Course Outline

Scaling Out: Effective Cluster Computing with Distributed Dask

Course Summary

Description

This class addresses the transition from working successfully on a single server or experimenting with a minimal cluster to achieving successful, reliable, repeatable use of larger Dask compute clusters. We focus on a deep dive into all of the critical components in a distributed Dask cluster, how they work together, and how you can configure them to maximize throughput and minimize costs.

Objectives

At the end of this course, students will understand:

- What components make up a distributed Dask cluster and what purposes they serve
- How to configure cluster resources to meet your workload needs
- How to identify problems, debug, and troubleshoot successfully

Topics

- Introduction
- Distributed Dask: Cast of Characters
- Basic Operation of Dask Clusters
- Tasks
- Distributed Data
- Resource usage and Resilience
- Best Practices, Debugging
- Use Case Example: Orchestrating Batch ML Scoring
- Q & A Discussion

Audience

This course is intended for engineers or data scientists who typically work with large data clusters.

Prerequisites

Students should have experience in Python and Pandas and/or SQL programming, both at a basic level.

Duration

One day

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I. Introduction
- A. About Dask and Coiled Computing: Making scale-out computing easier
- B. Simplest distributed cluster: manual setup
- C. Changes in transitioning to distributed environment
  Storage, fast universal memory access, single shared executable
- D. Implications for users (devs) and admin (IT)

II. Distributed Dask: Cast of Characters
- A. Client, Scheduler, Nanny, Worker
- B. Where these services are located, their relationships and roles
- C. Supporting Players: cluster resource manager (e.g., k8s, Coiled Cloud, YARN, etc.)

III. Basic Operation of Dask Clusters
- A. User perspective
- B. Creating clusters with helper tools: Cloud Provider, Coiled Cloud, etc.
- C. Cluster API
- D. Sizing your cluster
- E. Scaling your scaling – manual/automatic
- F. Admin perspective
  - a. CLI: dask-scheduler and dask-worker
  - b. Managing the worker environment
  - c. Additional admin concerns (security, tagging, and costs)

IV. Tasks
- A. Submitting tasks and directing output
- B. Scheduling policy
- C. Finding your tasks and data (programmatically)
- D. Seeing your tasks and data: the Dask Dashboard

V. Distributed Data
- A. Source data via tasks
- B. Source data scatter
- C. Storing data worker-local
- D. Handling output (result) data, direct parallel write vs. gather/result

VI. Resource usage and Resilience
- A. Output spill location and resource management
- B. Work stealing
- C. Loss of processes
- D. Loss of storage on workers

VII. Best Practices, Debugging
- A. Dashboard information pages
- B. Additional GUIs (e.g., profiler)
- C. Review of best practices
- D. Remote debugging
- E. client.run

VIII. Use Case Example: Orchestrating Batch ML Scoring
- A. Source data on disk
- B. ML model
- C. Options for inference, pros/cons
- D. Supplying dependencies via code or container image
- E. Basic workflow
- F. Improvements and optimizations (e.g., batch size)

IX. Q & A, Discussion