



# **STAC Update: Big compute**

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- Non-trivial Monte Carlo
  - Heston-based Greeks for multi-asset, path-dependent options with early exercise
  - Metrics: Speed, capacity, quality, efficiency
- Numerous reports
  - Some public, some in the STAC Vault
- Premium STAC members get:
  - Reports in STAC Vault
  - Detailed config info on public and private reports
  - Code from vendor implementations of the benchmarks

[www.STACresearch.com/a2](http://www.STACresearch.com/a2)

# Intel 2S Server System S9200WK with Cascade Lake-AP CPUs

- SUT ID: INTC190903
- STAC Pack for Intel Parallel Studio XE (Rev M)
- Intel Server System S9200WK
- 2 x 48-core Intel Xeon Platinum 9242 (Cascade Lake-AP) processors
- RHEL 7.7
- 384GB DRAM



[www.STACresearch.com/INTC190903](http://www.STACresearch.com/INTC190903)

# Space efficiency vs. all other publicly disclosed results

- Solution based on Cascade Lake-AP set a new record for space efficiency (STAC-A2.β2.HPORTFOLIO. SPACE\_EFF)
- This beats the previous record, also held by Intel (SUT ID INTC181012), by 8.5%



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# Efficiency vs. most recently benchmarked using GPUs

Compared to the most recently benchmarked solution using GPUs (SUT ID NVDA181105), this system:

- Had over 1.8x the space efficiency (STAC-A2.β2.HPORTFOLIO.SPACE\_EFF)
- Was within 20% of the energy efficiency (STAC-A2.β2.HPORTFOLIO.ENERG\_EFF)



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# Versus a “non-AP” Cascade Lake based solution

Compared to a similar 2S system using Platinum 8280 (“Cascade Lake”) processors (SUT ID INTC190402), this system had:

- 2.67x the space efficiency  
(STAC-A2.β2.HPORTFOLIO.SPACE\_EFF)
- 1.84x the throughput  
(STAC-A2.β2.HPORTFOLIO.SPEED)
- 1.35x the energy efficiency  
(STAC-A2.β2.HPORTFOLIO.ENERG\_EFF)



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# AI Benchmark PoCs

- **Deep time series (training)**
  - Small number of largely homogeneous data types per symbol
  - Long, dense time series
- **Wide time series (training)**
  - Large collection of heterogeneous data types per symbol
  - Often have regular but different frequencies
- **Entity matching (inference)**
  - Match descriptions of companies to unique identifiers
- **NLP (training)**
  - Topic modeling of business description in annual report required from U.S. firms
  - STAC Study in Vault: “*Scaling a common machine learning workload in the cloud*”
  - Excerpts on STAC website: [STAC Study Excerpts - NLP topic modeling 2018](#)

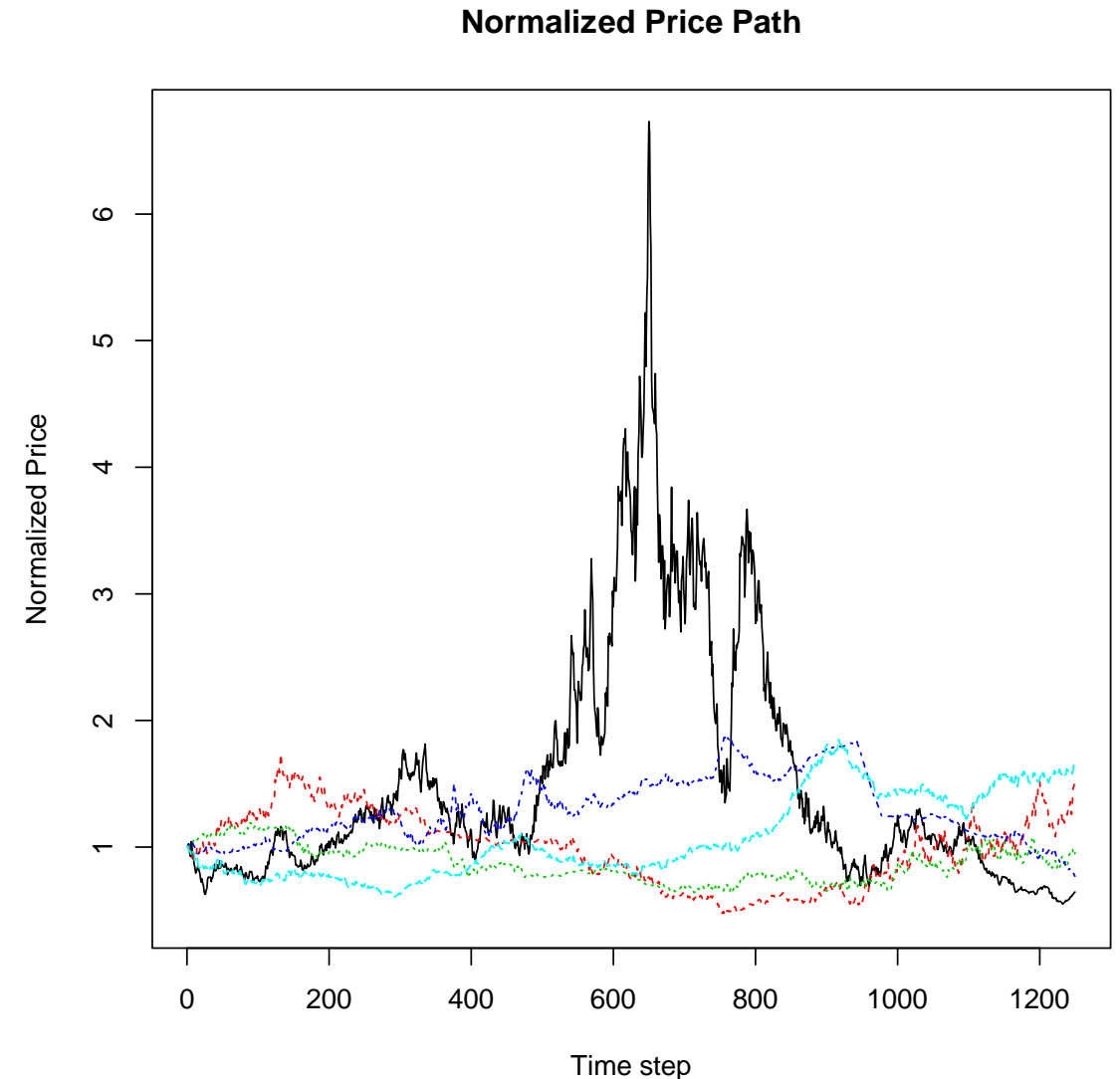
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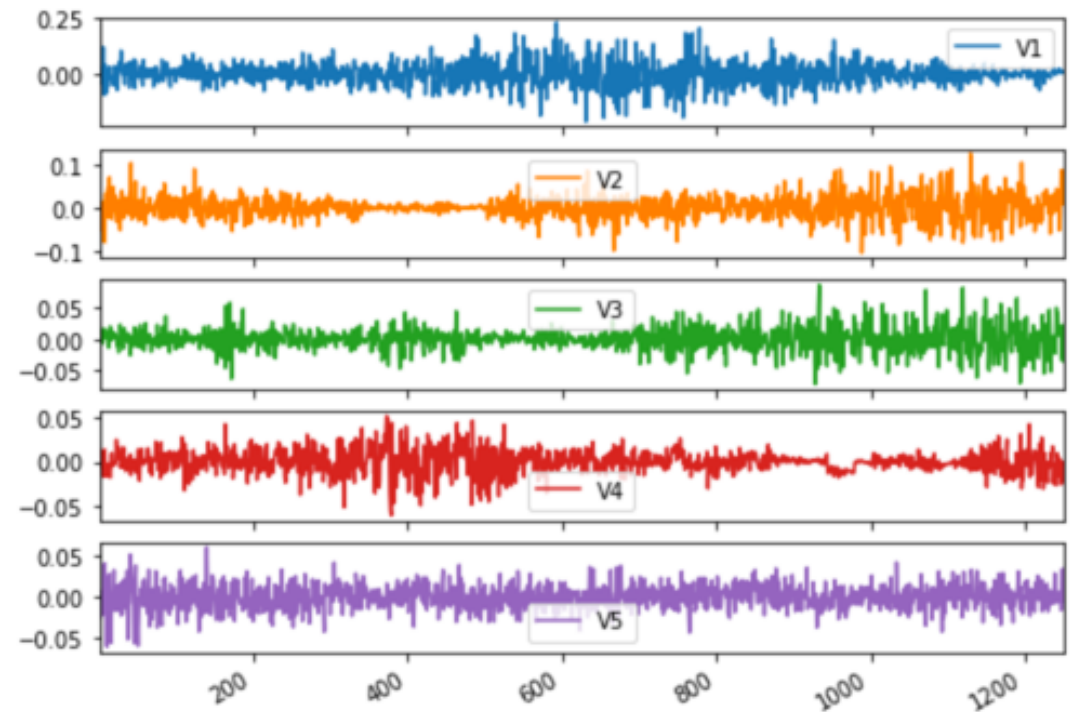
# AI benchmarks in finance: the data problem

- Core challenge with real data
  - Don't know underlying "real" signal
  - Neural net models are opaque
  - How to judge relative performance?
- Simulated data advantages
  - Specify signal
  - Vary attributes e.g. signal:noise ratio
  - Compare model performance to reality



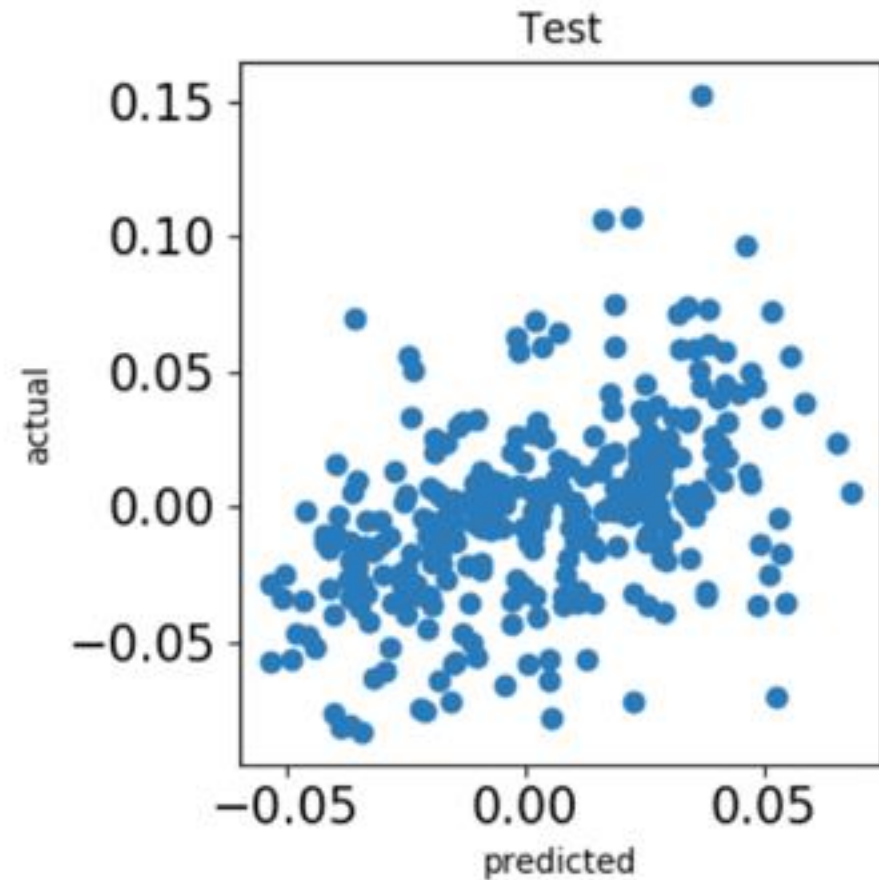
# Deep Time Series: Status of crawl, walk, run approach

- Leveraged existing STAC-A2 path generation
- Generate simulated market data with known stochastic properties
- Apply random correlation matrix
- Overlay simple signals
- Can this data be modeled?



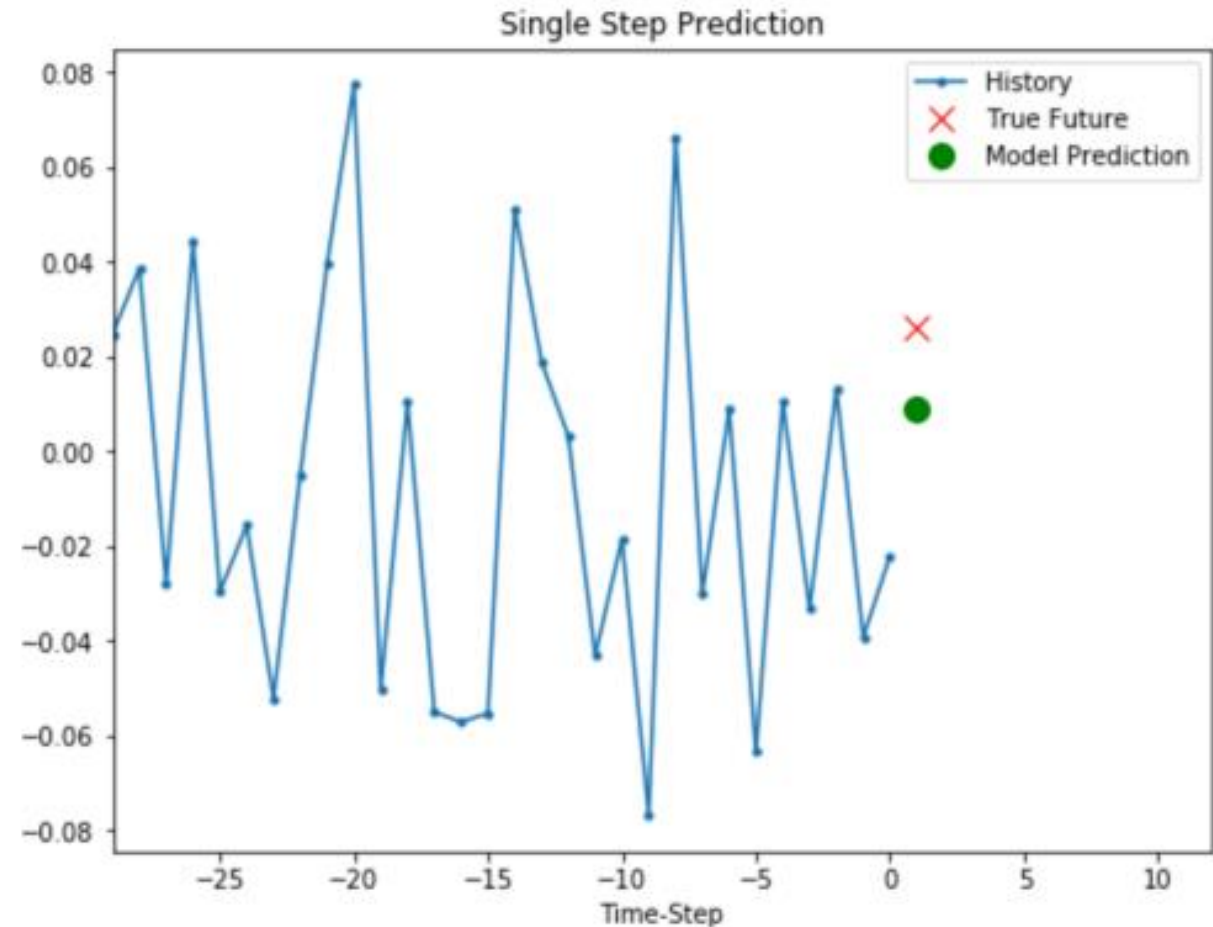
# Deep Time Series: Status of crawl, walk, run approach

- On “tiny” problem, an AutoML tool finds models with tunable accuracy
- First does automated feature engineering
- Runs for a couple of hours



# Deep Time Series: Status of crawl, walk, run approach

- Preliminary work with LSTM in Tensorflow
- Experimenting with attributes of the overlaid signal
- Experimenting with attributes of the model
- Results will be provided to the Working Group



# Deep Time Series: Next steps and call to action

- We have data generator and outlines of the work load
- Will refine in the next couple of months
- Will post outstanding issues and ways to help in online forum
- To join the STAC-AI Working Group:
  - Go to [www.STACresearch.com/ai](http://www.STACresearch.com/ai), right side of the page
  - If you see the “Group Members” list, you’re already in it
  - If not, click on the “Enable Me” button

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