#### Azure MapReduce

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# Agenda

- Recap of Azure Cloud Services
- Recap of MapReduce
- Azure MapReduce Architecture
- Pairwise distance alignment implementation
- Next steps

# **Cloud Computing**

- On demand computational services over web
  - Backed by massive commercial infrastructures giving economies of scale
  - Spiky compute needs of the scientists
- Horizontal scaling with no additional cost

   Increased throughput
- Cloud infrastructure services
  - Storage, messaging, tabular storage
  - Cloud oriented services guarantees
  - Virtually unlimited scalability
- Future seems to be CLOUDY!!!

# **Azure Platform**

- Windows Azure Compute
  - net platform as a service
  - Worker roles & web roles
- Azure Storage
  - Blobs
  - Queues
  - Table
- Development SDK, fabric and storage

## MapReduce

- Automatic parallelization & distribution
- Fault-tolerant
- Provides status and monitoring tools
- Clean abstraction for programmers
  - map (in\_key, in\_value) ->
     (out\_key, intermediate\_value) list
  - reduce (out\_key, intermediate\_value list) ->
     out\_value list

## Motivation

 Currently no parallel programming framework on Azure

– No MPI, No Dryad

- Well known, easy to use programming model
- Cloud nodes are not as reliable as conventional cluster nodes

# Azure MapReduce Concepts

- Take advantage of the cloud services
  - Distributed services, Unlimited scalability
  - Backed by industrial strength data centers and technologies
- Decentralized control
- Dynamically scale up/down
- Eventual consistency
- Large latencies

Coarser grained map tasks

Global queue based scheduling



#### 1. Client driver loads the map & reduce tasks to the queues



#### 2. Map workers retrieve map tasks from the queue



3. Map workers download data from the Blob storage and start processing



4. Reduce workers pick the tasks from the queue and start monitoring the reduce task tables



5. Finished map tasks upload the results to Blob storage. Add entries to the respective reduce task tables.



#### 6. Reduce tasks download the intermediate data products



7. Start reducing when all the map tasks are finished and when a reduce task is finished downloading the intermediate data products

## Azure MapReduce Architecture

- Client API and driver
- Map tasks
- Reduce tasks
- Intermediate data transfer
- Monitoring
- Configurations

# Fault tolerance

- Use the visibility timeout of the queues
  - Currently maximum is 3 hours
  - Delete the message from the queue only after everything is successful
  - Execution, upload, update status
- Tasks will rerun when timeout happens
  - Ensures eventual completion
  - Intermediate data are persisted in blob storage
  - Retry up to 3 times
- Many retries in service invocations

	Apache Hadoop	Microsoft Dryad [25]	Twister [19]	Azure Map
	[24] /(Google MR)			Reduce/Twister
Programming	MapReduce	DAG execution,	Iterative	MapReduce will
Model		Extensible to	MapReduce	extend to Iterative
		MapReduce and other		MapReduce
		patterns		
Data Handling	HDFS (Hadoop	Shared Directories &	Local disks and	Azure Blob Storage
	Distributed File	local disks	data management	
	System)		tools	
Scheduling	Data Locality; Rack	Data locality;	Data Locality;	Dynamic task
	aware, Dynamic	Network	Static task	scheduling through
	task scheduling	topologybased	partitions	global queue
	through global	run time graph		
	queue	optimizations; Static task		
		partitions		
<b>Failure Handling</b>	Re-execution of	Re-execution of failed	Re-execution of	Re-execution of
	failed tasks;	tasks; Duplicate	Iterations	failed tasks;
	Duplicate execution	execution of slow tasks		Duplicate execution
	of slow tasks			of slow tasks
Environment	Linux Clusters,	Windows HPCS cluster	Linux Cluster	Window Azure
	Amazon Elastic Map		EC2	Compute, Windows
	Reduce on EC2			Azure Local
				Development
				Fabric
Intermediate	File, Http	File, TCP pipes, shared-	Publish/Subscribe	Files, TCP
data transfer		memory FIFOs	messaging	

# Why Azure Services

- Virtually unlimited scalable distributed services
- No need to install software stacks
  - In fact you can't 🙂
  - Eg: NaradaBrokering, HDFS, Database
- Zero maintenance

Let the platform take care of you

• Availability guarantees

# API

- ProcessMapRed(jobid, container, params, numReduceTasks, storageAccount, mapQName, reduceQName,List mapTasks)
- Map(key, value, programArgs, Dictionary outputCollector)
- Reduce(key, List values, programArgs, Dictionary outputCollector)

# Develop applications using Azure MapReduce

- Local debugging using Azure development fabric
- DistributedCache
  - Bundle with Azure Package
- Compile in release mode before creating the package.
- Deploy using Azure web interface
- Errors logged to a Azure Table

## SWG Pairwise Distance Alignment

- SmithWaterman-GOTOH
- Pairwise sequence alignment

- Align each sequence with all the other sequences



## AzureMR SWG Performance **10k Sequences**



Number of Azure Small Instances

## AzureMR SWG Performance 10k Sequences



## AzureMR SWG Performance on Different Instance Types



## AzureMR SWG Performance on Different Data Sizes



## Next Steps

- In the works
  - Monitoring web interface
  - Alternative intermediate data communication mechanisms
  - Public release
- Future plans
  - AzureTwister
    - Iterative MapReduce

# Thanks!!

• Questions? 😳

# References

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