



# Introducing STAC-ML

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# History

- Driven by financial firms
  - Motivation: market making, hedging, customer pricing, etc.
- STAC-ML working group has refined the original POC idea into a finished benchmark specification
- Tech vendors provided crucial input
- But control ultimately rests with users – i.e., those who must deliver business value from technology in the real world
  - Like all STAC Benchmarks

# Latest Status

- Off to the races!
- The benchmark specifications, test harness, reference implementation, and documentation are released
- 5 vendor implementations are currently underway
- Test Harness engineered to allow end-users to mark their own stacks to market

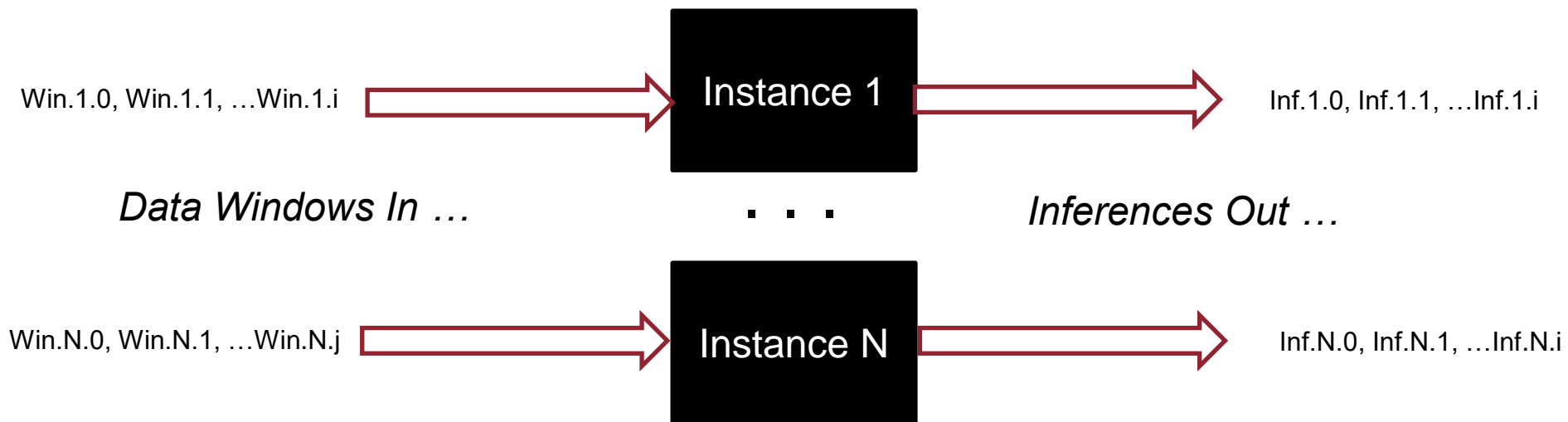
<http://www.STACresearch.com/ml>

# Basics

- LSTM models that simulate real models derived from market data
- Goal: isolate inference performance
  - Inference engine software
  - Underlying processors, memory, accelerators, etc.
  - Anything required to optimally use the former with the latter (e.g., data transfer to processor memory)
- Metrics:
  - Latency, throughput, power efficiency, space efficiency, error
- Benchmarks allow any level of precision (including mixed-precision)

# Scale Dimensions

- Model size
  - Three are currently specified
  - Input data window scales with model size
- Number of Model Instances running in parallel
  - As specified by the SUT provider



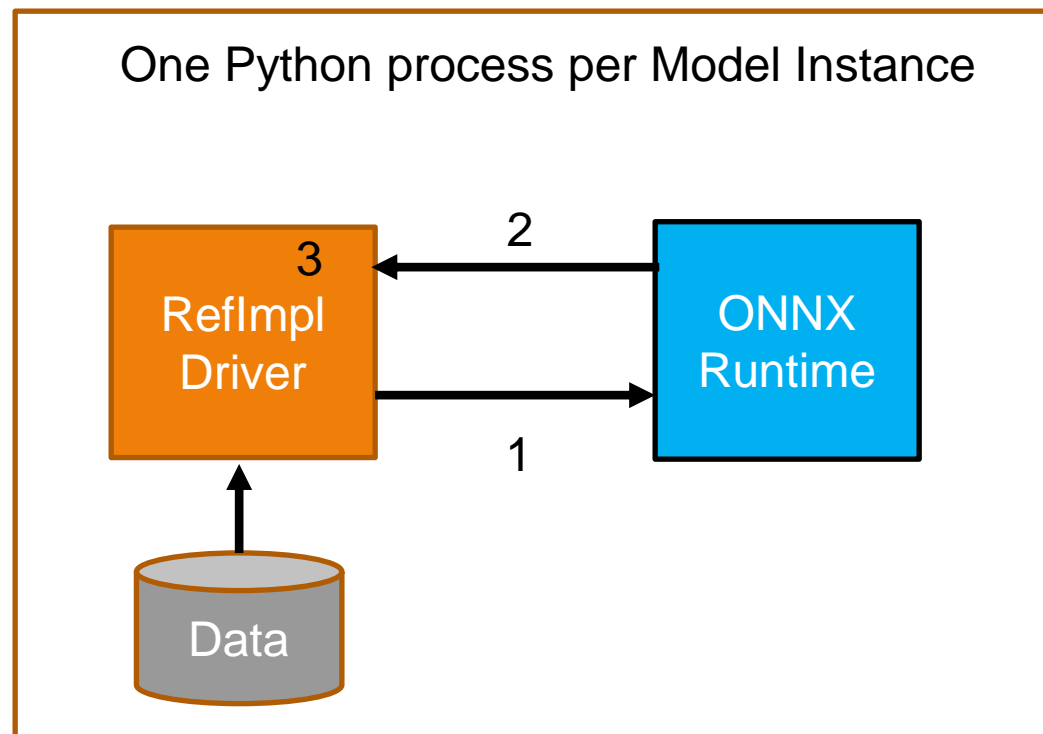
# Use Cases and Optimizations

- Different Use Cases:
  - Trading – Latency Optimization
  - Backtesting – Throughput Optimization
- Optimization tradeoffs (latency vs throughput vs efficiency vs error) are up to the SUT provider
  - The tests collect all metrics every time, no matter the optimization goal
  - Any quantization scheme allowed, if used consistently

# Speaking of Tradeoffs...

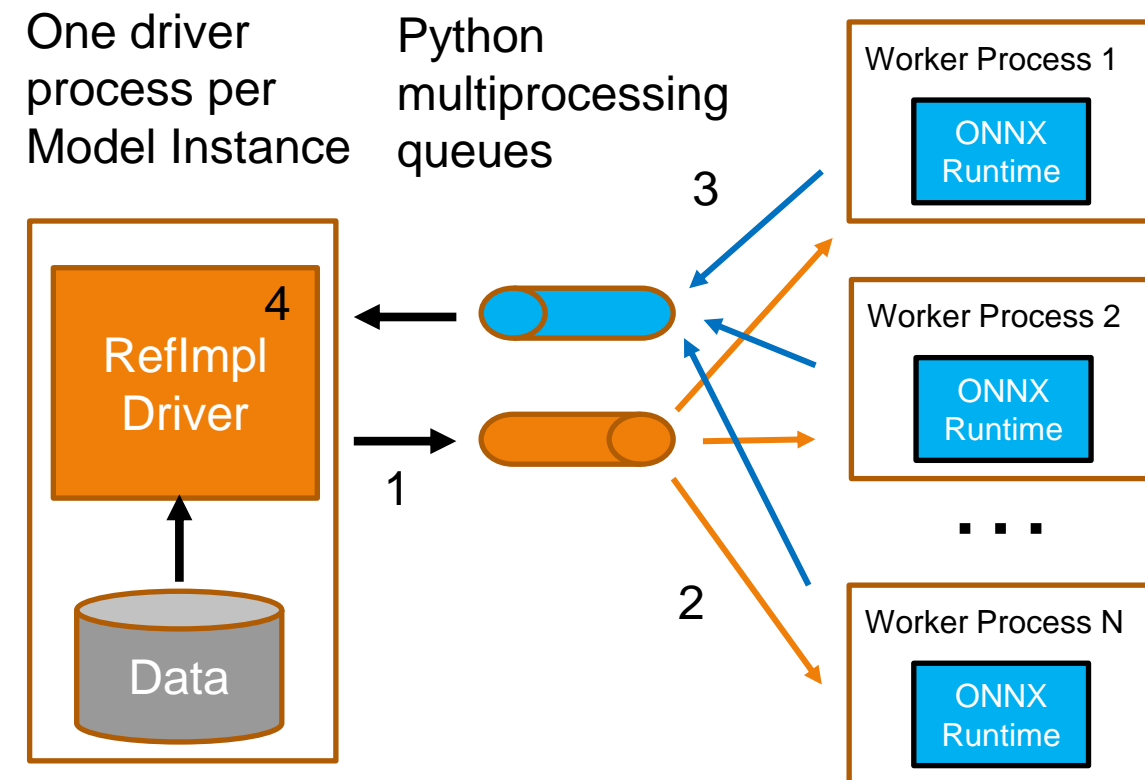
- STAC has just published the first audit reports, representing internal research
- 2 SUTs:
  - Latency Optimized
  - Throughput Optimized
- Same Software (different tuning):
  - Pure Python Implementation
  - STAC-ML Markets (Inference) Naive Implementation
  - Unmodified ONNX 1.11.0 runtime
- Same Hardware:
  - 60-vCPU @ 3.1Ghz, Sole-Tenant cloud instance, 240GiB Memory
- Models: Standard benchmark models at default FP32 precision

# Serial Implementation.



1. Driver calls ONNX runtime with data window; waits for response
  2. ONNX returns inference value
  3. Driver stores value in memory
- Repeat...

# Parallel Implementation



1. Driver inserts data window(s) into queue; waits for response
  2. Workers take windows from queue
  3. Workers store results in queue
  4. Driver gets result from queue and orders it in memory
- Repeat...

Multiple worker processes per Model Instance

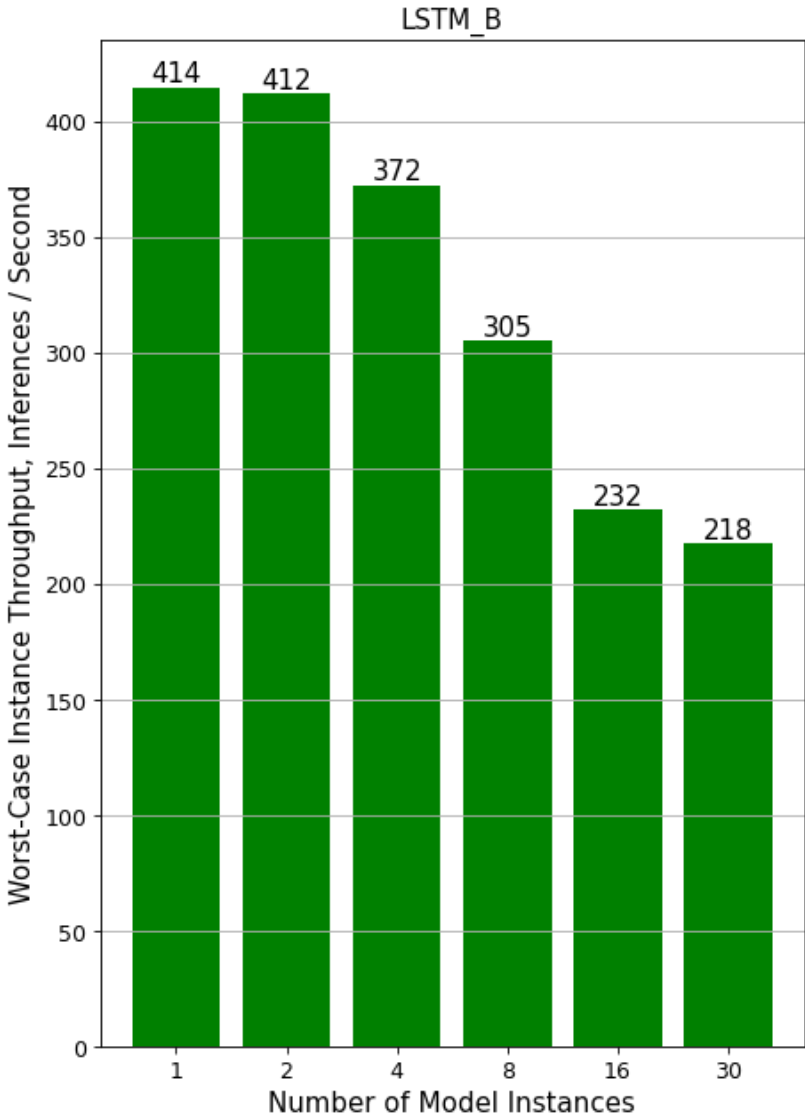
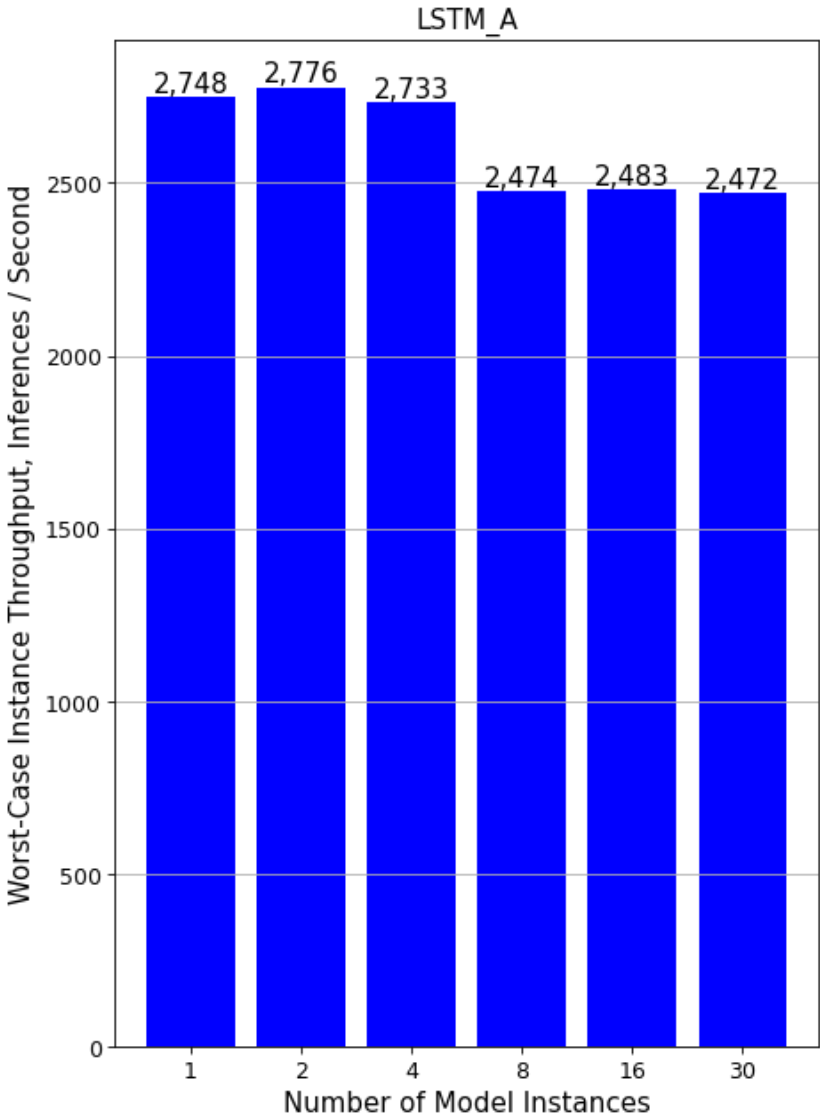


# Optimization 'Knobs'

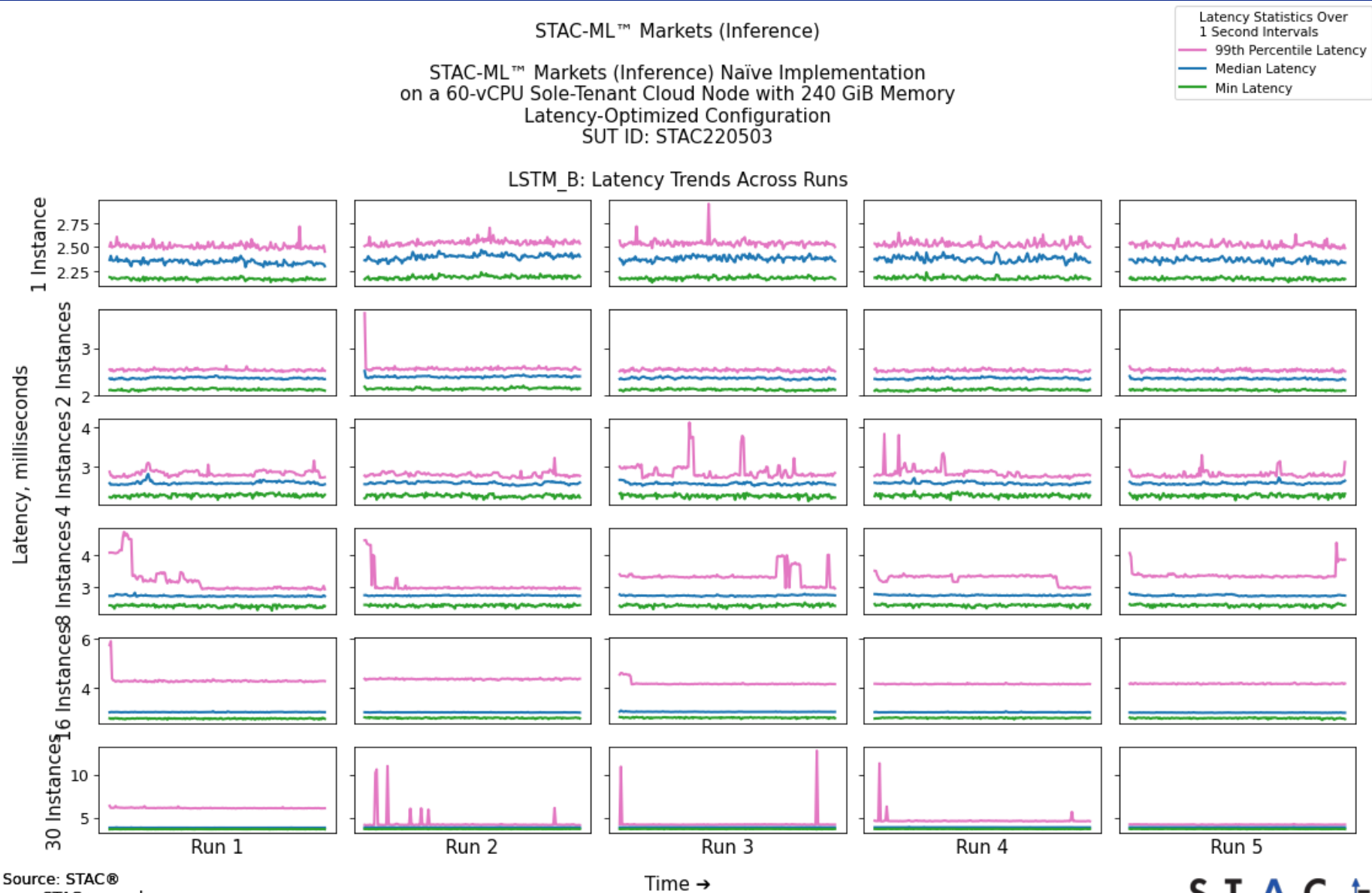
- Number of Model Instances
  - We looked at 1, 2, 4, 8, 16 and 30 instances
- Number of ONNX threads
  - ONNX can (sometimes) effectively utilize multiple threads per model
- Number of Parallel Instances
  - Note: 1 Parallel Instance == Serial Model
  - We have a choice of allocating a HW thread to either an ONNX thread or a Python process
  - The optimal choice again varies by model and optimization goal
- *A Research Note describing the optimization experiment is available in the STAC Vault.*

# Worst-Case Instance Throughput Comparison, LSTM\_A vs LSTM\_B

LSTM\_A vs. LSTM\_B: Instance Throughput of Latency-Optimized Configurations



# Latency Trends



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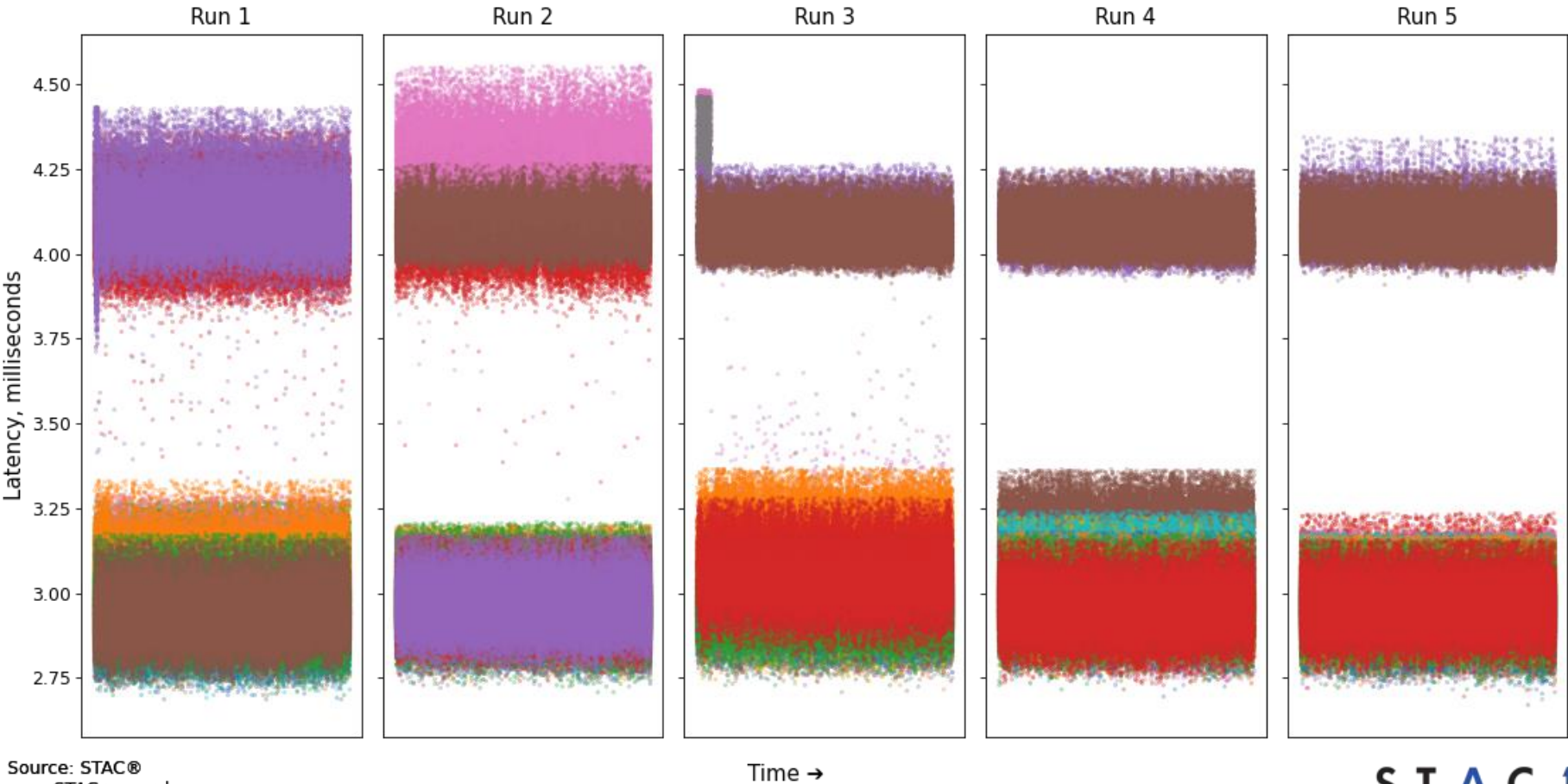
# Latency Details

STAC-ML™ Markets (Inference)

STAC-ML™ Markets (Inference) Naïve Implementation  
on a 60-vCPU Sole-Tenant Cloud Node with 240 GiB Memory  
Latency-Optimized Configuration  
SUT ID: STAC220503

LSTM\_B, 16 Model Instances

Detailed Latency over Time, Samples Above the 99.0th% Quantiles per Model Instance/Run Omitted



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# How to get Access

- All STAC subscribers can access
  - Audited and publicly published STAC Reports
- Premium Subscribers can access
  - Benchmark Specifications
  - Highly detailed configuration information
  - Extensive, detailed visualizations and tables on Performance, Efficiency, and Error
  - Code for test harness, generating post-test visualizations, and STAC Packs
  - Additional reports and research in the confidential STAC Vault™

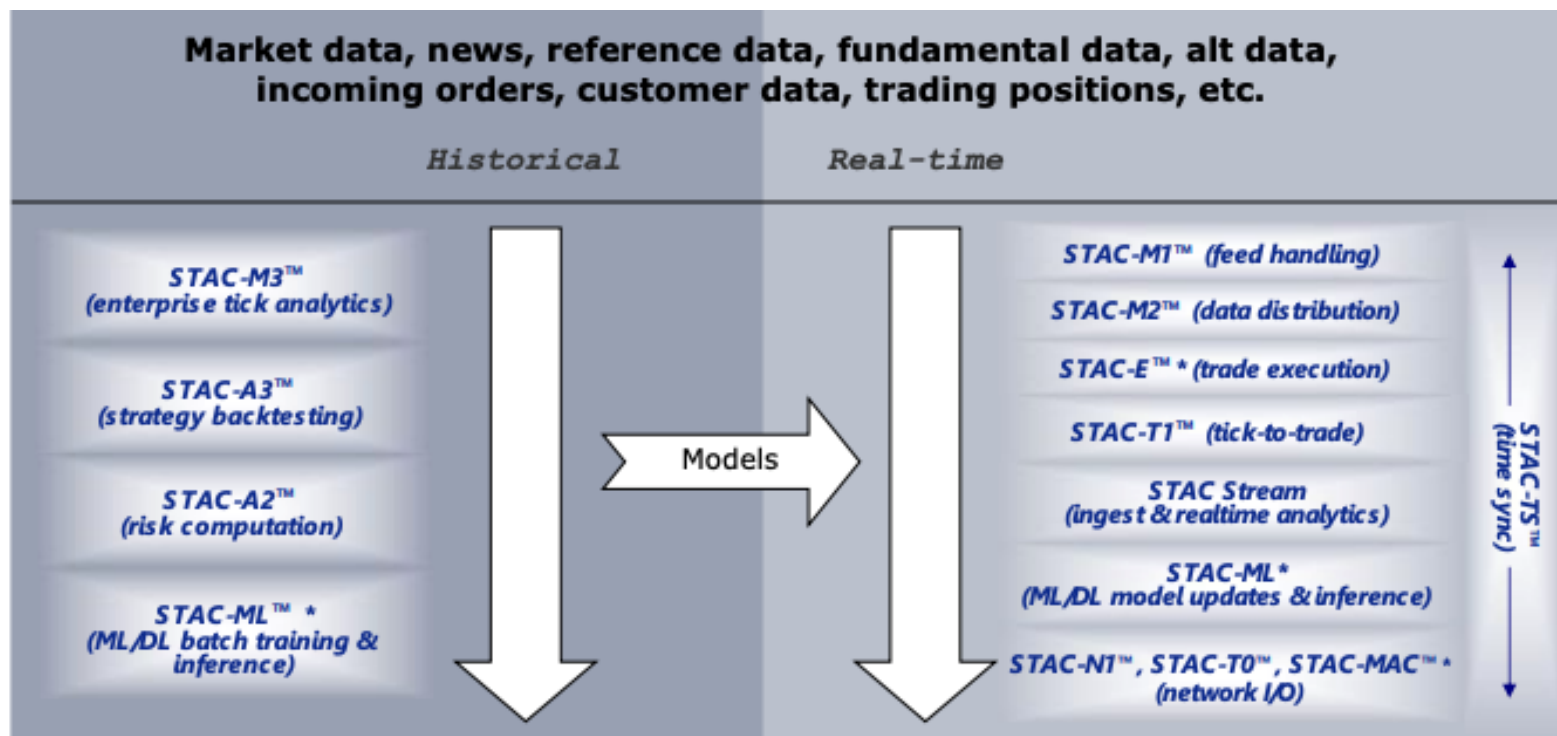
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Tradeflow STAC Track

Analytics STAC Track

Included in both previously existing STAC Tracks

No need to do anything different



# Machine Learning STAC Track

- New STAC Track that includes
  - STAC-ML Markets (Inference)
  - Future STAC-ML benchmarks
- Free trial for the remainder of 2022
  - For those responsible for ML research and infrastructure
  - Full access, including STAC Vault content
  - To request the trial:

[council@STACresearch.com](mailto:council@STACresearch.com)