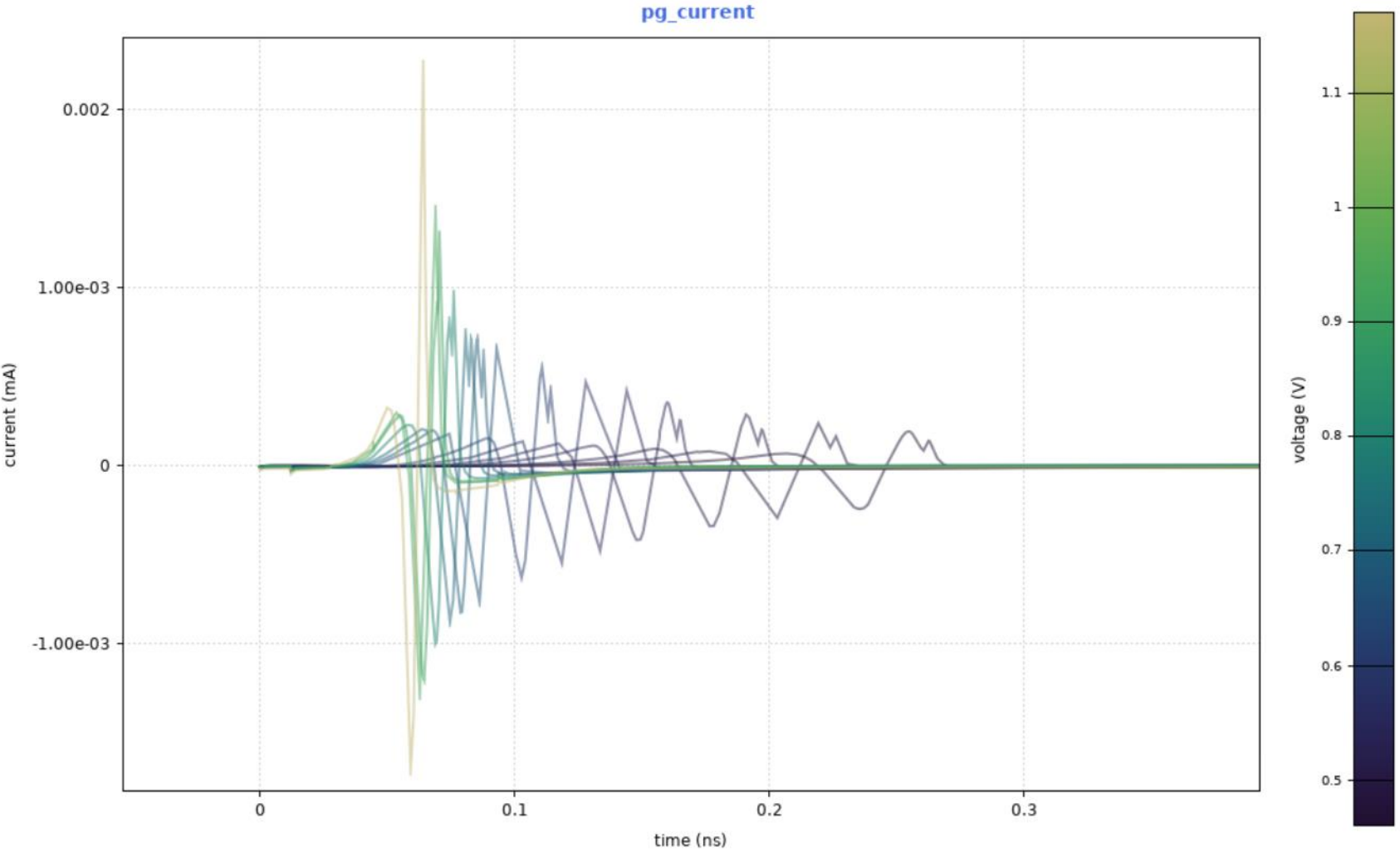
James Marquez

Ken De Lathouwer

Megan Marsh



# Modeling Jagged Waveforms is Hard



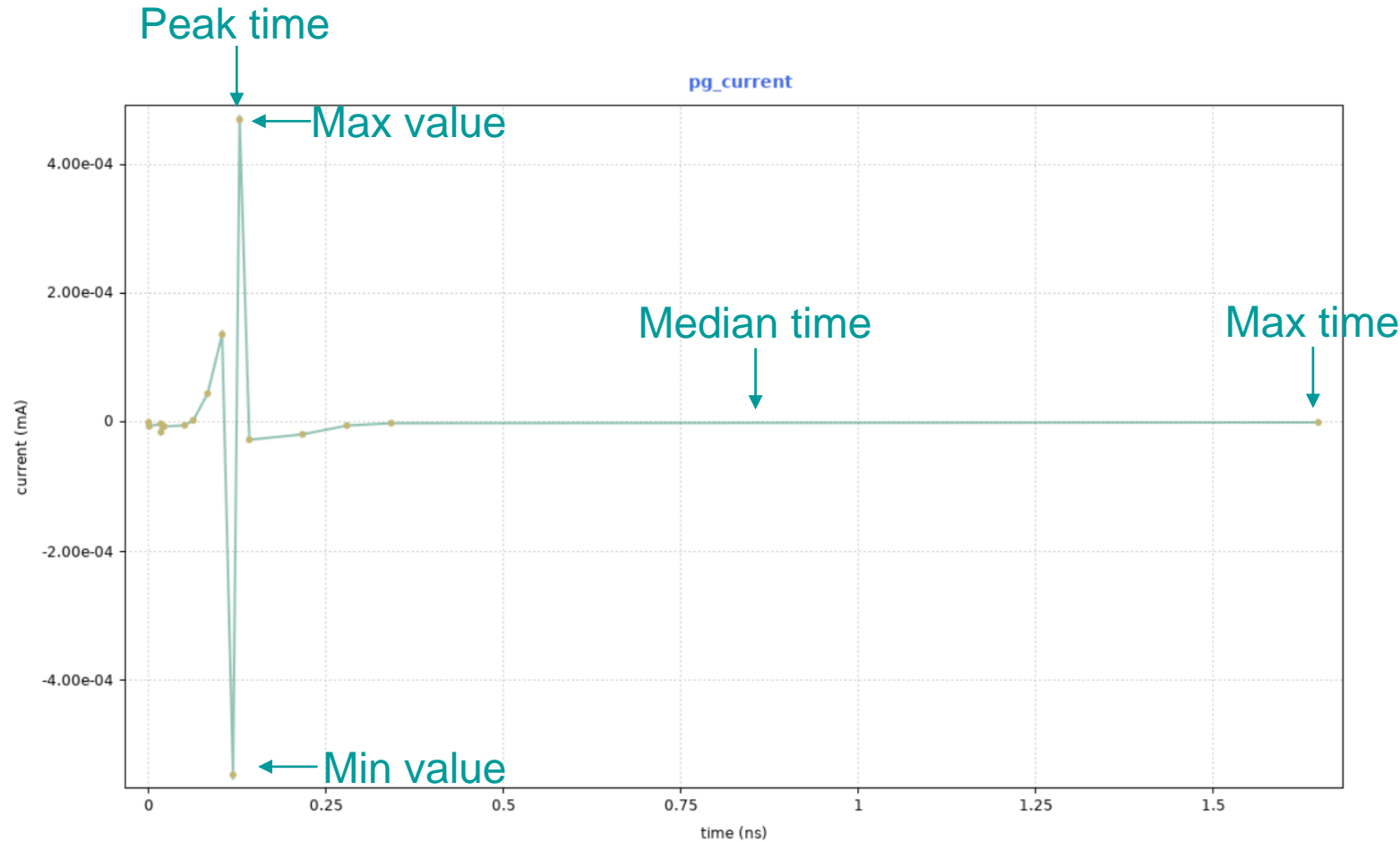
Waveforms are shapes, not values

Accuracy of the shape is really important

Jagged shapes are really hard to model

Modeling the jagged shape directly fails

# Modeling Jagged Waveforms Using Their Parameters



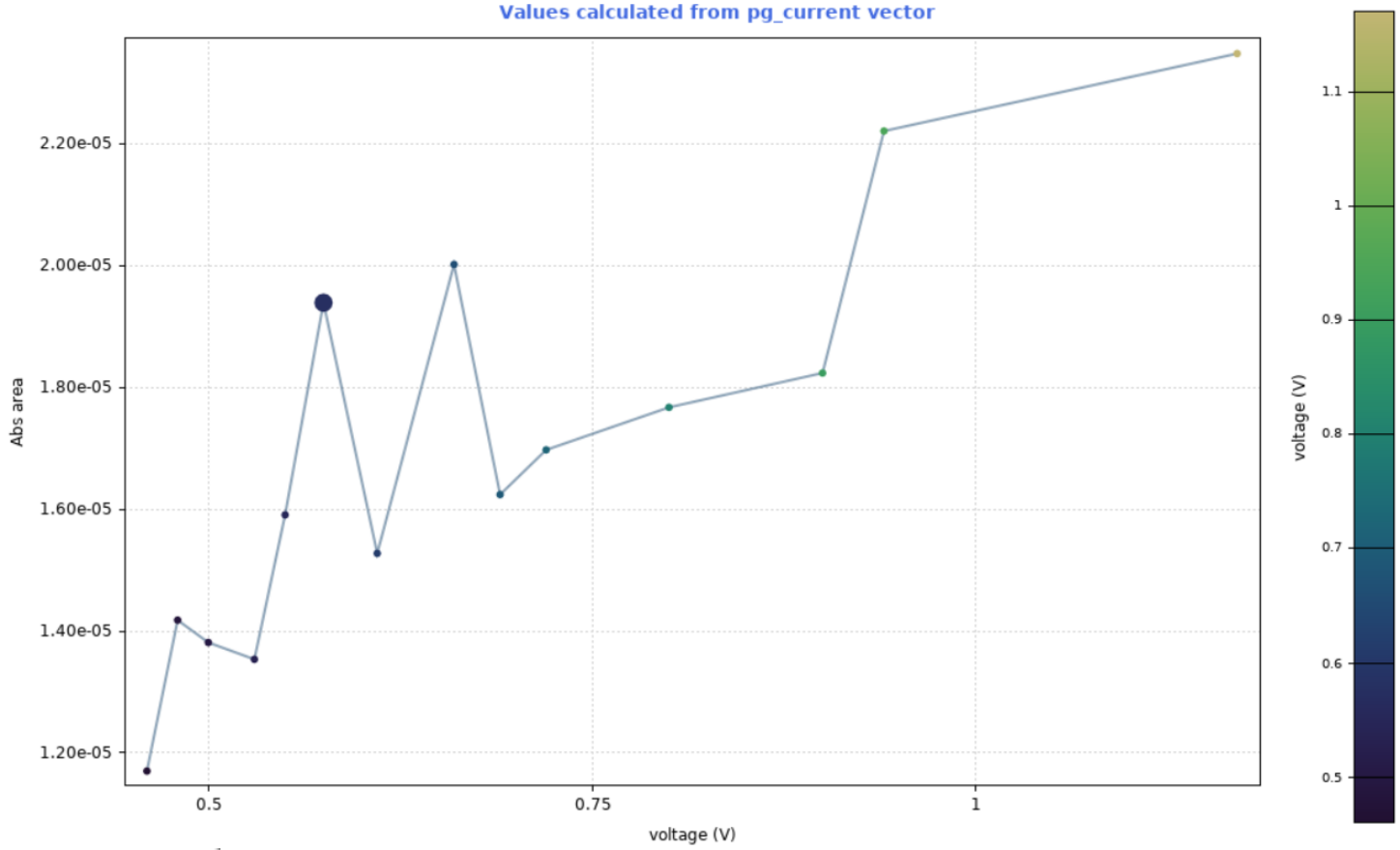
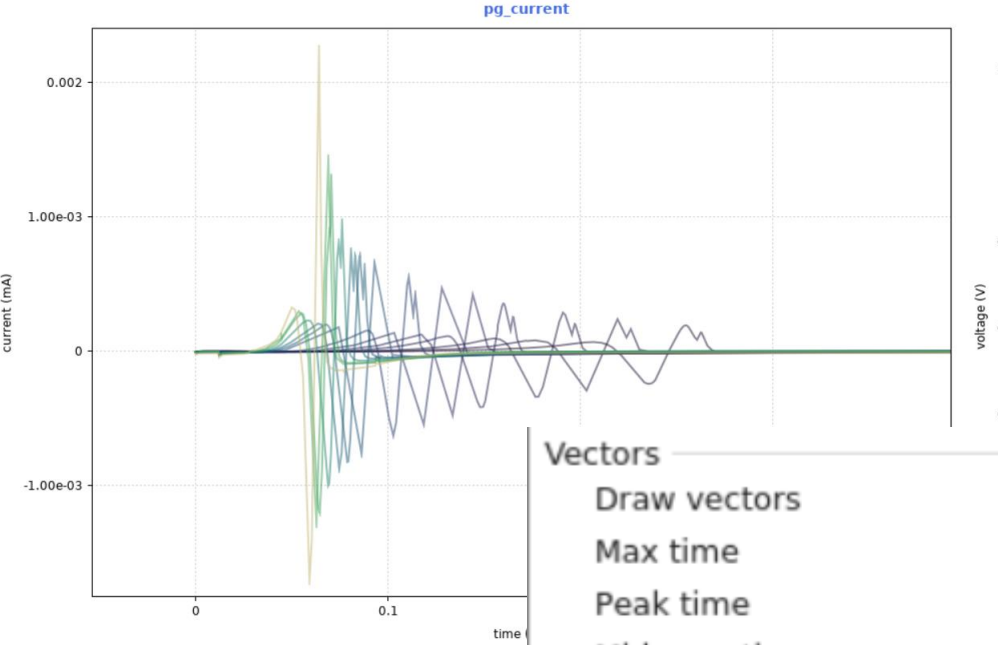
Observation: The general shape holds across input dimensions

Observation 2: Waveforms are defined by continuous numeric parameters

Can we model and predict the parameters of a waveform?

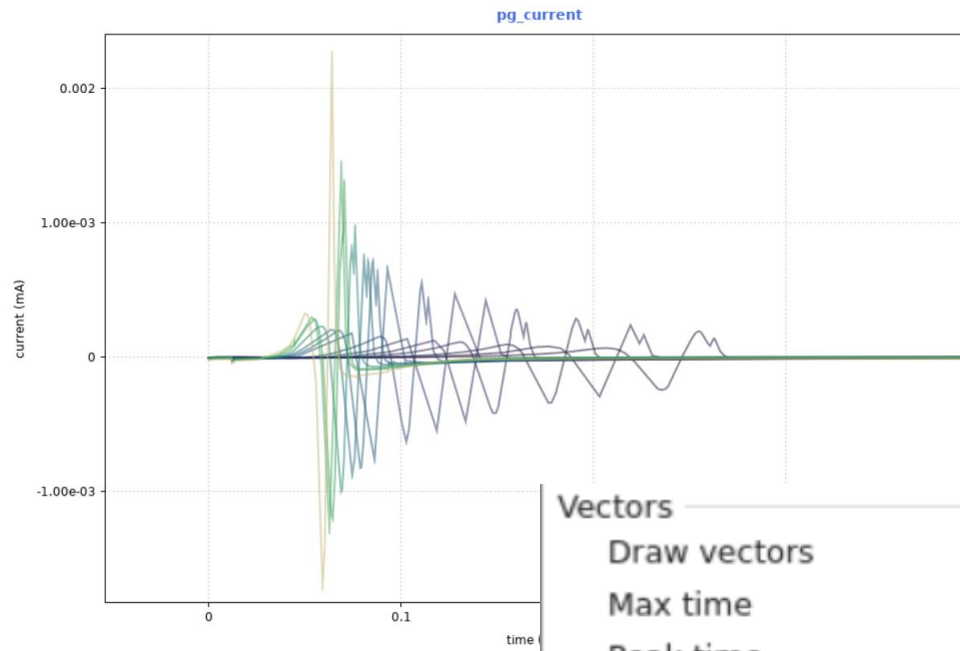
Can we use those predicted parameters to morph the waveform to other input conditions?

# Absolute Area: Too Noisy to Model

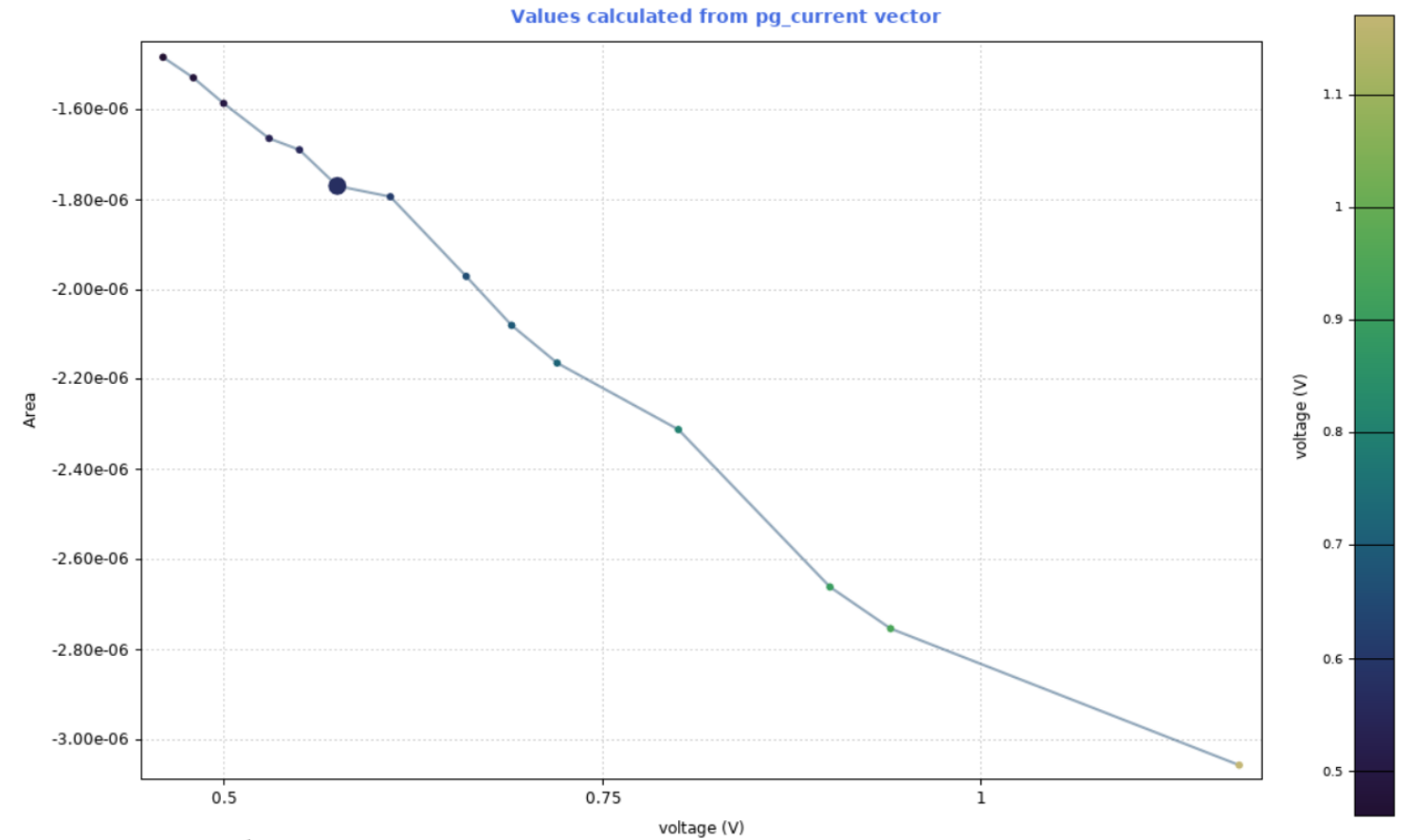


- Vectors**
- Draw vectors
  - Max time
  - Peak time
  - Mid-area time
  - Median time
  - Max value
  - Min value
  - Peak value
  - Last value
  - Area
  - Abs area**

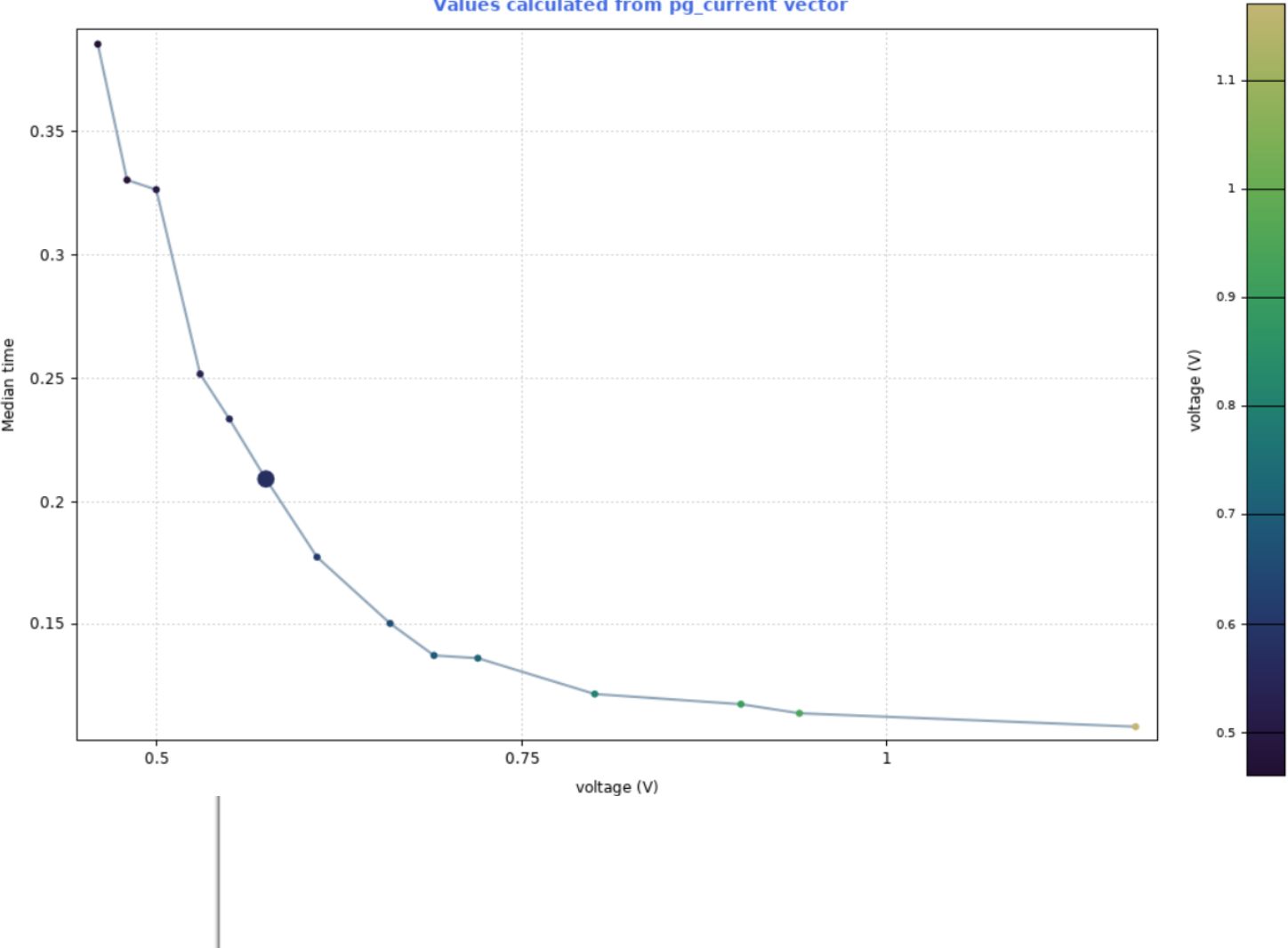
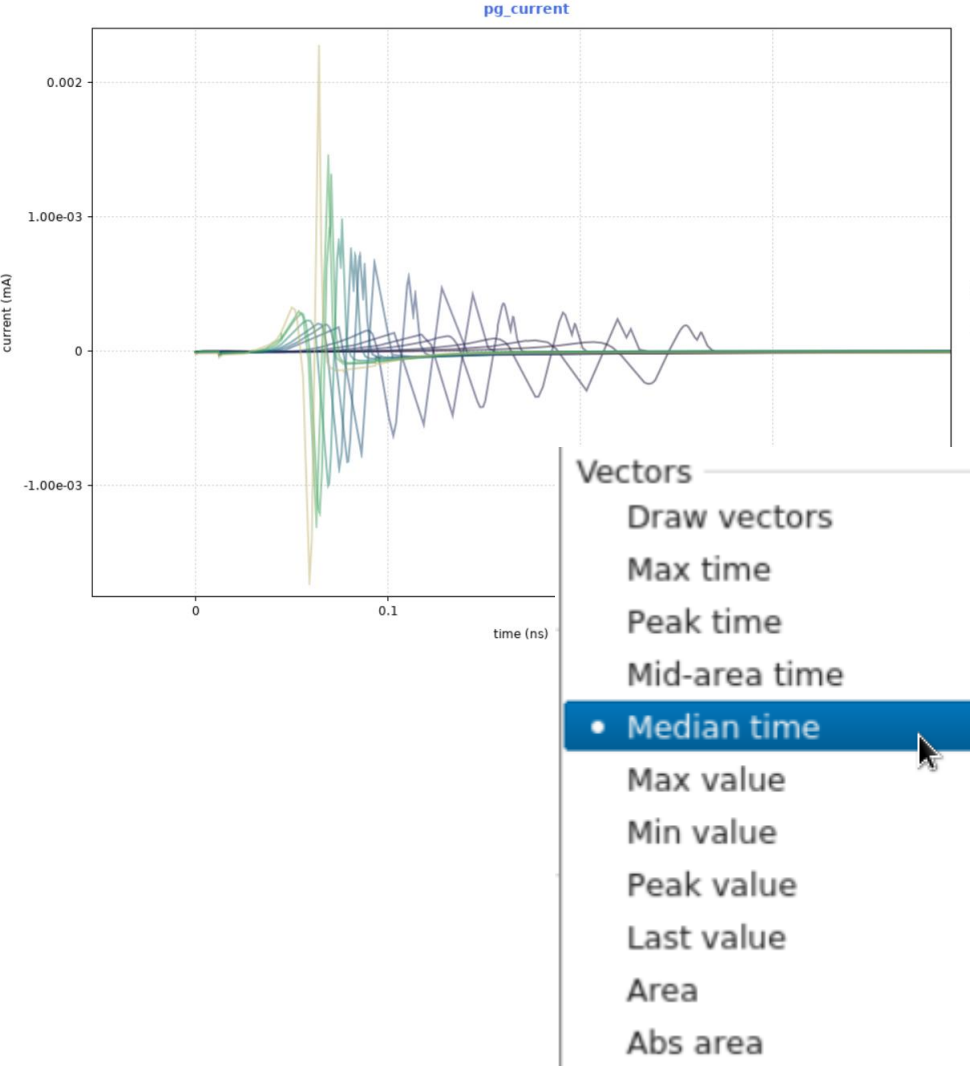
# Area: Pretty Smooth – Models Well



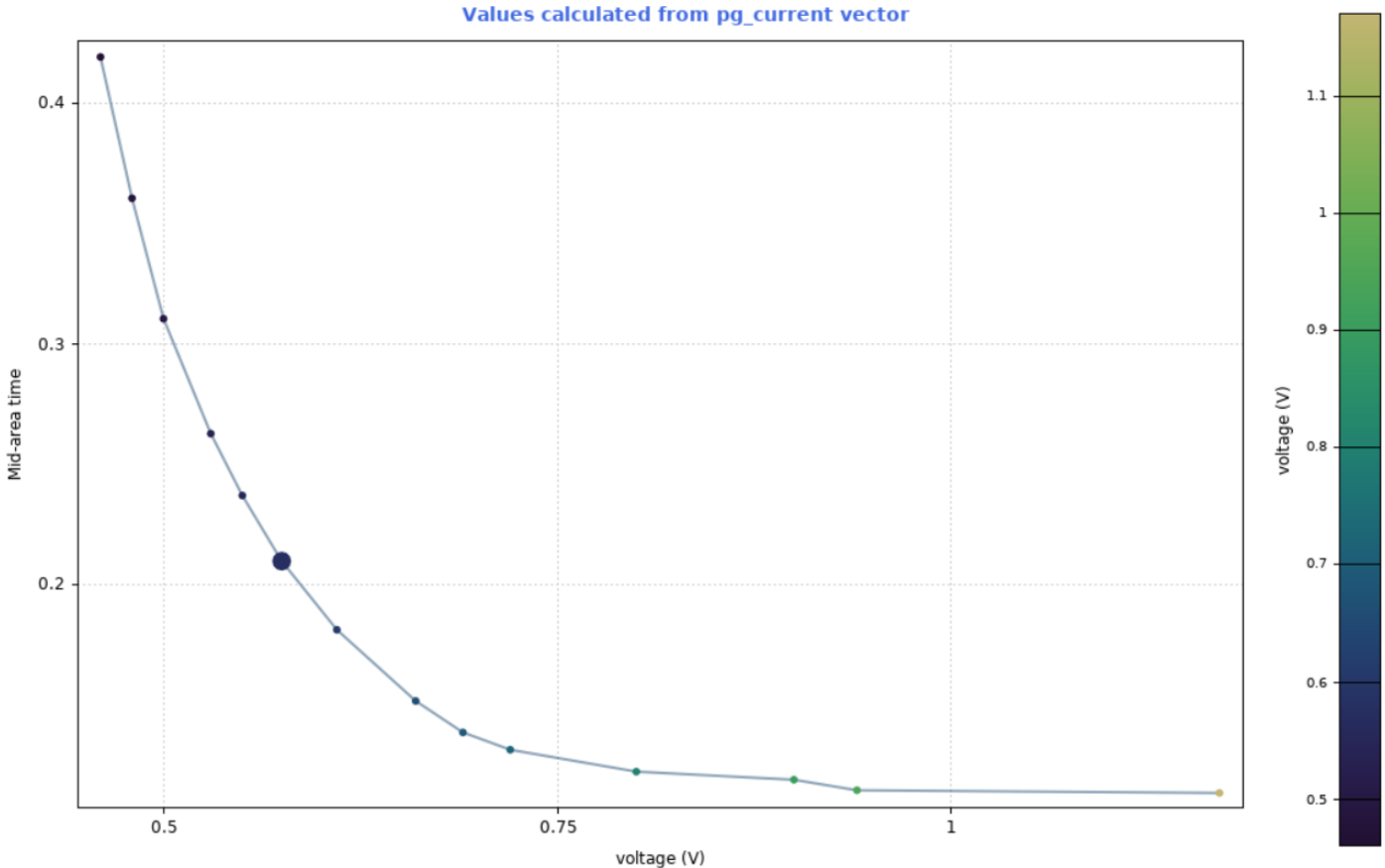
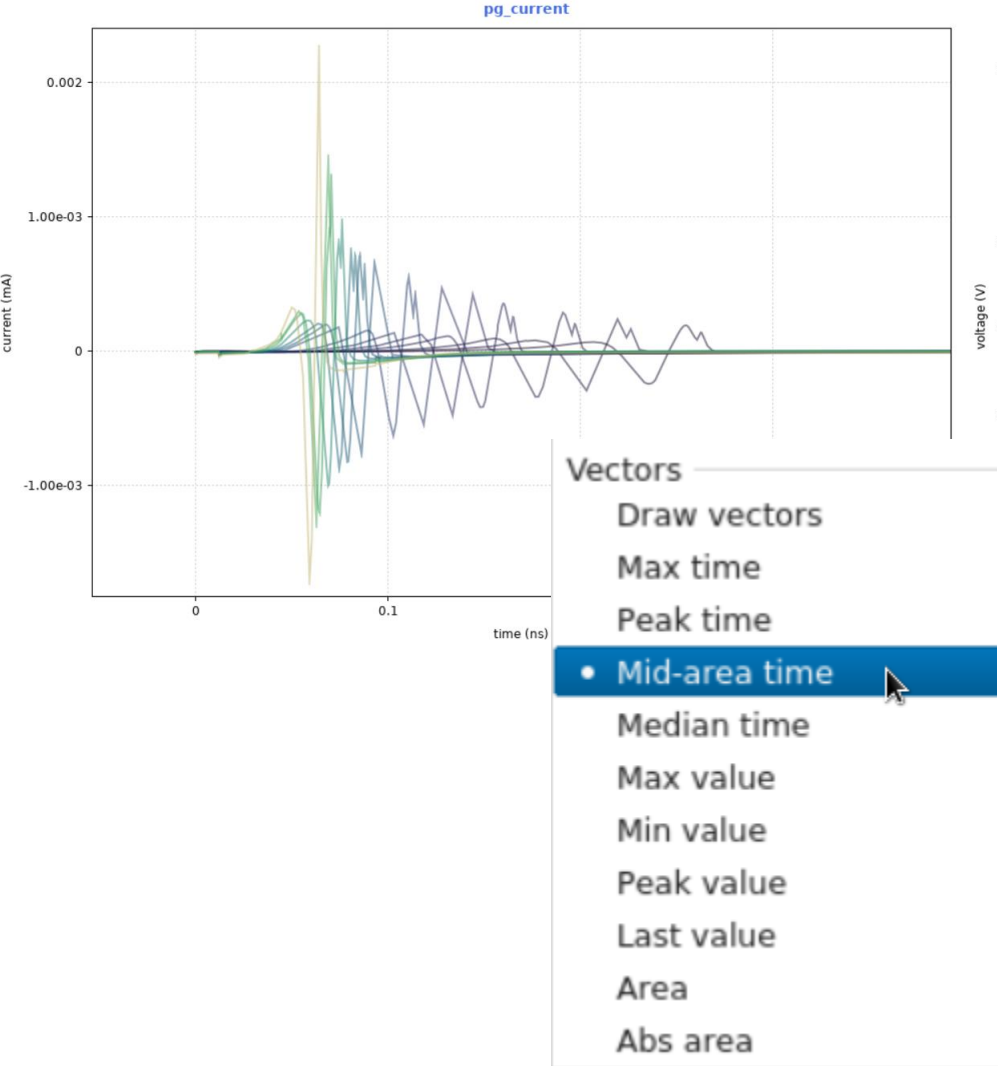
- Vectors
- Draw vectors
  - Max time
  - Peak time
  - Mid-area time
  - Median time
  - Max value
  - Min value
  - Peak value
  - Last value
  - Area**
  - Abs area



# Median Time: Not Bad - A Little Noisy

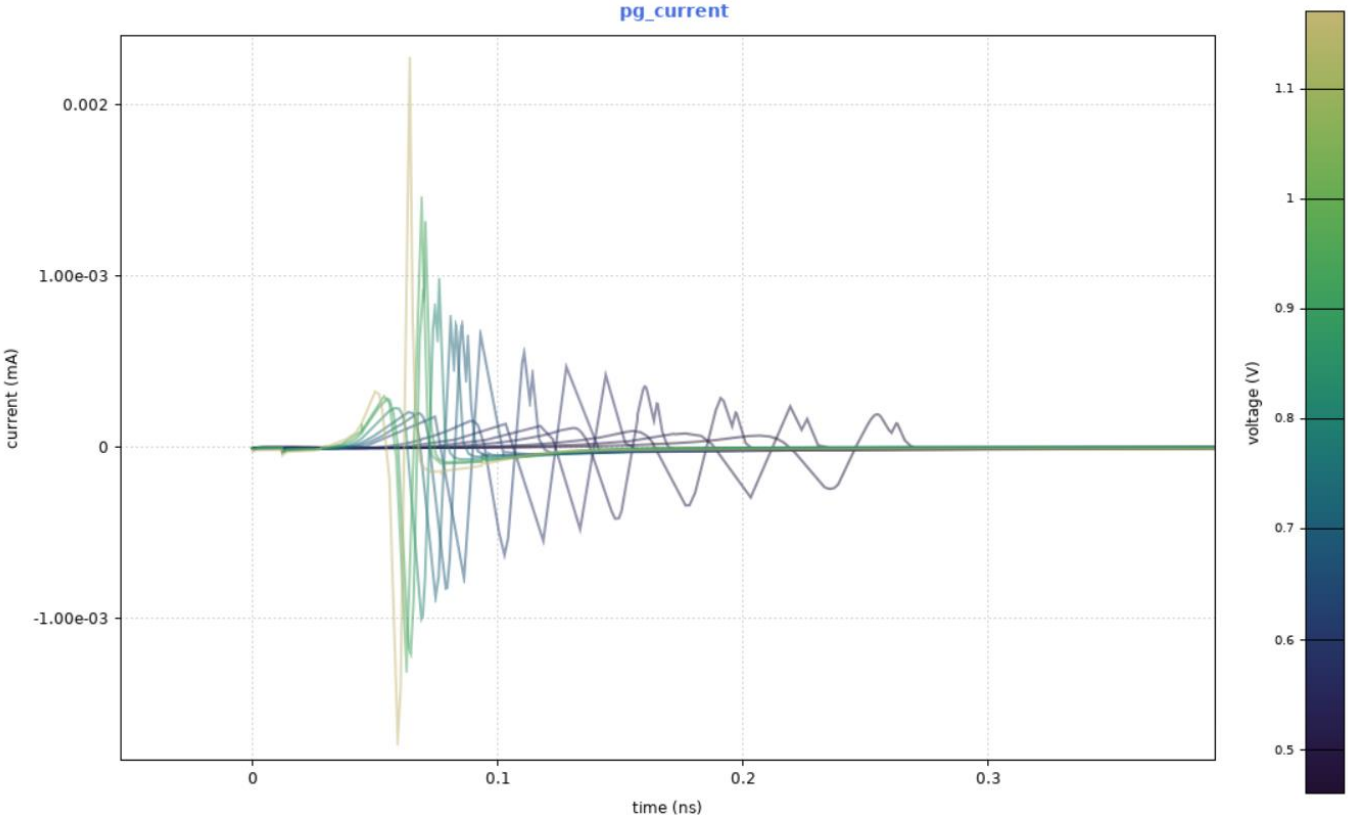


# Mid-Area Time: Smoother and Models Well



# Improvement by Modeling Jagged Waveforms Using Better Parameters

	% In Tolerance (All corners)	% In Tolerance (WC Corners)
Original parameters	~80%	~50%
Best parameters	>99%	>99%



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## Getting the Right Answer With Machine Learning

Levels 3 and 4 are useful for production engineering applications

Leveling up is hard, necessary, and has a ton of great research problems remaining

| Thank you