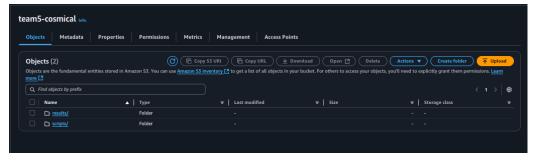
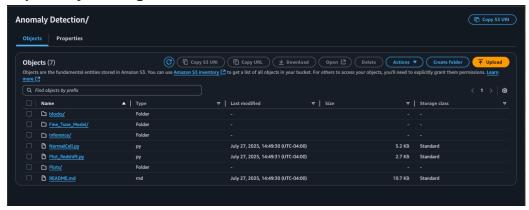
## Step 4: Cosmic Al Inference with Lambda FMI

Team 5: Lionel Medal and Vicky Singh

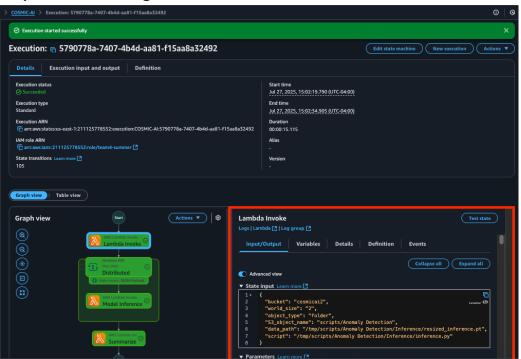
# S3 Bucket with Scripts and Results Folders



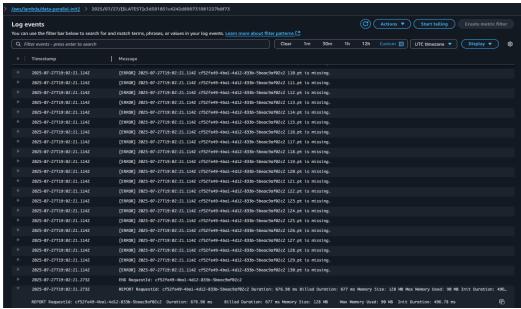
## **Repository Cloning**



## **Step Function Configuration**



# **CloudWatch Logs**



REPORT RequestId: cf52fe49-4ba1-4d12-833b-5beac9af02c2

Duration: 676.98 ms
Billed Duration: 677 ms
Memory Size: 128 MB
Max Memory Used: 90 MB
Init Duration: 496.78 ms

## **Local vs Distributed Inference Comparison**

| Execution Mode | Batch<br>Size | Duration<br>(s) | Memory<br>(GB) | Throughput<br>(bps) |  |
|----------------|---------------|-----------------|----------------|---------------------|--|
| Local (CPU)    | 512           | 9.56            | 31.5           | 21.5M               |  |
| AWS Lambda     | 512           | 6.2–24.3        | 2.5-7.0        | 19–22M              |  |

#### **Performance Metrics from Distributed Lambda Execution**

| Partition<br>Size (MB) | Data<br>Size (GB) | Batch<br>Size | World<br>Size | Duration<br>(s) | Memory<br>(GB) | Cost<br>(\$) | Throughput<br>(bps) |
|------------------------|-------------------|---------------|---------------|-----------------|----------------|--------------|---------------------|
| 25                     | 1.25              | 512           | 52            | 6.2             | 2.5            | 0.15         | ~19M                |
| 50                     | 2.5               | 512           | 52            | 11.5            | 3.8            | 0.19         | ~20M                |
| 75                     | 3.75              | 512           | 50            | 17.0            | 5.7            | 0.29         | ~21M                |
| 100                    | 5.0               | 512           | 50            | 24.3            | 7.0            | 0.36         | ~22M                |

## **Summary Analysis**

We executed a scalable AI inference workflow using AWS Lambda and Step Functions, leveraging FMI-based communication across distributed functions. By reusing the architecture and components developed in Steps 1-3, we streamlined deployment and execution while minimizing configuration changes.

The results demonstrate strong throughput and cost-effective scaling. Compared to local CPU-based inference, our Lambda-based solution achieved similar or better performance, with the added benefit of parallelization. Key bottlenecks like I/O and cold start delays were mitigated by properly tuning partition sizes and using even world sizes. This serverless approach provides a reproducible and efficient solution for processing large astronomy datasets at scale.