

STAC-ML Update

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STAC-ML Markets (Training) Benchmark: Underway

- Existing ML training benchmarks are not specific to Finance:
 - They focus on <u>qualitative</u> problems
 - Finance requires good <u>quantitative</u> models
- We spoke to many both inside and outside of the Working Group
- Came back to the Working Group with several candidate use cases
 - Value to the end user
 - The ability to fairly evaluate the quality of benchmark solutions
- Consensus Focus on complex derivative modelling
- Now detailing a proposal Join us!





STAC-ML Markets (Inference)

NEWS FLASH

- First audited results from Groq published today!
- Will get to that shortly...

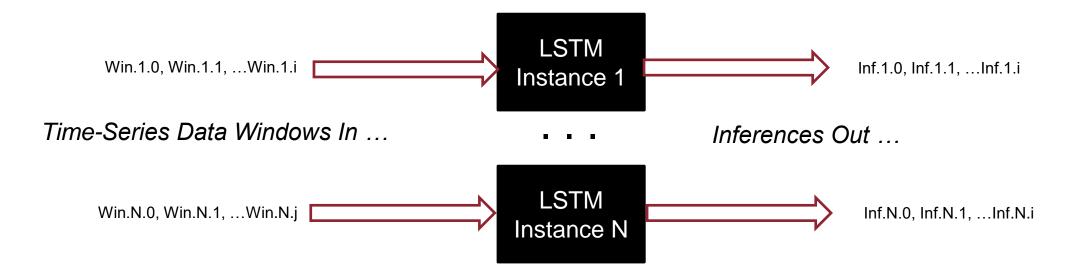


Background - STAC-ML Markets (Inference)

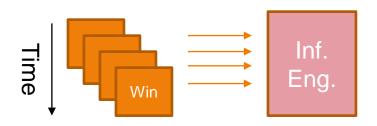
- STAC-ML provides a framework for full-stack evaluation
- Three users of STAC-ML
 - STAC
 - Vendors
 - Financial firms
- I will talk about all three



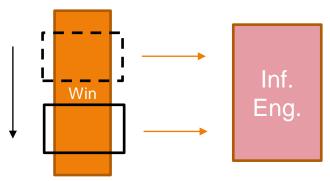
Time-Series Inference using LSTM Models: Perf./Eff./Scalability



Sumaco – Fixed, Unique Window



Tacana - Sliding Window (Streaming)





Information for three use cases

- 1. Latency at any cost
- 2. Throughput for a given latency
- 3. Throughput at any cost



Research Available to ML STAC-Track Subscribers

- GCP Cloud SUT
 - Latency- and Throughput-optimized configurations for ONNX inference
- TensorFlow Performance (on CPU)
 - Looked at different ways to configure TensorFlow for inference
- Azure Cloud-SUT Jamboree (Coming up)
- All research available via free trial for remainder of 2022
 - For those responsible for ML research and infrastructure

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STAC-ML Markets (Inference) Azure Cloud-SUT Jamboree!

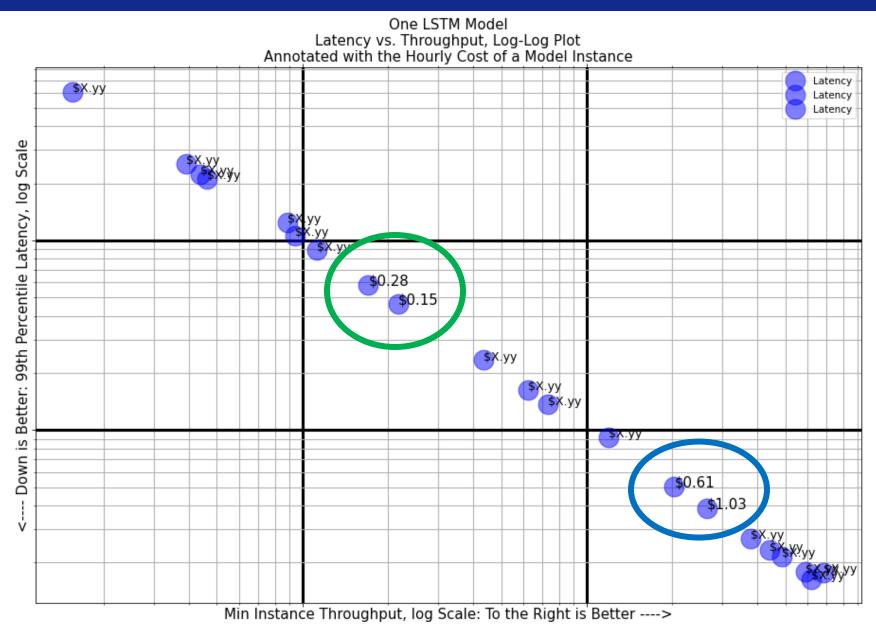
- Goal: compare 3 CPU architectures for inference
 - Intel, AMD, Ampere (ARM)
- Used the STAC "Naive" Python implementation with ONNX
- Tested on Microsoft Azure

Thanks to Microsoft for supporting the STAC community by providing credits for this research!

- Tested two configs for each VM (latency optimized, tput optimized)
- All 6 reports & comparison report will soon be in the STAC Vault
- No vendors participated in the setup and optimization of the SUTs

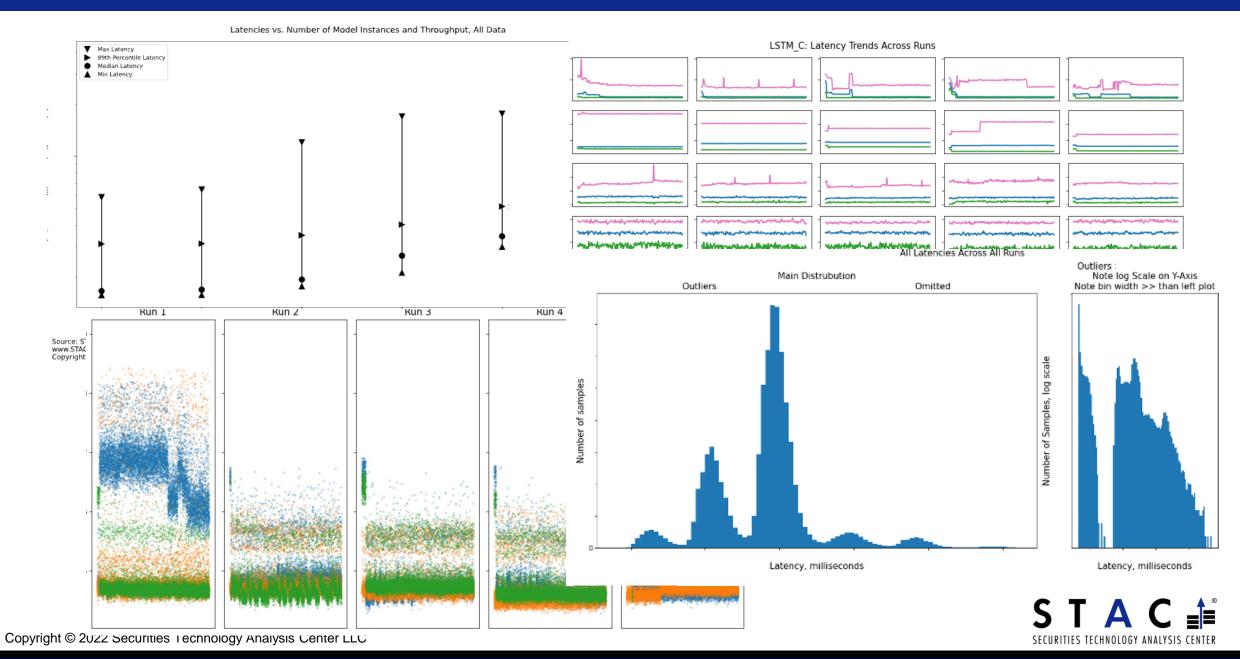


Research Summary Note: Business-oriented comparisons

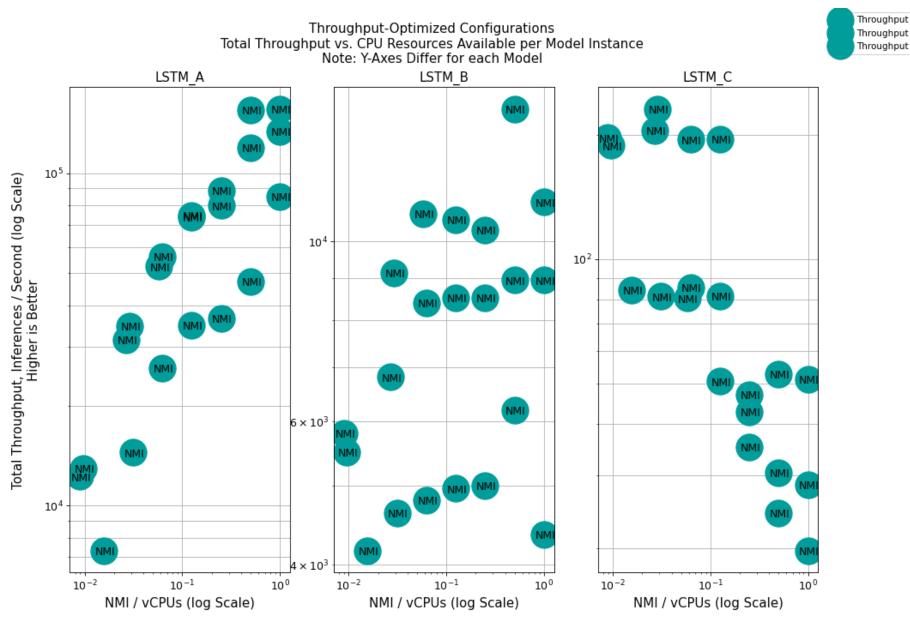




Detailed analysis available for each SUT



Research Summary Note: Throughput Performance Comparisons





First public tested SUT!

- STAC-ML Pack for GroqWare[™] (Rev A)
 - Version of STAC "Naive" implementation adapted for GroqWare™ APIs
- GroqWare[™] SDK 0.9.0.5 devtools and runtime
- Python 3.8.15; NumPy 1.23.4
- Ubuntu Linux 22.04.1 LTS
- GroqNode™ GN1-B8C-ES:
 - 8 x GroqCard[™] 1 Accelerators (GC1-010B)
 - 2 x AMD EPYC™ 7413 24-core CPUs @ 2650 MHz
 - 16 slots x 64GiB DDR4 1024GiB Total



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Results highlights - Groq

- For small model LSTM_A, across 1, 2 and 4 simultaneously running model instances (NMI):
 - Worst case 99th percentile latency was 56.4 µsec
 (STAC-ML.Markets.Inf.LSTM_A.S.4.LAT.v1)
 - 99th percentile latencies varied 1% (from 55.9 to 56.4 µsec)
 (STAC-ML.Markets.Inf.LSTM_A.S.[1,2,4].LAT.v1)
 - The widest spread from minimum to 99th percentile latency was 6% (53.4 to 56.4 μsec) (STAC-ML.Markets.Inf.LSTM_A.S.4.LAT.v1)



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Results highlights - Groq

- For large model LSTM_C, across all NMI tested:
 - Worst case 99th percentile latency was 2.27 ms (STAC-ML.Markets.Inf.S.LSTM_C.8.LAT.v1)
 - 99th percentile latencies varied by 2% (from 2.72 to 2.77 ms)
 (STAC-ML.Markets.Inf.S.LSTM_C.[1,2,4,8].LAT.v1)
 - The widest spread from minimum to 99th percentile latency was 3% (2.68 to 2.77 ms)
 (STAC-ML.Markets.Inf.S.LSTM_C.8.LAT.v1)



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STAC-ML tools are ready for you, too

- Vendor implementations See how it works
- Test harness software and analysis tools Test your own stacks
 - In fact, test your own models!



