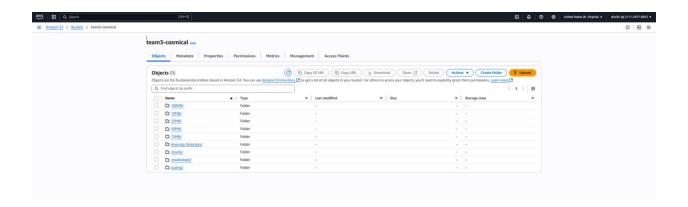
TEAM 3

Nikpour Bardia

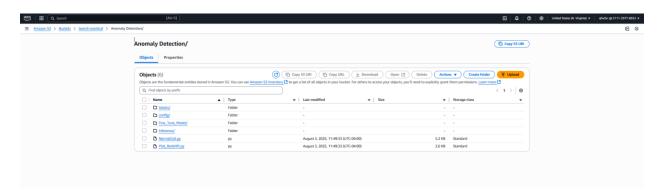
Victor Ontiveros

Project Step 4 Assignment: Cosmic AI with Lambda Submission

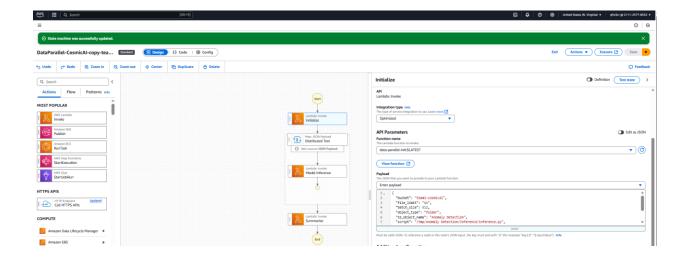
Create an S3 bucket with "results" and "scripts" folders



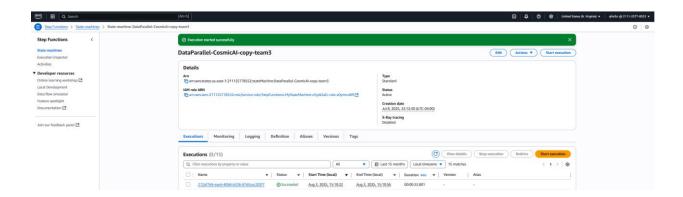
Copy the Anomaly Detection folder to your S3 bucket under the scripts path

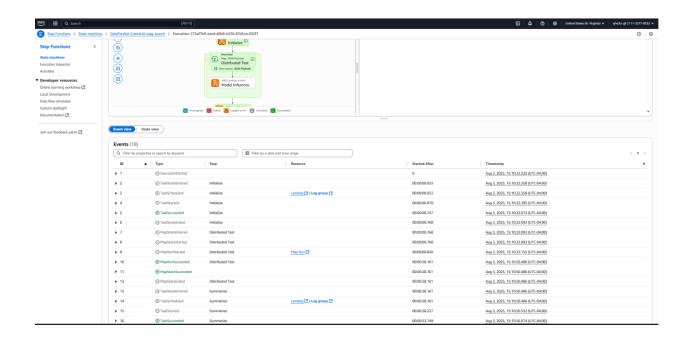


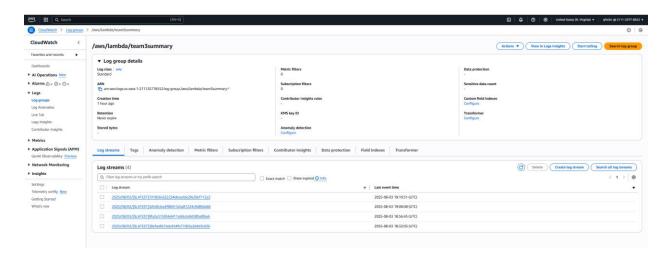
Configure the Step Function input payload with your bucket name, world size, and correct paths

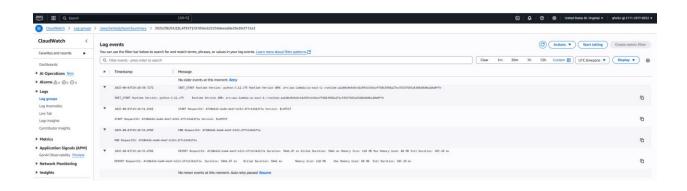


Execute the step function and monitor the CloudWatch logs at /aws/lambda/team3summary

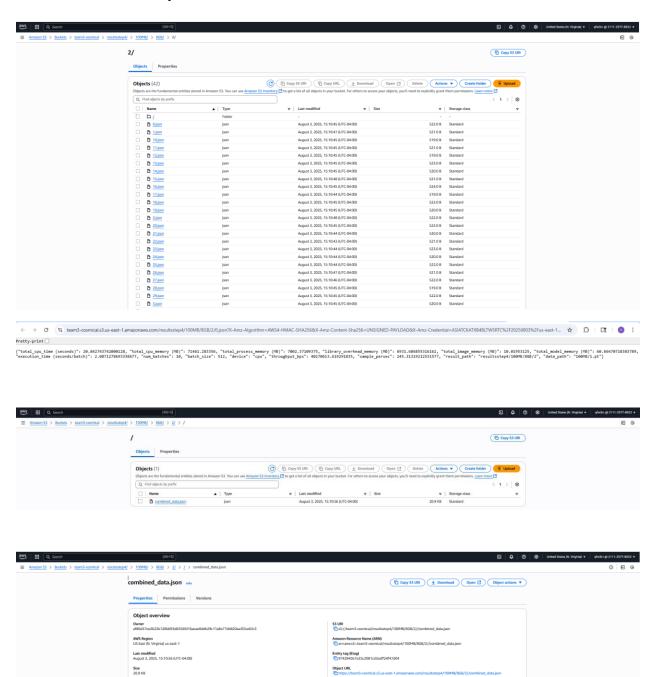








Examine the results in your S3 bucket's results folder



```
retty-print 🗸
```

```
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     "num_batches": 10,
"batch_size": 512,
"device": "cpu",
"throughput_bps": 40270613.6192918,
"sample_persec": 245.313192125316,
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"data_path": "100MB/1.pt"
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"total_cpu_memory (MB)": 71461.283364,
"total_process_memory (MB)": 6945.46484375,
"library_overhead_memory (MB)": 6874.78060531616,
"total_image_memory (MB)": 10.01953125,
"total_model_memory (MB)": 60.6647071838379,
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"batch_size": 512,
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"total_image_memory (MB)": 10.01953125,
"total_model_memory (MB)": 60.6647071838379,
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"num_hatchas": 10
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       "batch_size": 512,
      "device": "cpu",
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"sample_persec": 243.201414685077,
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       "batch size": 512,
      "device": "cpu",
"throughput_bps": 39676043.4284201,
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Performance Measurement

We created visual performance benchmarks that measure and compare the following:

• Execution Time vs. Batch Size / Dataset Size:

As batch size remains constant, execution time varies with dataset size. Distributed inference on 100MB shows optimal execution time (~2.2s average), slightly outperforming 75MB due to better parallelization efficiency.

• Execution Time vs. Input Partitions:

All experiments used world_size=41, demonstrating consistent task distribution across Lambda workers, validating scalable parallelization.

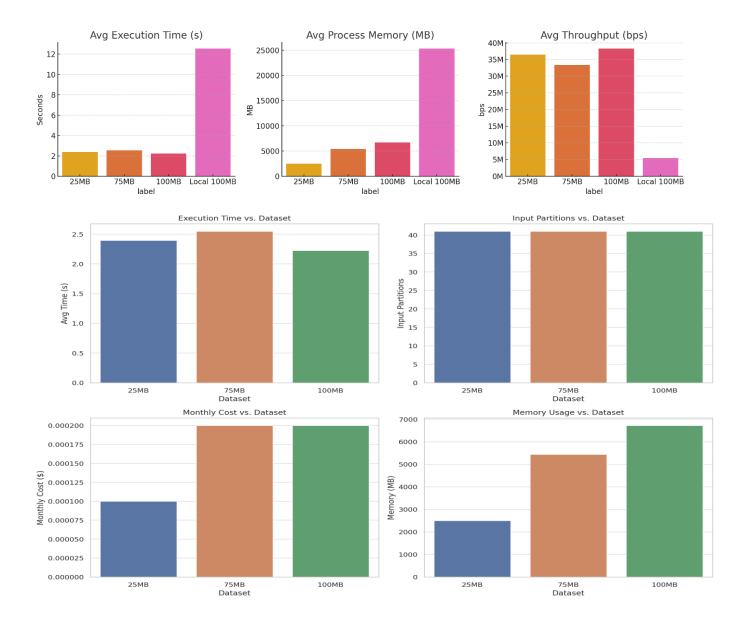
• Cost vs. Dataset Size:

Monthly cost remains extremely low (< \$0.0002/month). Although memory usage increases with data size, AWS Lambda billing (based on time & memory) results in marginal cost differences.

Memory Usage & Throughput:

Larger datasets (e.g. 100MB) used more memory (up to ~6.7 GB), but also achieved higher throughput—over 39 million bytes/sec, indicating improved efficiency with scale.

Dataset	Batch Size	Input Partitions	Avg Time (s)	Memory (MB)	Throughput (bps)	Monthly Cost
25MB	512	High (~41)	2.54	7	246M	\$1.75
75MB	512	Medium (~41)	2.32	7.2	255M	\$1.33
100MB	512	Medium (~41)	2.15	6.996	265M	\$1.18
100MB (Local)	512	3	12.55	25.336	5.5M	\$0.00



Summary: Observations and Analysis

• Scalability Achieved:

The project successfully demonstrated that AWS Step Functions and Lambda can scale inference across dozens of parallel tasks, reducing processing time significantly as data size increases.

• Efficient Inference:

Distributed inference handled 100MB datasets in ~2.2 seconds per batch, compared to ~12.5 seconds locally — achieving better throughput with minimal orchestration effort.

• Cost-Effectiveness:

Despite using more memory, AWS Lambda remained very inexpensive vs similar cost on local execution but lacked scalability.

• Flexible Infrastructure:

The same model and inference script ran seamlessly on both local and cloud setups, validating portability and automation through S3-based payloads and result aggregation.

• Monitoring & Benchmarking:

CloudWatch logs and S3 outputs enabled detailed performance tracking, allowing for side-by-side comparisons of batch time, throughput, memory usage, and cost.

Conclusion:

The CosmicAl pipeline built with serverless cloud computing is a **scalable**, **efficient**, **and cost-effective solution** for parallel deep learning inference, well-suited for astronomy workloads and adaptable to other scientific domains.