

STAC Update - STAC-AI AI STAC (NYC)

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Goals for the next 25 minutes

- (Re-)Introduce the STAC-AI™ LANG6 (Inference-Only) benchmark
 - Where it came from;
 - What is does and why it's different;
 - Where it stands and where it's going
- Discuss the results of the STAC240903a/b and STAC242211a/b audits
- Cover how you and your organization can be involved!



The Critics Have Spoken: STAC-AI™ LANG6 (Inference-Only) is a hit!

Thank you for sharing your work - it was genuinely exciting to see such thoughtful performance engineering being applied to LLM systems. Your benchmark sets a new standard for what meaningful LLM performance characterization should look like.

www.claude.ai November 15, 2024



Origin: LLM Basics

Large Language Models are AI systems trained on vast amounts (trillions of words) of text and other
data, whose purposes are to understand, generate and manipulate human languages of all types natural, programming, mathematical etc.

Applications:

- Text (code) completion, generation, optimization
- Translation and summarization
- Question answering and chatbots

BB What is the best known STAC Benchmark?

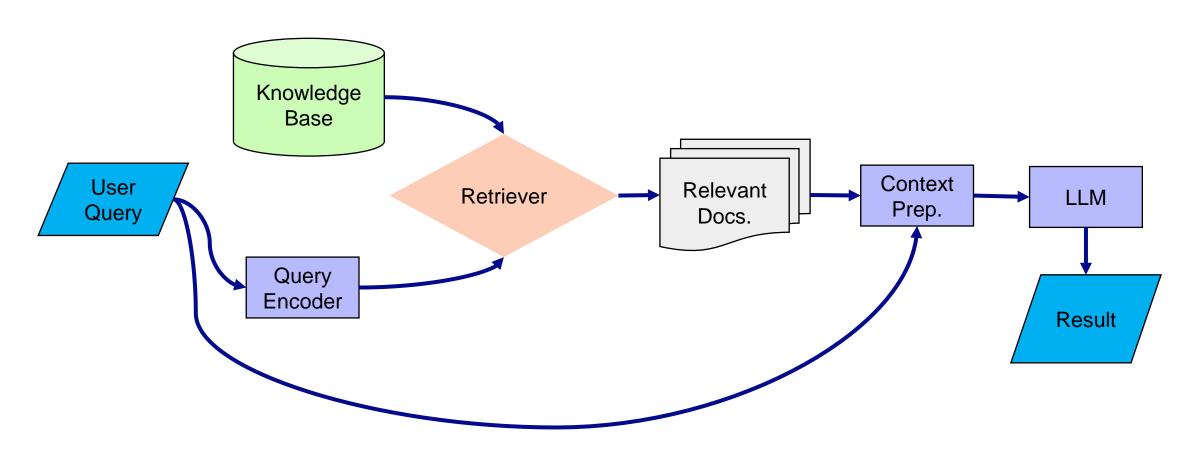
STAC-A3 is one of the most well-known STAC benchmarks. It focuses on financial analytics, specifically measuring the performance of risk calculations and Greeks computations for options pricing models.

- Limitations:
 - Lack of true understanding or reasoning
 - Biased and / or incorrect outputs
 - No real-time or task-specific knowledge updates



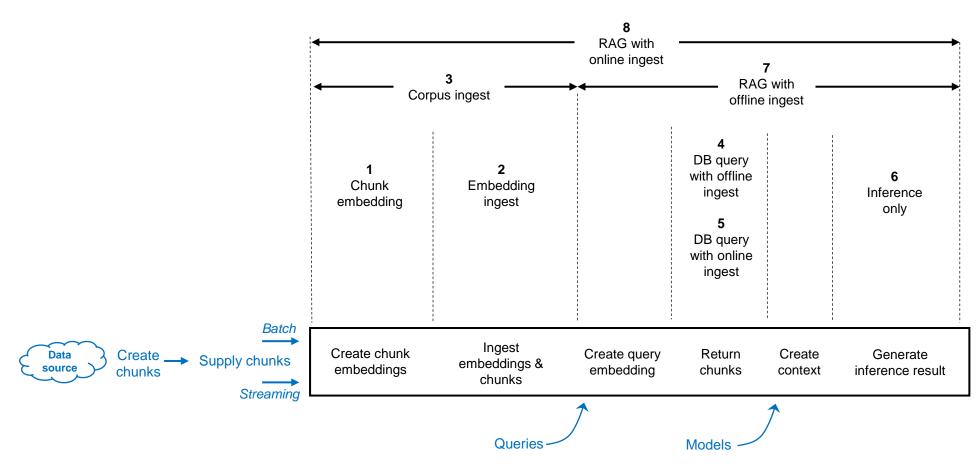
Retrieval-Augmented Generation (RAG) Pipeline

Augments the LLM's general knowledge with evidence specific to a particular query





STAC-AI™ RAG Benchmarks Landscape





STAC-AI™ LANG6 (Inference-Only) Overview

- STAC-AI[™] LANG6 (Inference-Only) models the LLM server-side of a RAG application [Or generic LLM server]
 - Does not include any interaction with clients (no external networking)
 - RAG retrieval has already been accomplished [if necessary]
 - Input data and output results remain on the server
 - Current benchmark models are Llama-3.1-8B / 70B –Instruct
 - Expect new models to be approved by WG periodically
 - Official benchmark data sets are based on the analysis of EDGAR filings







STAC-AI™ LANG6 (Inference-Only) Status

- Specification Rev. D has just been published
- A Rev.-D compatible Test Harness is available to subscribers
- STAC has completed 4 internal audits on the Paperspace GPU cloud
 - Public reports will be published soon!

- NOTE: STAC-AI™ LANG6 (Inference-Only) and STAC-ML™ Markets (Inference) are almost unrelated
 - STAC-ML models a low-latency ML-based trading application
 - Market Data → Mechanical Trading Decision
 - STAC-All currently models high-level workloads for financial document analysis
 - Documents → Insights



How STAC-AI™ LANG6 (Inference-Only) Differs from Other LLM Benchmarks

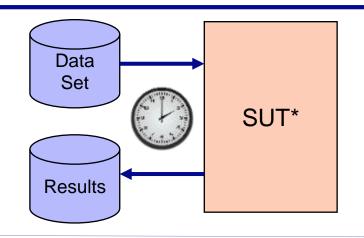
- This is an infrastructure performance benchmark, not a data science challenge
 - An aid to capacity planning, cost estimation, etc.
- We focus on realistic workloads from the financial domain
 - Not toy examples
- The metrics are business-oriented and human-oriented, not LLM-architecture oriented
- STAC provides detailed tabulations and visualizations of results
 - Not a single-metric 'leaderboard' presentation



Request Modes

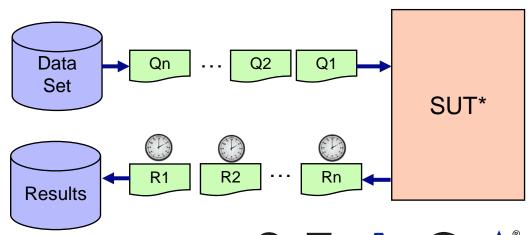
Batch Mode:

- The entire Data Set is processed and timed in one go
- Essentially 2 inference performance metrics:
 - Throughput: Overall words per second generated
 - Inference Rate: Overall inferences per second



Interactive Mode:

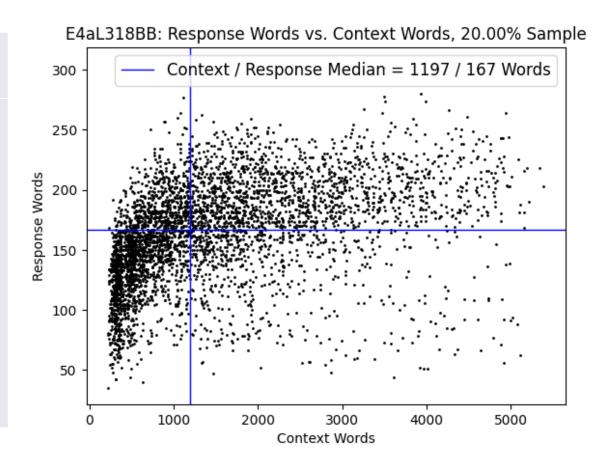
- Models a Poisson (random) arrival process with mean arrival rate λ
- SUT streams the output
- Tests may include multiple λ
- Many inference performance metrics





Data Set: EDGAR4a/b*

Data Set	Prompt type	Document type
EDGAR4	Summarization of the relationship of a company to one of various physical and financial concepts such as commodities, currencies, interest rates and real estate sectors.	EDGAR 10-K paragraphs from a single security 10-K filing, selected by RAG retrieval. Prompts are generated for each of 5 preceding years, for symbols in the current Russel 3000 index.

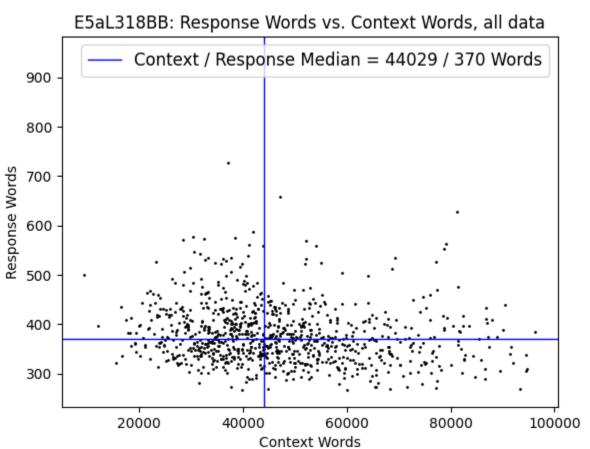




^{*} EDGAR4a is for Llama-3.1-8B-Instruct; EDGAR4b for the 70B model

Data Set: EDGAR5a/b*

Name	Prompt type	Document type
FDGAR5	A set of questions covering several different aspects of a complete 10-K filing.	Complete text of an EDGAR 10-K filing for randomly selected Russel 3000 symbols from one of the randomly selected last 5 years. [Not a RAG Workload, per se]

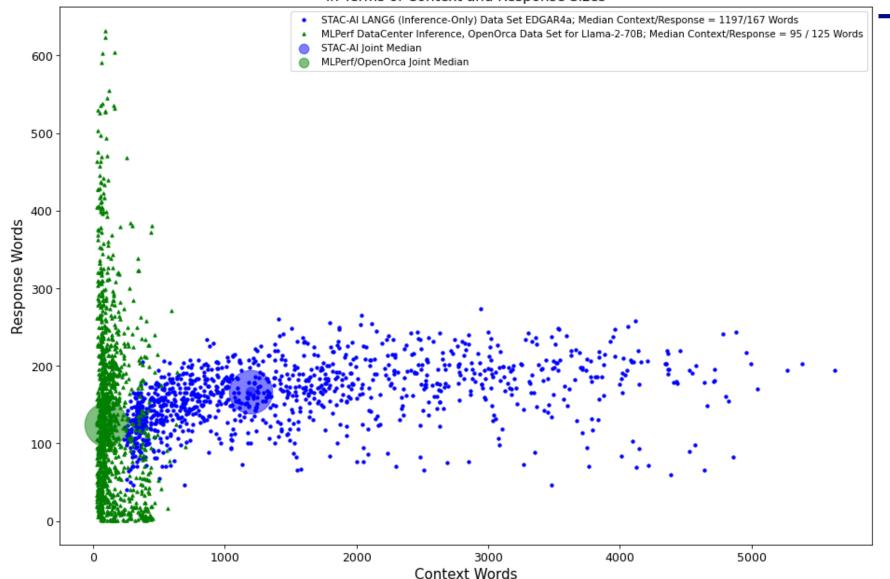




^{*} EDGAR5a is for Llama-3.1-8B-Instruct; EDGAR4b for the 70B model

Comparison with MLPerf (Llama-2-70B / OpenOrca)

Comparing STAC-Al LANG6 (Inference-Only) EDGAR4a and MLPerf OpenOrca Data Sets in Terms of Context and Response Sizes



We measured batch inference rates on NVDIA A100 GPUs: The inference rate and throughput of the less challenging MLPerf OpenOrca Data Set are more than 5x the rate of the STAC-AI™ data set on Llama-3.1-70B-Instruct

Note: OpenOrca was designed as a data science challenge, not as a performance benchmark.

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Metrics Illustrations Follow: STAC240903a/b and STAC241122a/b Audits

- Paperspace cloud
- STAC Reference Implementation; vllm/vllm-openai:v0.5.5 container
- Ubuntu 22.04; Xen Hypervisor
- STAC240903a/b
 - 8 x NVIDIA A100-SXM4-80GB GPUs
 - 2 x Intel® Xeon® Gold 6342 CPUs + 708GiB Memory
- STAC241122a/b
 - 8 x NVIDIA H100 80GB HBM3 GPUs
 - 2 x Intel® Xeon® Platinum 8458P CPUs + 1.6TB Memory
- SUTs
 - a: Llama-3.1-8B-Instruct, BF16
 - b: Llama-3.1-70B-Instruct, BF16

No vendors participated in these benchmarks.

All cloud services were purchased by STAC at standard retail pricing.

STAC does not endorse any commercial hardware or software product or service.



GPU Configurations

	H100	Batch Configurations			
Model	Workload	Max Context, Tokens	GPUs / Model Instance	Model Instances	Batch Workers
Llama-3.1-8B-Instruct	EDGAR4a	10K	1	8	128
Llama-3.1-8B-Instruct	EDGAR5a	128K	1	8	8
Llama-3.1-70B-Instruct	EDGAR4b	10K	2	4	32
Llama-3.1-70B-Instruct	EDGAR5b	128K	4	2	2

Note: All but the optimal number of Batch Workers were identical between the A100 and H100 in our testing.

Interactive parallelism is driven by the Interactive arrival rate.



STAC240903a: Batch Report Card

SUT ID: STAC240903a

Batch Report Card

* = STAC-AI.LANG6.[Model].[Data Set]

Model	Llama-3.1-	Llama-3.1-8B		
Data Set	EDGAR4a	EDGAR5a		
SUT Variant	L318Bm10KB	L318BB		
*.BATCH.INF_RATE.v1				
Inference Rate	24.0	0.431		
Inferences / sec				
*.BATCH.TPUT.v1				
Throughput	3,917	164		
Words / sec				
*.BATCH.LOAD.v1				
Load Time	73.5	74.5		
seconds				
*.BATCH.FIDELITY.v1	98.76%	97.90%		
Fidelity, %	98.7078			
*.BATCH.HOUR_EFF.v1				
Hourly Efficiency	554.2K	23.22K		
Words / USD				

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STAC241122a: Interactive Report Card

Interactive Report Card

* = STAC-AI.LANG6.[Model].[Data Set]

Model		Llama-3.1-8B				
Data Set	EDG	AR4a	EDG	EDGAR5a		
SUT Variant	E4aI		E5aI			
Lambda	44.0	33.0	0.820	0.600		
*.INTERACTIVE.TPUT.v1						
Throughput	7,102	5,351	303	227		
Words / sec						
*.INTERACTIVE.REACT.v1						
Median Reaction Time	0.0889	0.0710	9.92	5.97		
seconds						
*.INTERACTIVE.RESP.v1						
Median Response Time	10.2	4.15	32.7	15.5		
seconds						
*.INTERACTIVE.OUT_RATE.v1						
5p Output Rate	11.7	32.8	10	16.1		
Words / second						
*.INTERACTIVE.OUT_PROF.v1	10.7	20.0	0.71	10.0		
5p Output Profile Words / second	10.7	30.0	9.71	12.3		
*.INTERACTIVE.LOAD.v1 Load Time	101	101	94.5	94.5		
seconds	101	101	94.5	94.5		
*.INTERACTIVE.FIDELITY.v1						
Fidelity, %	98.54%	98.65%	96.00%	97.69%		
*.INTERACTIVE.HOUR_EFF.v1						
Hourly Efficiency	537.2K	404.7K	22.90K	17.17		
Words / USD	J37.2K	404.78	22.30K	1/.1/5		

Small reductions in interactive arrival rates are paid back with much larger improvements in the user experience.

(Or in performance when chaining operations)

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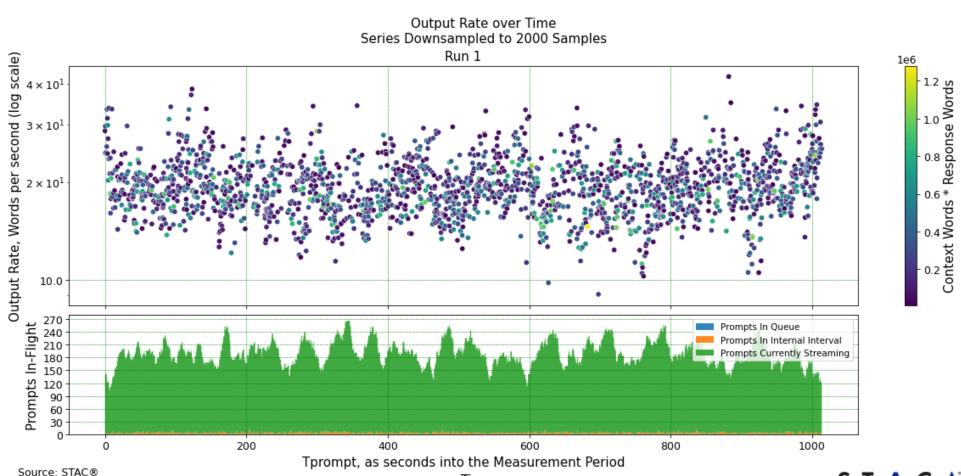
STAC240903a/E4aL318BI: Output Rate over Time

STAC-AI™ LANG6 (Inference-Only)

STAC-Al™ Reference Implementation for vLLM OpenAl Server on 8 x NVIDIA A100-SXM4-80GB GPUs in the Paperspace Cloud Running Llama-3.1-8B-Instruct

SUT ID: STAC240903a

Model: Llama-3.1-8B Data Set: EDGAR4a SUT Variant: L318Bm10KI $\lambda = 21.5$

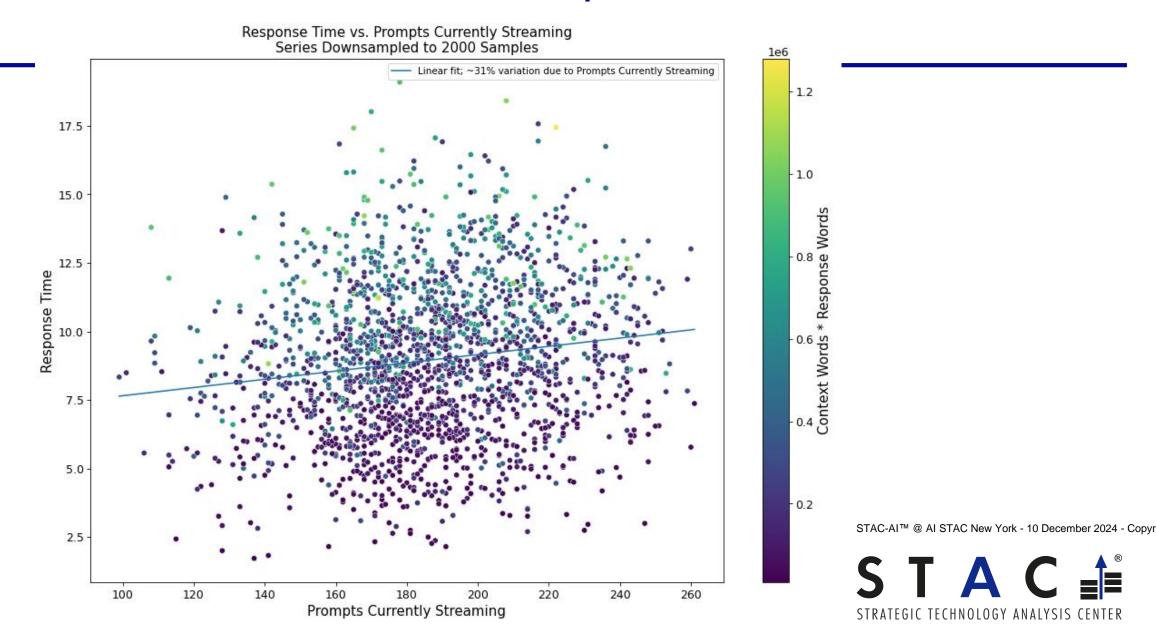


Time →

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STAC240903a/E4aL318BI: Response Time vs. Business



Key Comparisons between H100 and A100 as Observed in These Tests

- 8B Model Batch Mode
 - H100 averages 2.1x the inference rate and throughput of A100
 - H100 averages 1.1x the price-performance of A100
- 70B Model Batch Mode
 - H100 averages 2.4x the inference rate and throughput of A100
 - H100 averages 1.3x the price-performance of A100



Using STAC-AI™ LANG6 (Inference-Only)

- Dual Use: Public / Vault Reports; Private Testing
- Public Reports:
 - Compare vendor-optimized SUTs
- Vault Reports:
 - Vendor results
 - STAC research
- Private Testing
 - Latency-efficiency-throughput tradeoffs for deployment sizing
 - Public cloud vs. API-cloud vs. on-prem costs
 - Large-Language-Models as a Service:
 - Time of day and / or regional effects
 - Adherence to SLAs?



Possible Future Directions

NB: The path forward always depends on input from the Working Group!

- Implement other benchmarks from the RAG pipeline
- New representative LLM Workloads
- New quality metrics
- Training / fine-tuning benchmarks
- Multi-modal inference
- Multi-model inference (Agents?)



How to get involved

Join the working group



STACresearch.com/ml



from this domain, or if you would like to participate in this group, please click the button below.

Enable me! >

2 Access the test harness



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