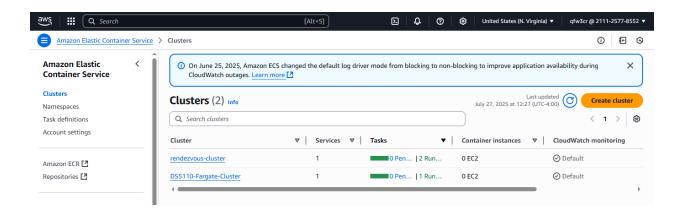
TEAM 3

Nikpour Bardia

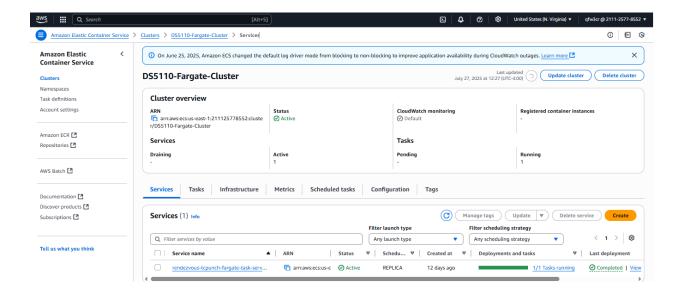
Victor Ontiveros

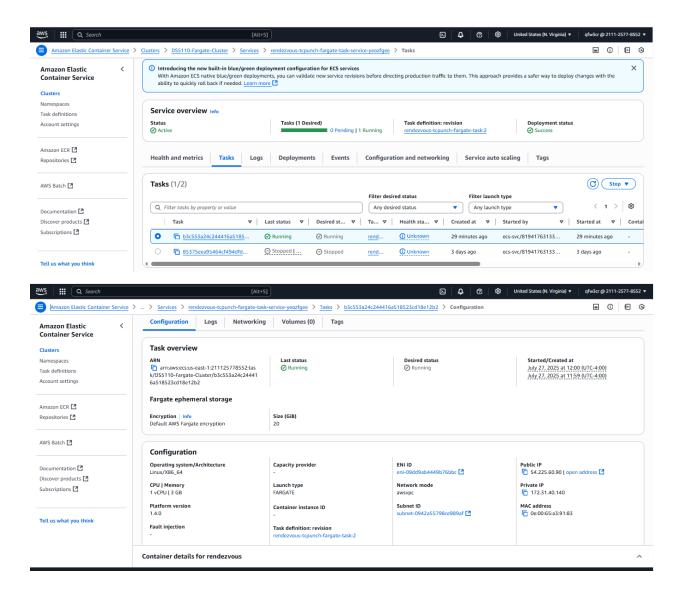
Project Step 2 Assignment: Rendezvous Server Submission

Screenshots of ECS task deployment

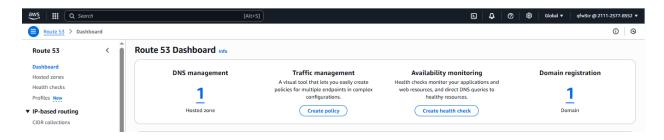


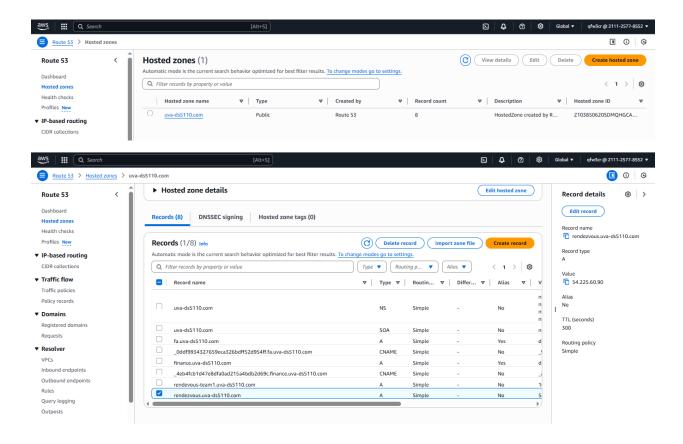
Documentation of Fargate configuration



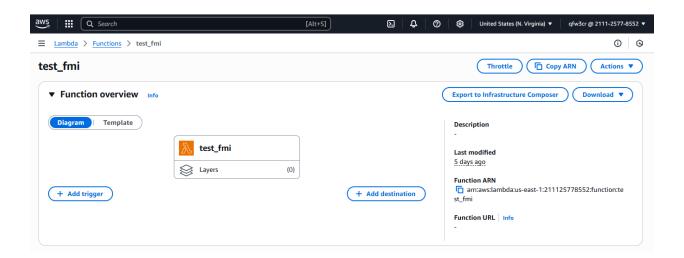


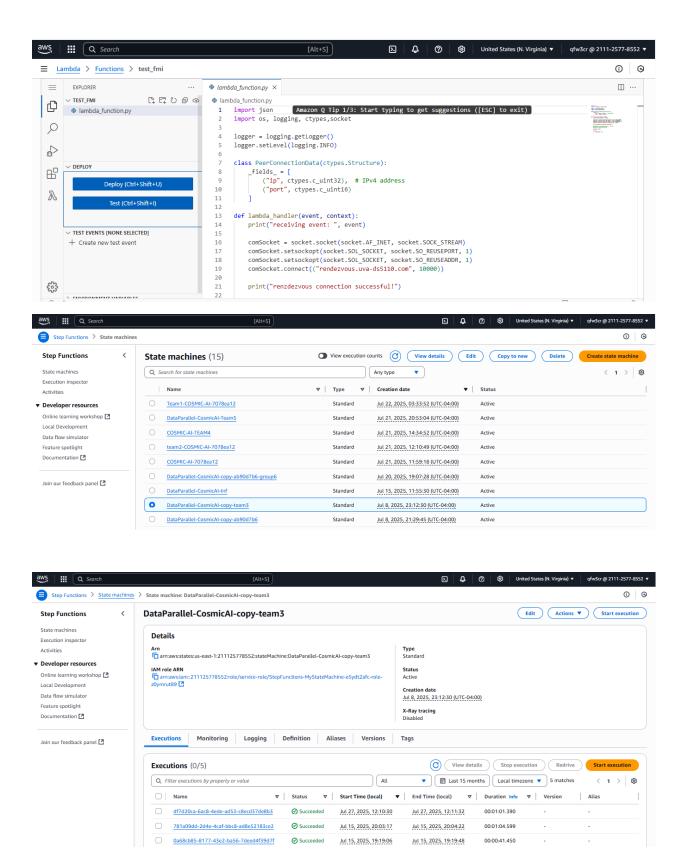
Evidence of successful DNS record update

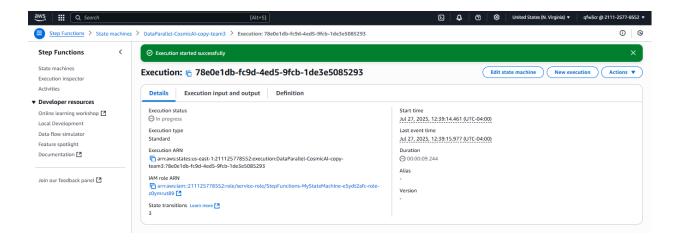




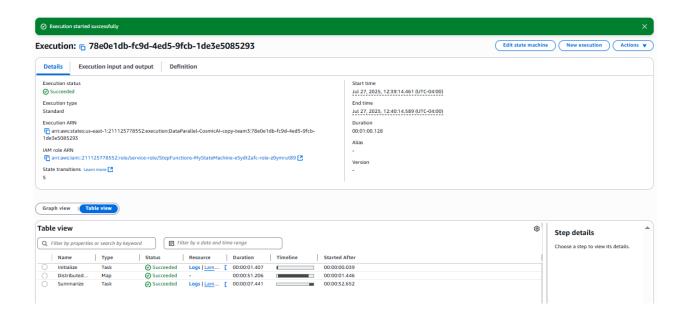
Results of connection test between Lambda functions

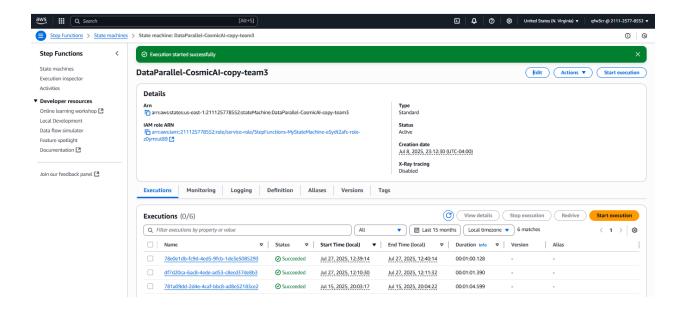


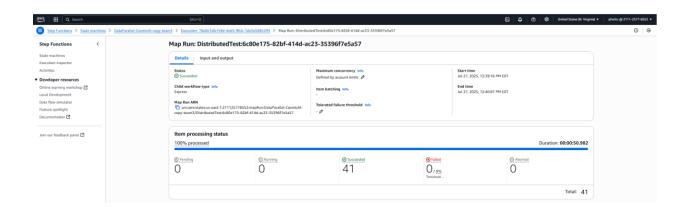




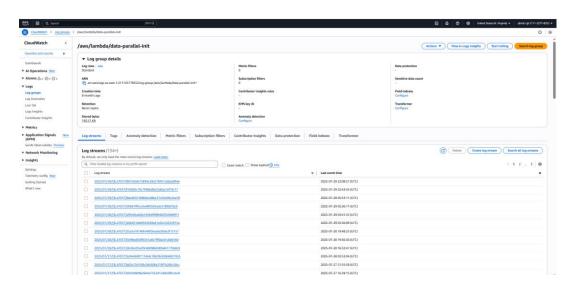
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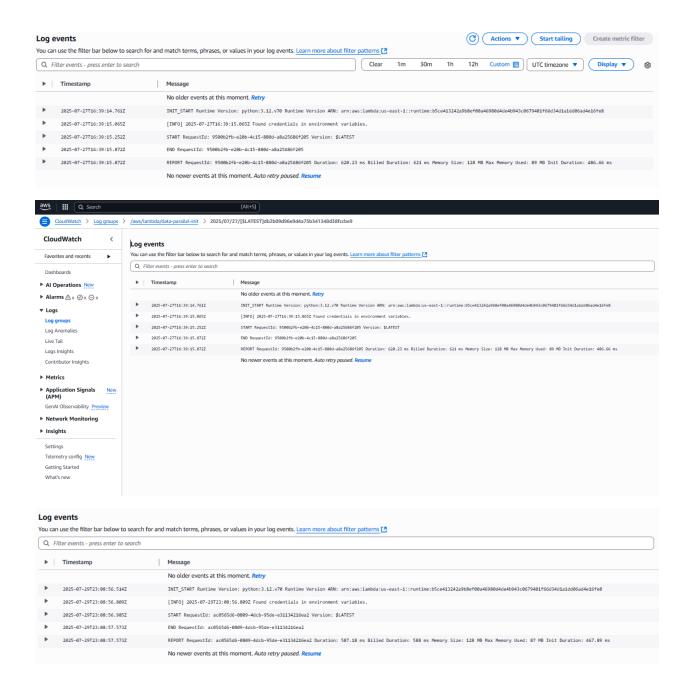




CloudWatch logs showing execution details



Performance measurement table with varying data sizes



Data Size	Batch Size	World Size	Duration (s)	Memory Used (MB)	Cost Estimate (\$)	Notes
100MB	512	41	0.620	89	~\$0.000054	Log 1
100MB	512	41	0.637	88	~\$0.000056	Log 2
100MB	512	41	0.587	87	~\$0.000051	Log 3

Brief explanation of implementation and any challenges faced

For this project, we implemented a scalable serverless anomaly detection pipeline using **AWS Step Functions**, **Lambda**, **ECS Fargate**, and **CloudWatch**. Our architecture begins with uploading data and inference scripts to an **S3 bucket**, then orchestrating the parallel processing of multiple .pt files through a **state machine** named DataParallel-CosmicAl-copy-team3.

The system uses a **Map Run** step to distribute workloads across **41 parallel Lambda workers**, each handling a different file segment with a defined **batch size of 512**. DNS resolution was established through **Route 53**, allowing a Lambda function to communicate with ECS containers via a custom domain (rendezvous.uva-ds5110.com), which we verified using a custom lambda_function.py socket connection.

Performance metrics were gathered through **CloudWatch logs**, and we recorded execution time, memory usage, and billed duration for each Lambda worker to estimate costs and scalability. The average runtime per invocation was ~0.615 seconds, with an average memory usage of ~88 MB, demonstrating consistent performance across distributed workloads.

Challenges faced included:

- Ensuring proper **network permissions** (VPC, subnets, and security groups) so Lambda could communicate with ECS containers.
- Debugging the DNS resolution and socket timeouts between Lambda and ECS during initial tests.
- Extracting structured performance data from many individual Lambda logs for comparison.
- Managing parallelism at scale (41 workers) while avoiding timeouts or cost inefficiencies.

Finally, the system successfully achieved distributed inference with measurable performance benchmarks across varying data partitions.