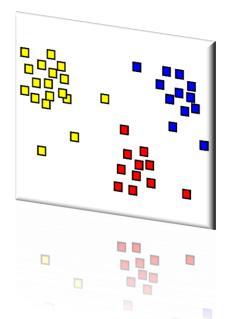


CloudClustering

Toward a scalable machine learning toolkit for Windows Azure



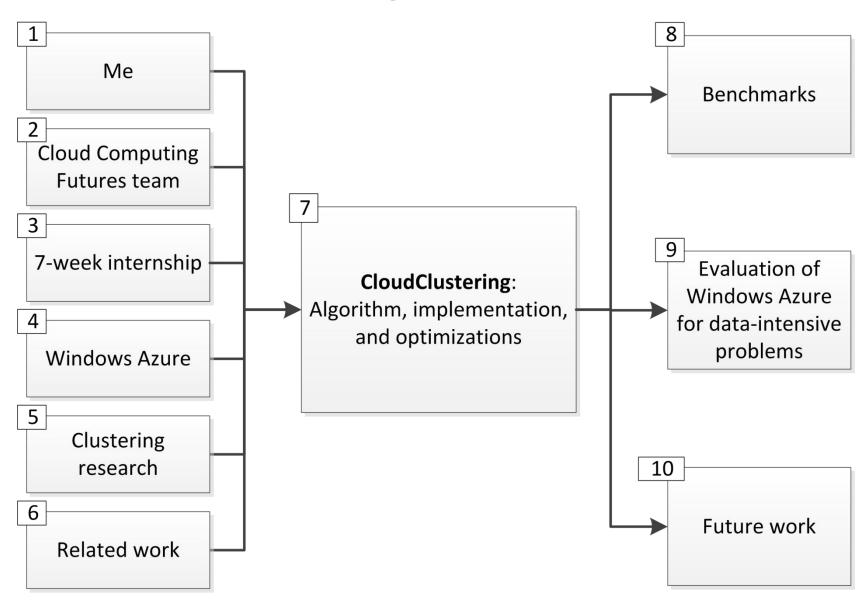
Ankur Dave

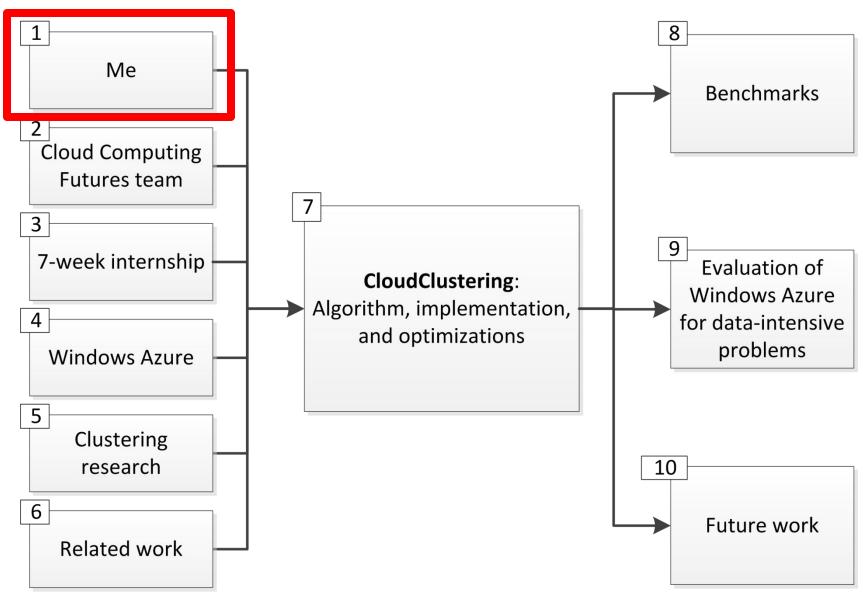
XCG Intern

Microsoft Research

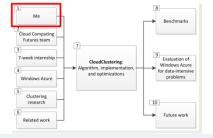
Mentors:

Roger Barga Wei Lu





About me



EDUCATION





UNIVERSITY OF CALIFORNIA, BERKELEY

WORK

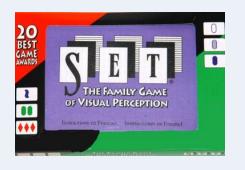






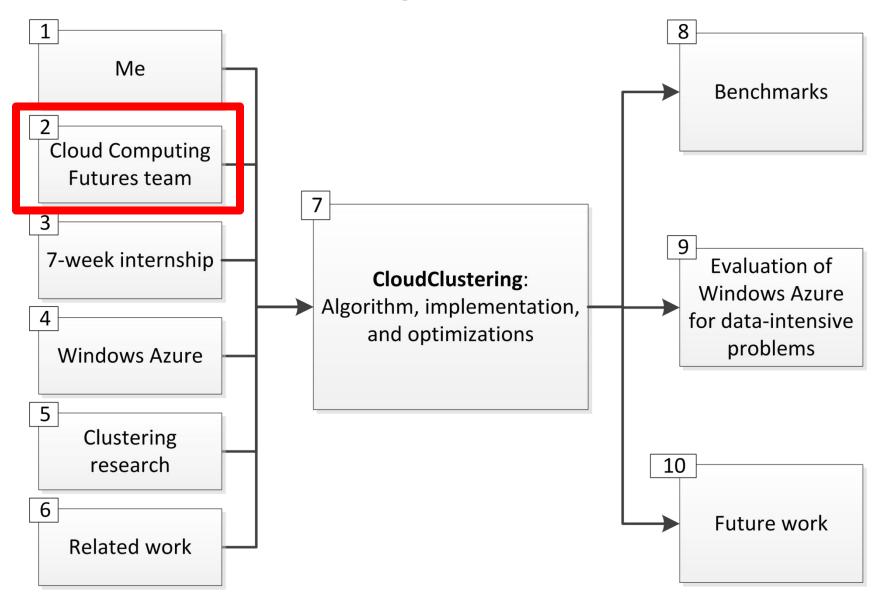


HOBBIES

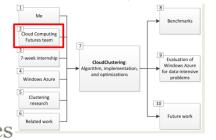








My team



Microsoft[®] Research

Cloud Computing Futures To create novel data center solutions, designs must be based on comprehensive optimization of all attributes, rather than gradually accruing incremental changes

based on current technologies and best practices. The Cloud Computing Futures team is tasked to invent on a large scale. Our goal is to reduce data center costs by four-fold or greater, including power consumption, while accelerating deployment and increasing adaptability and resilience to failures.



Azure Research Engagement



The Azure Research Engagement project aims to change the paradigm for scholarly and scientific research by extending the power of the computer into the cloud. We build the components of cloud technology and work with researchers in the field on projects that push the frontier of client and cloud computing.



Dan Reed

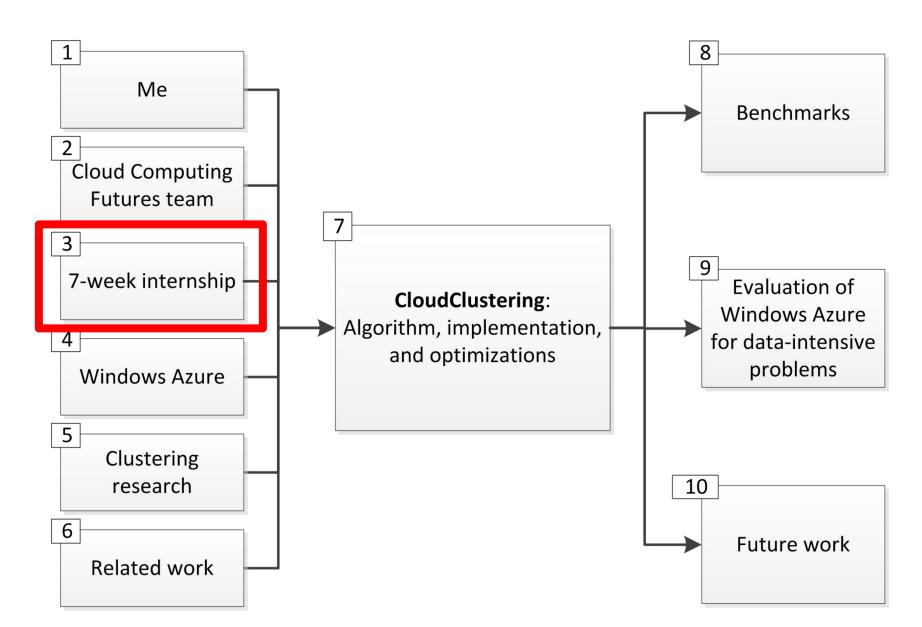
Dennis Gannon **Jaliya** Ekanayake

Jared Jackson

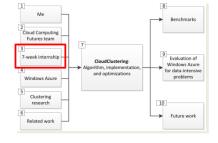
Nelson Araujo

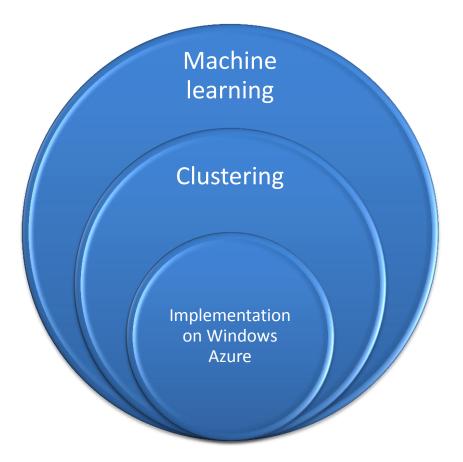
Roger Barga

Wei Lu



Internship: Domain





Goals:

- Build a scalable clustering algorithm on Azure
- Explore clustering and the cloud by reading papers
- Meet a variety of researchers at MSR

Week 1: Ramp-up; architecture planning

Week 2: Building CloudClustering base impl.

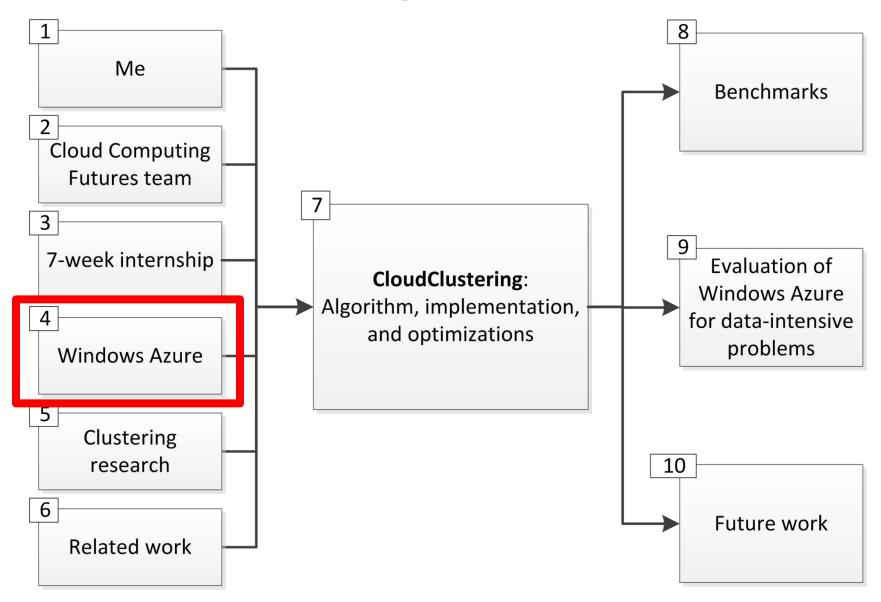
Week 3: Building CloudClustering base impl.

Week 4: Multicore parallelism with PLINQ

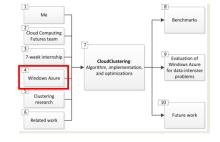
Week 5: Performance testing on Azure fabric

Week 6: Multicore w/threads; data affinity

Week 7: Presentation and report-out



Windows Azure

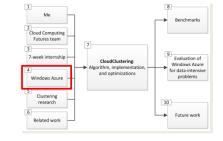


Exploring

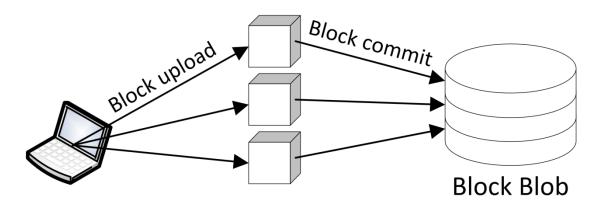


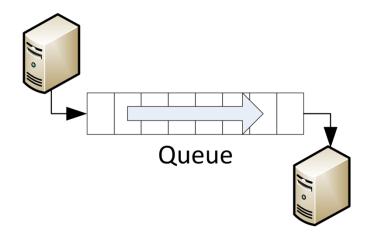
for data-intensive research

Windows Azure



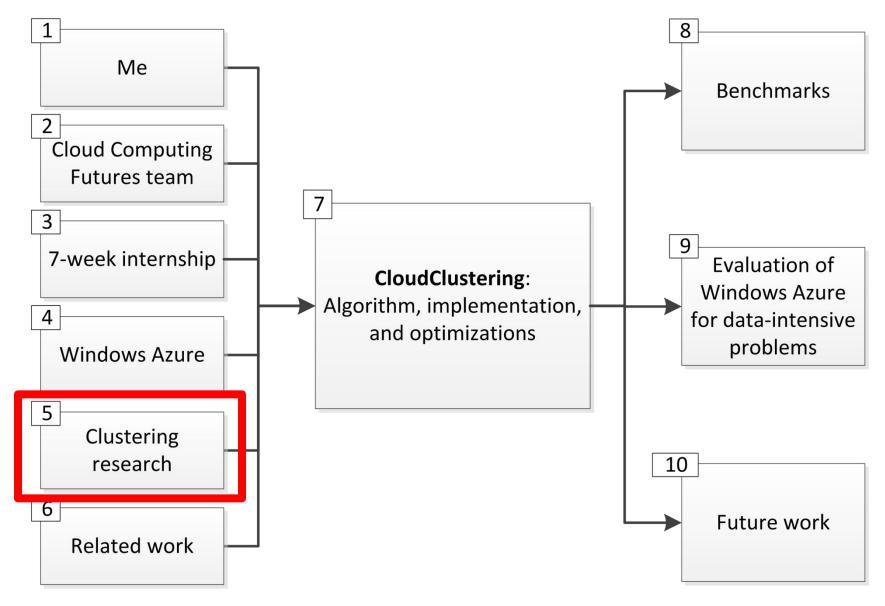


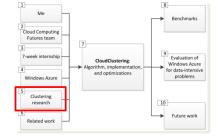


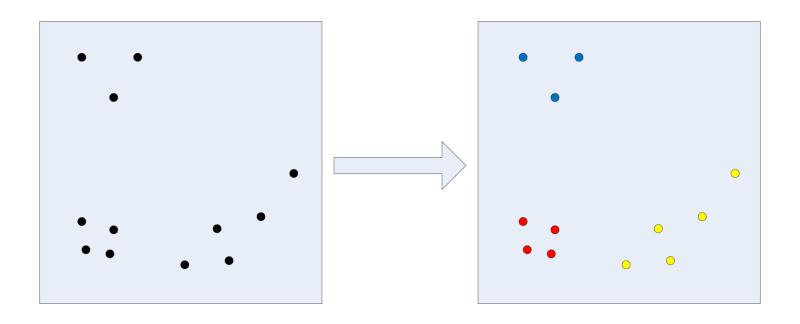


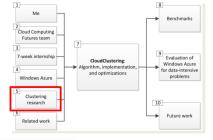
PartitionKey	RowKey	Col1	Col2

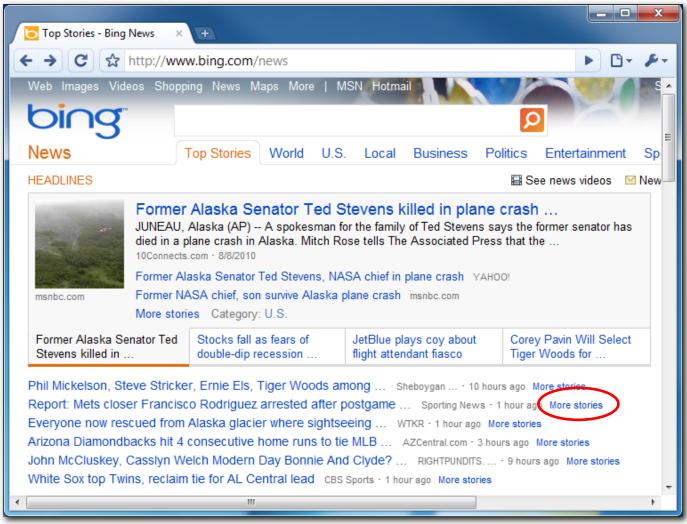
Table











Benchmarks

Evaluation of

Windows Azure for data-intensive

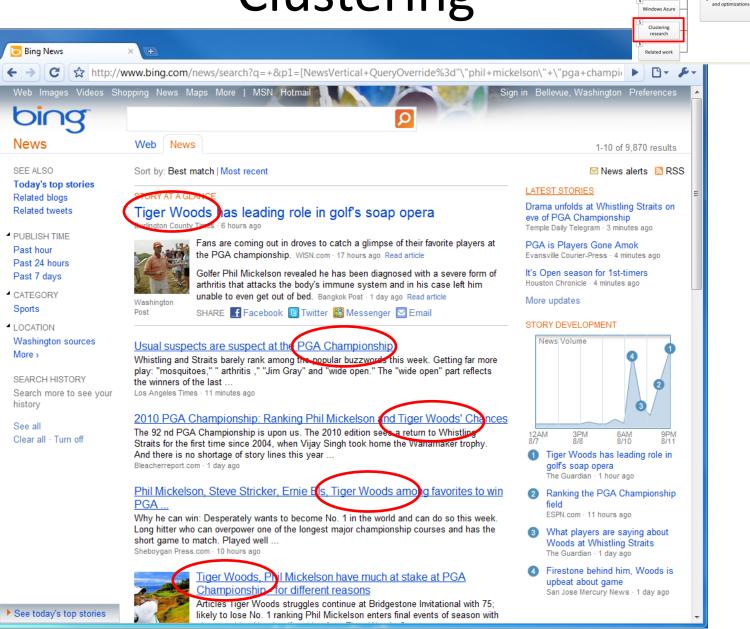
problems

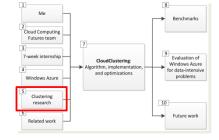
10 Future work

CloudClustering:

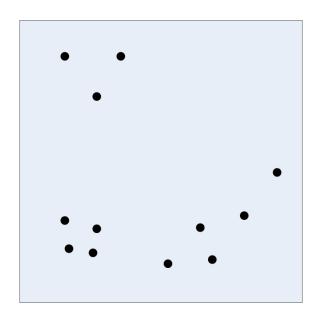
Cloud Computing

7-week internsh



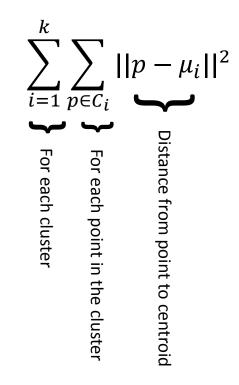


k-means



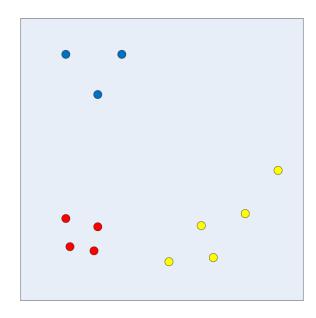
$$k = 3$$

 Minimizes the within-cluster sum of squares:

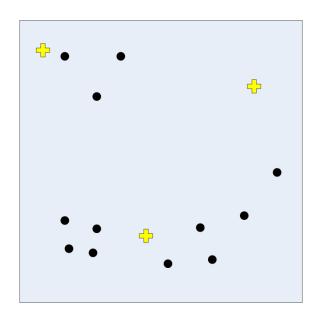


Me 2 Cloud Computing Futures team 3 7-week internship Algorithm, implementation, and optimizations Windows Azure for data-intensive problems Clustering research Related work 8 Benchmarks 9 Evaluation of Windows Azure for data-intensive problems

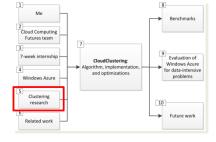
Target clustering



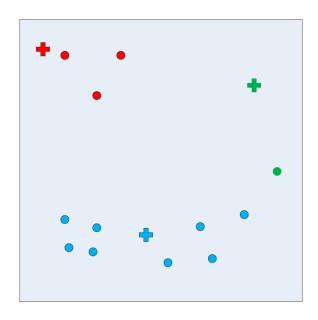
1. Initialization

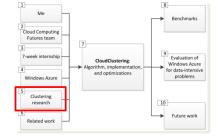


$$k = 3$$

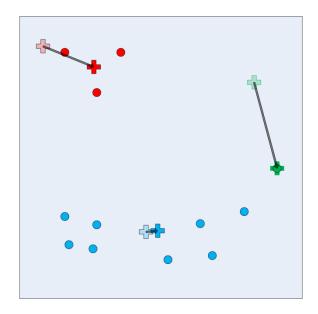


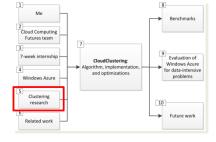
2. Assign Points to Centroids #1



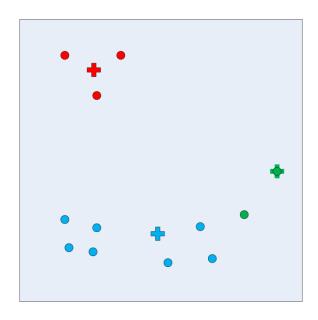


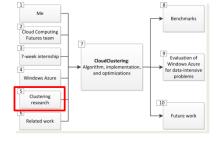
3. Recalculate Centroids #1



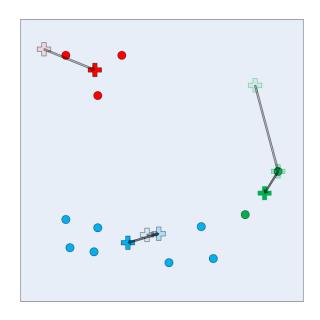


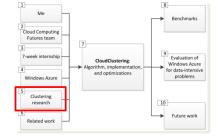
4. Assign Points to Centroids #2



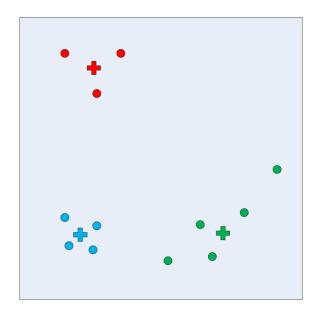


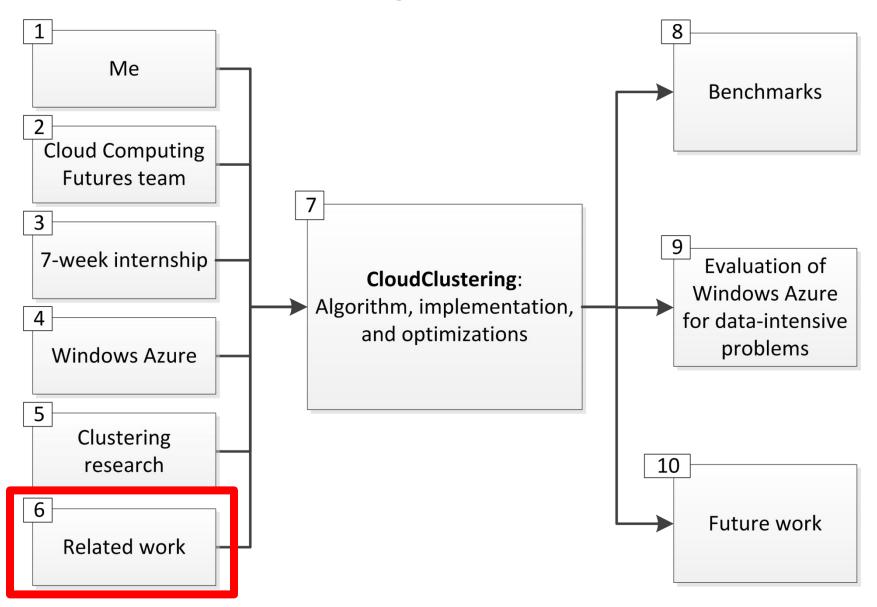
5. Recalculate Centroids #2



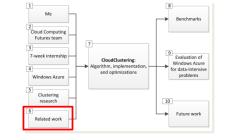


Stopping Condition

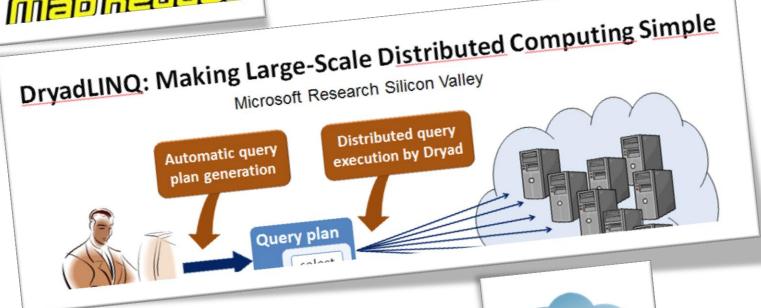




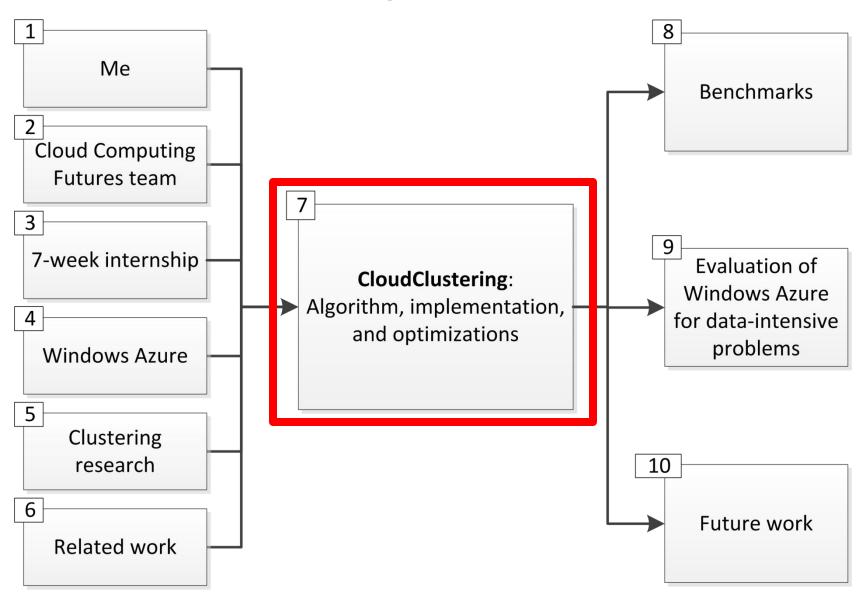
Related Work



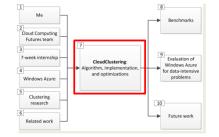




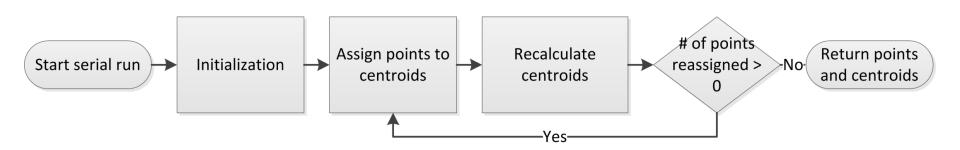




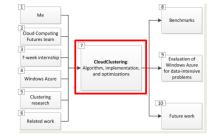
CloudClustering: Algorithm



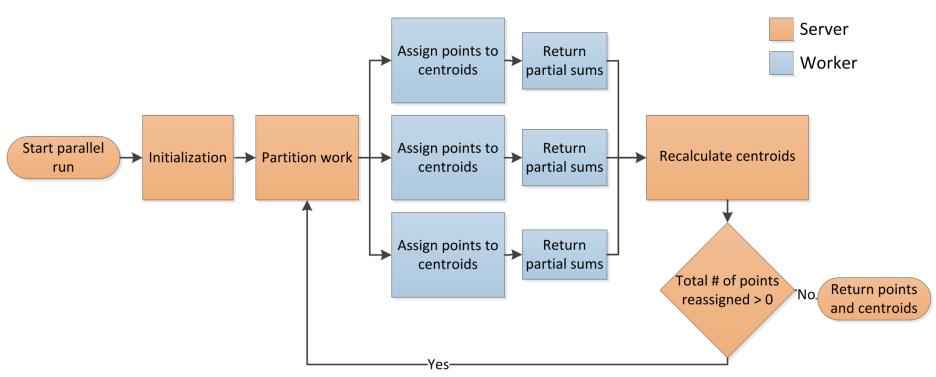
Conventional (Serial) k-means



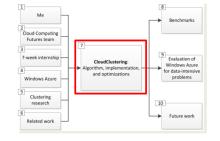
CloudClustering: Algorithm

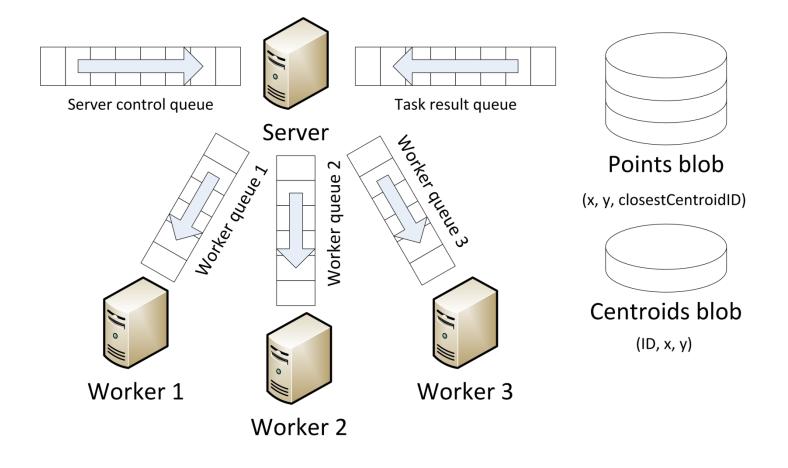


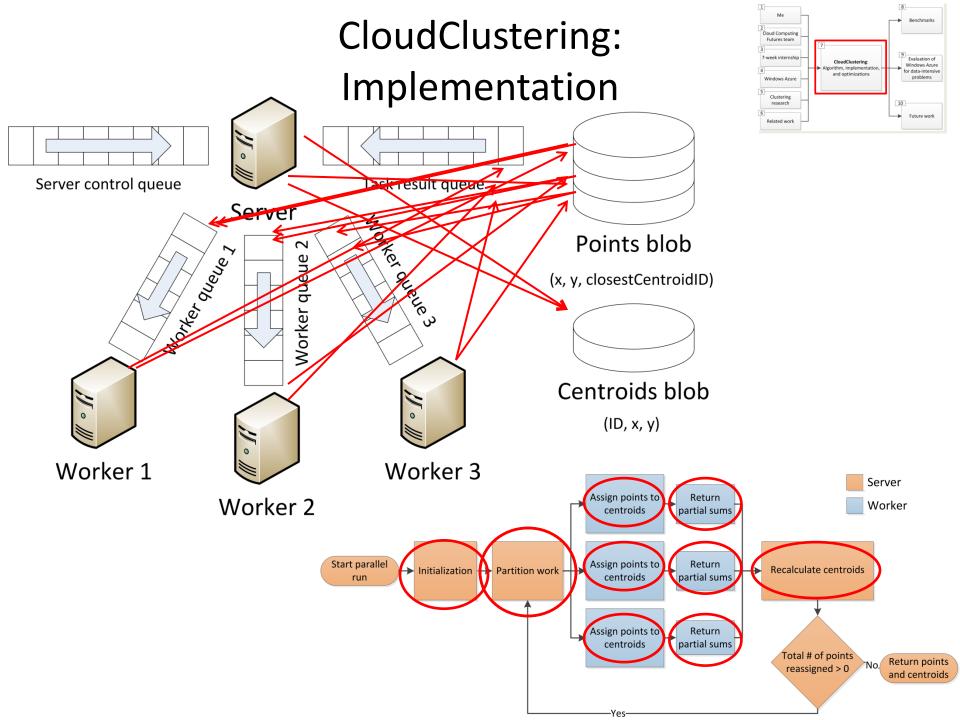
Parallel k-means



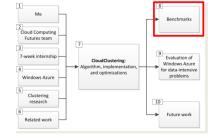
CloudClustering: Implementation





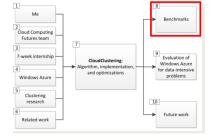


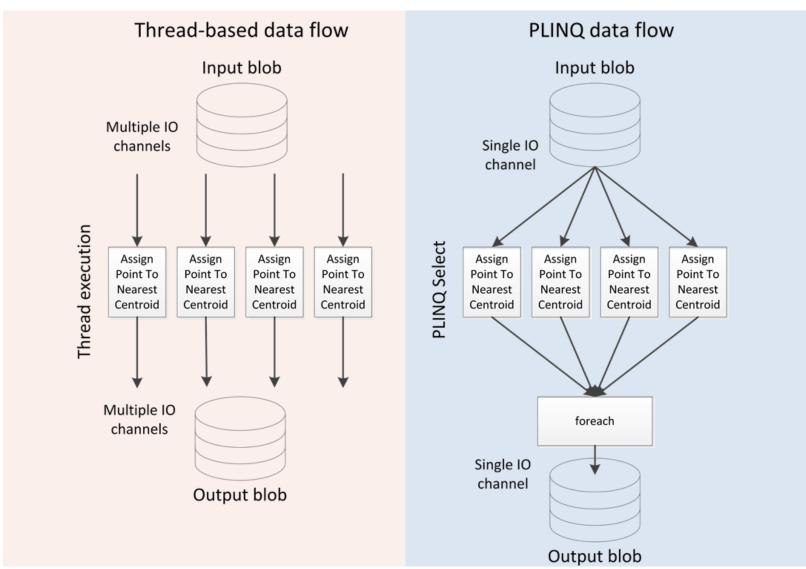
CloudClustering: Leveraging the cloud



- Multicore Parallelism
- Data Affinity
- Efficient Blob Concatenation
- Dynamic Scalability

CloudClustering: Multicore Parallelism





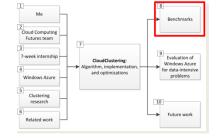
Thread-based

```
private void ProcessPoints()
   CloudBlockBlob pointsBlob = AzureHelper.GetBlob(task.Points);
   // Do the mapping and write the new blob
    int numThreads = Environment.ProcessorCount:
   PointsProcessedData[,] pointSumsPerCentroidPerThread = new PointsProcessedData[numThreads, task.K];
    int[] pointsChangedPerThread = new int[numThreads];
    string[][] blockIDsPerThread = new string[numThreads][];
   System.Threading.Tasks.Parallel.For(0, numThreads, threadID =>
       using (ObjectCachedStreamReader<ClusterPoint> stream = new ObjectCachedStreamReader<ClusterPoint>(pointsBlob,
ClusterPoint.FromByteArray, ClusterPoint.Size, AzureHelper.GetLocalResourceRootPath("cache"), task.JobID.ToString(),
task.PartitionNumber, task.M, subPartitionNumber: threadID, subTotalPartitions: numThreads))
            ObjectBlockWriter<ClusterPoint> writeStream = new ObjectBlockWriter<ClusterPoint>(pointsBlob, point =>
point.ToByteArray(), ClusterPoint.Size);
            foreach (var point in stream)
               // Assign the point to the nearest centroid
               Guid oldCentroidID = point.CentroidID;
               int closestCentroidIndex = centroids.MinIndex(centroid => Point.Distance(point, centroid));
               Guid newCentroidID = point.CentroidID = centroids[closestCentroidIndex].ID;
               // Write the updated point to the writeStream
               writeStream.Write(point);
                // Update the number of points changed
               if (oldCentroidID != newCentroidID)
                    pointsChangedPerThread[threadID]++;
                // Update the point sums
                if (pointSumsPerCentroidPerThread[threadID, closestCentroidIndex] == null)
                    pointSumsPerCentroidPerThread[threadID, closestCentroidIndex] = new PointsProcessedData();
               pointSumsPerCentroidPerThread[threadID, closestCentroidIndex].PartialPointSum += point;
               pointSumsPerCentroidPerThread[threadID, closestCentroidIndex].NumPointsProcessed++;
            // Collect the block IDs from writeStream
            writeStream.FlushBlock();
            blockIDsPerThread[threadID] = writeStream.BlockList.ToArray();
    // Combine the per-thread block lists and write the full block list to a blob. Then include that as part of TaskResult
   List<string> blockIDs = new List<string>();
    foreach (string[] blockIDsFromThread in blockIDsPerThread)
       blockIDs.AddRange(blockIDsFromThread);
    CloudBlob blockIDsBlob = AzureHelper.CreateBlob(task.JobID.ToString(), Guid.NewGuid().ToString());
    using (Stream stream = blockIDsBlob.OpenWrite())
       BinaryFormatter bf = new BinaryFormatter();
       bf.Serialize(stream, blockIDs);
    TaskResult.PointsBlockListBlob = blockIDsBlob.Uri:
    // Total up the per-thread pointSumsPerCentroid
    TaskResult.PointsProcessedDataByCentroid = new Dictionary<Guid, PointsProcessedData>();
    for (int i = 0; i < task.K; ++i)
       Guid centroidID = centroids[i].ID;
       TaskResult.PointsProcessedDataByCentroid[centroidID] = new PointsProcessedData();
       for (int j = 0; j < numThreads; ++j)</pre>
            if (pointSumsPerCentroidPerThread[j, i] != null)
               TaskResult.PointsProcessedDataByCentroid[centroidID].PartialPointSum += pointSumsPerCentroidPerThread[j,
il.PartialPointSum:
               TaskResult.PointsProcessedDataByCentroid[centroidID].NumPointsProcessed += pointSumsPerCentroidPerThread[j,
il.NumPointsProcessed;
    // Total up the per-thread numPointsChanged
    TaskResult.NumPointsChanged = 0;
    foreach (int threadPointsChanged in pointsChangedPerThread)
       TaskResult.NumPointsChanged += threadPointsChanged;
```

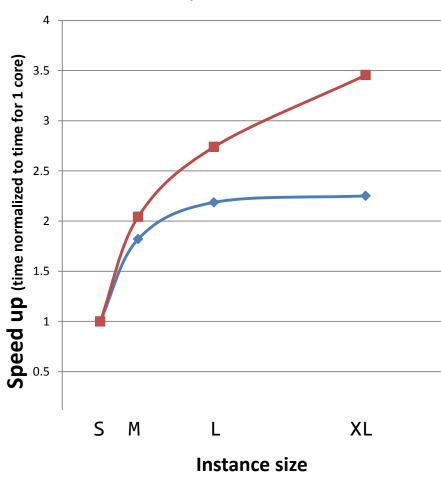
PLING

```
CloudBlockBlob pointsBlob = AzureHelper.GetBlob(task.Points);
    // Do the mapping and write the new blob
    using (ObjectStreamReader<ClusterPoint> stream = new ObjectStreamReader<ClusterPoint>(pointsBlob.
ClusterPoint.FromByteArray, ClusterPoint.Size, task.PartitionNumber, task.M))
       var assignedPoints = stream.AsParallel().Select(AssignClusterPointToNearestCentroid);
       ObjectBlockWriter<ClusterPoint> writeStream = new ObjectBlockWriter<ClusterPoint>(pointsBlob, point =>
point.ToByteArray(), ClusterPoint.Size);
        TaskResult.NumPointsChanged = 0:
       TaskResult.PointsProcessedDataByCentroid = new Dictionary<Guid, PointsProcessedData>();
       // Pipelined execution -- see http://msdn.microsoft.com/en-us/magazine/cc163329.aspx
        foreach (var result in assignedPoints)
            // Write the point to the new blob
            writeStream.Write(result.Point):
            // Update the number of points changed counter
            if (result.PointWasChanged)
                TaskResult.NumPointsChanged++;
            // Add to the appropriate centroid group
            if (!TaskResult.PointsProcessedDataByCentroid.ContainsKey(result.Point.CentroidID))
                TaskResult.PointsProcessedDataByCentroid[result.Point.CentroidID] = new PointsProcessedData();
            TaskResult.PointsProcessedDataByCentroid[result.Point.CentroidID].NumPointsProcessed++;
            TaskResult.PointsProcessedDataByCentroid[result.Point.CentroidID].PartialPointSum += result.Point;
       // Send the block list as part of TaskResult
       writeStream.FlushBlock();
       TaskResult.PointsBlockList = writeStream.BlockList;
private ClusterPointProcessingResult AssignClusterPointToNearestCentroid(ClusterPoint clusterPoint)
    ClusterPoint result = new ClusterPoint(clusterPoint):
    result.CentroidID = centroids.MinElement(centroid => Point.Distance(clusterPoint, centroid)).ID;
    return new ClusterPointProcessingResult
       Point = result.
       PointWasChanged = clusterPoint.CentroidID != result.CentroidID
    };
```

CloudClustering: Multicore Parallelism



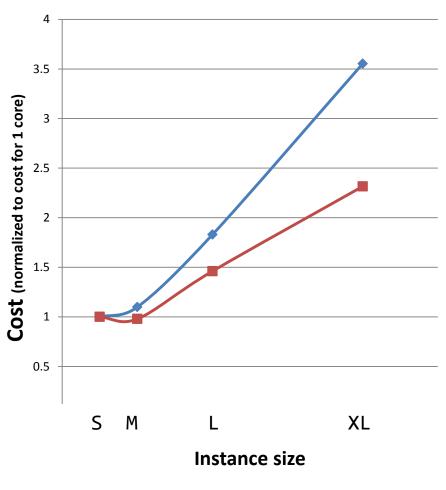
Scale-up: Speed up for varying instance size, PLINQ vs. Threads

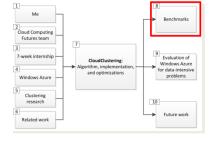


Threads speedup

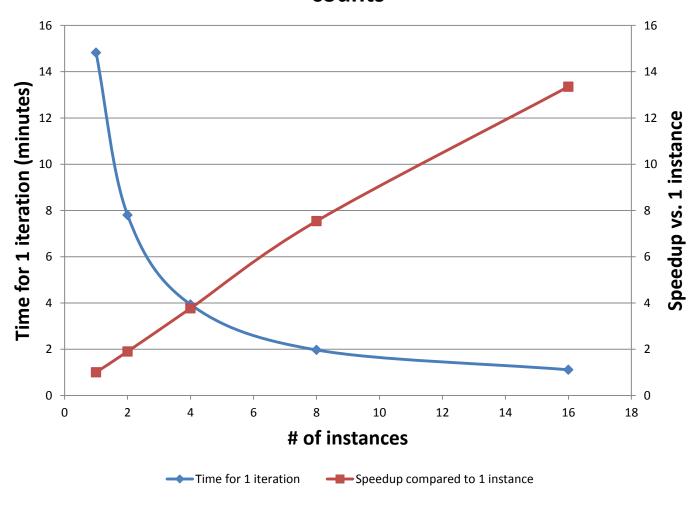
→ PLINQ speedup

Scale-up: Cost for varying instance size, PLINQ vs. Threads

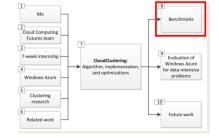


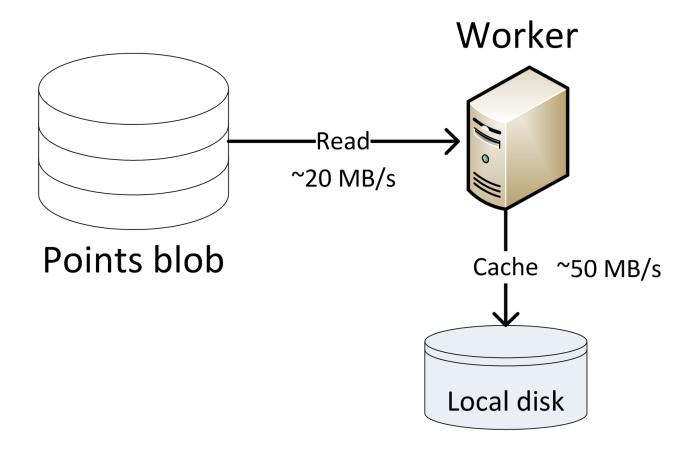


Scale-out: Time and speedup for varying instance counts

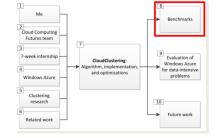


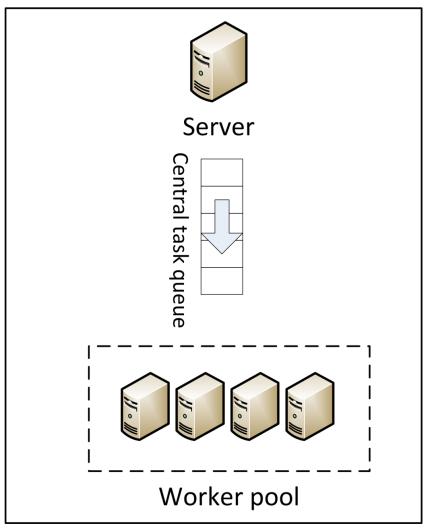
CloudClustering: Data Affinity

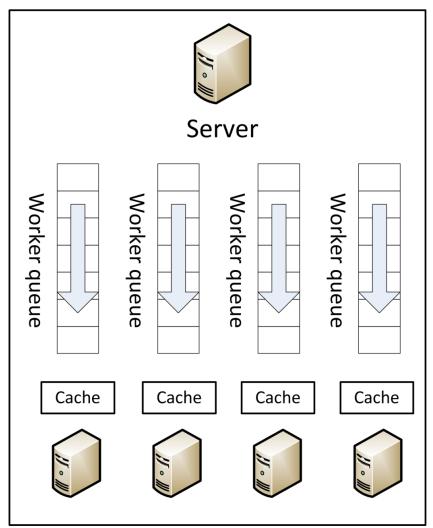




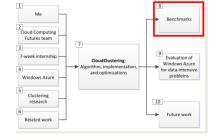
CloudClustering: **Data Affinity**



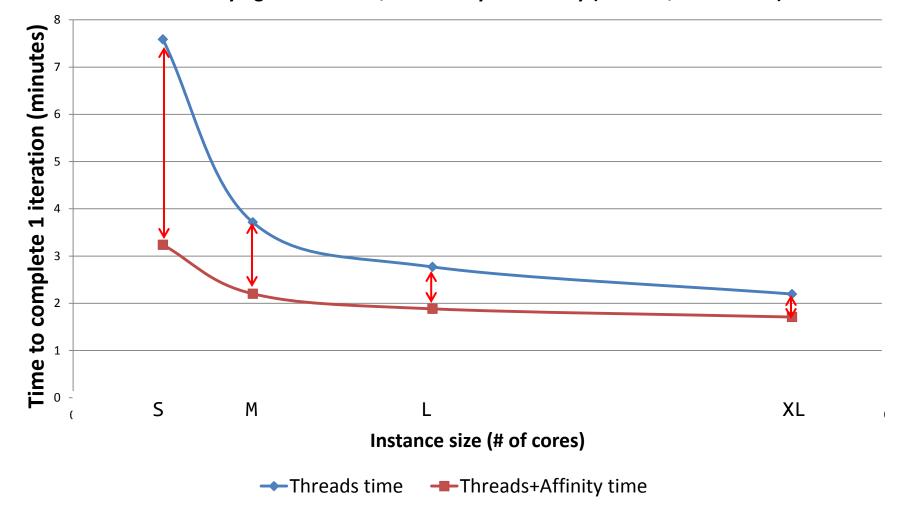




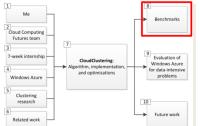
CloudClustering: Data Affinity

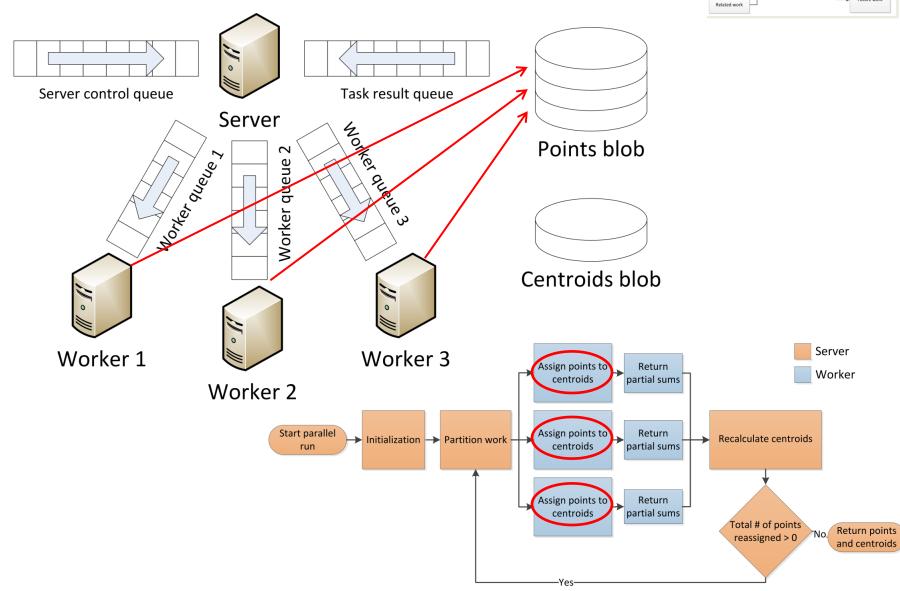


Time for varying instance size, No Affinity vs. Affinity (Threads, 4 instances)



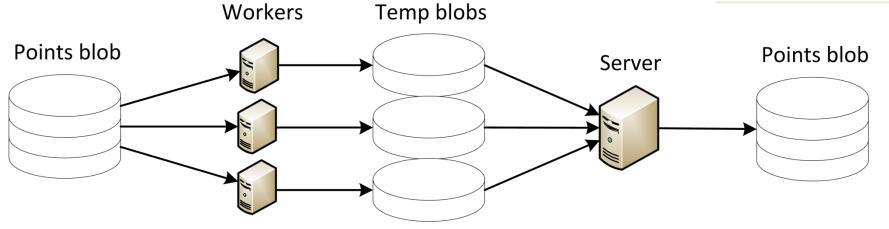
CloudClustering: Efficient Blob Concatenation

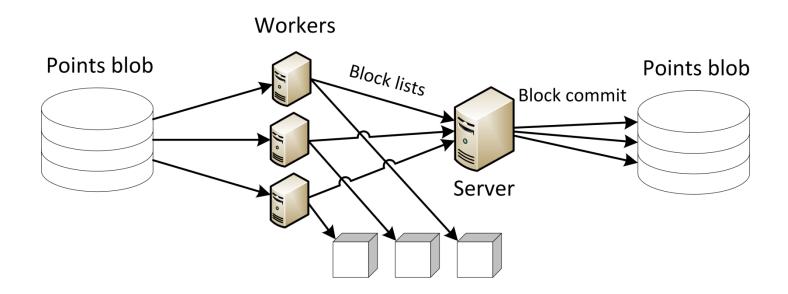




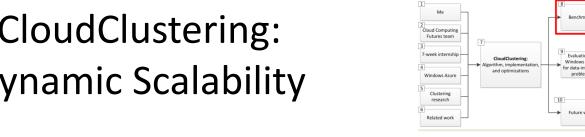
CloudClustering: Efficient Blob Concatenation

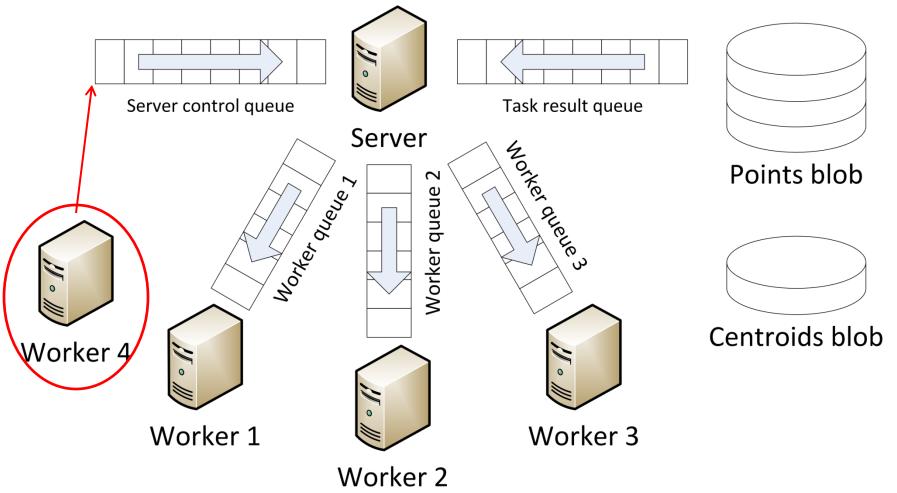




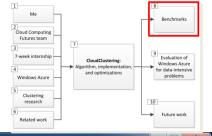


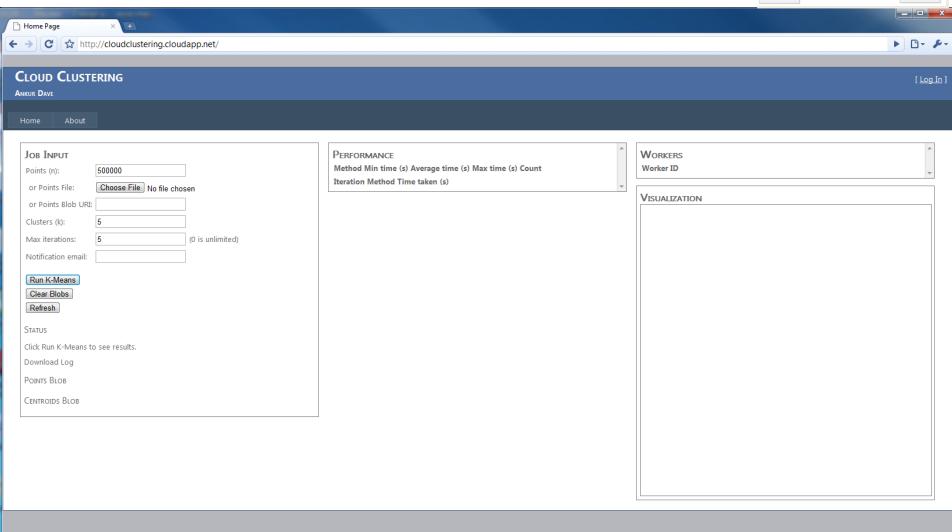
CloudClustering: **Dynamic Scalability**



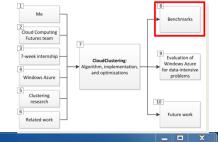


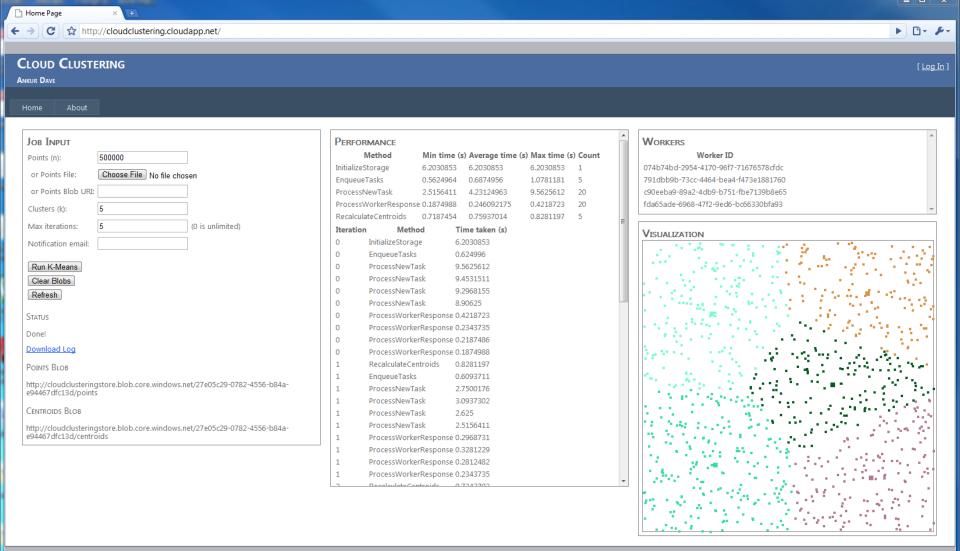
CloudClustering: Demo



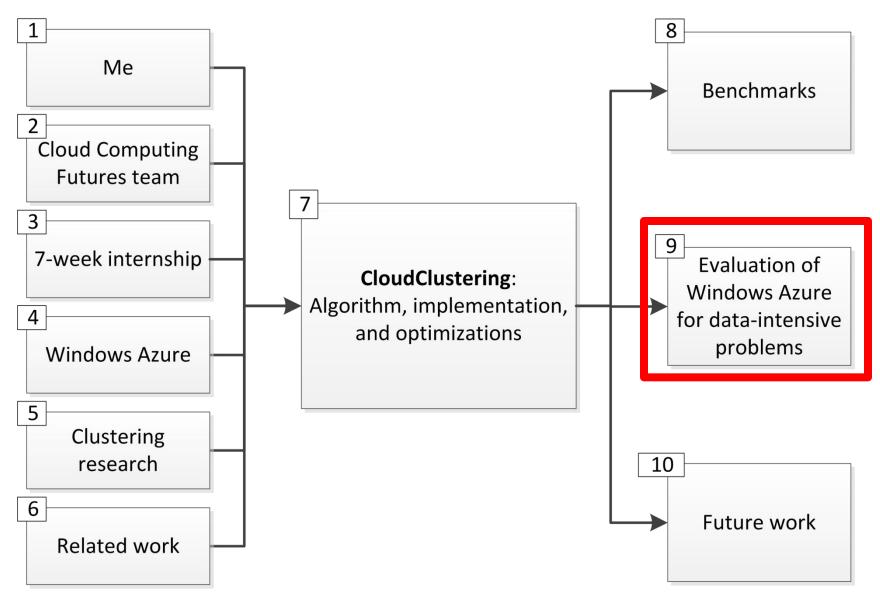


CloudClustering: Demo

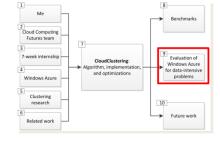




Agenda



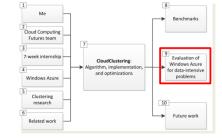
Windows Azure: Benefits





Azure is an **appropriate level of abstraction** for data-intensive algorithms like k-means.

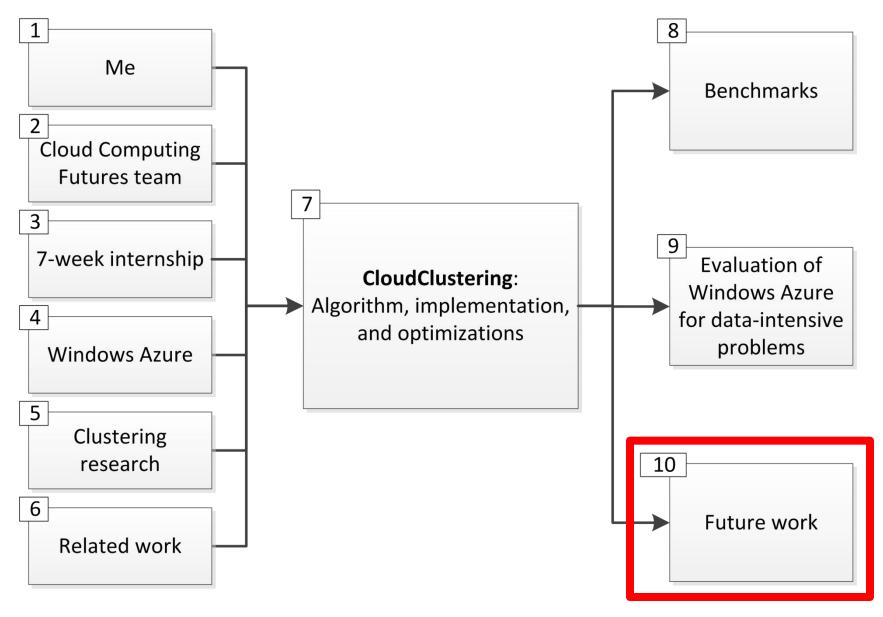
Windows Azure: Potential Problem Areas



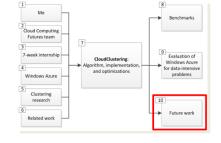
- On the cloud, cost scales directly with usage
 - Sub-linear speedups are not good enough!
- For data-intensive algorithms, data affinity gives great performance... but there's a tradeoff
 - Dynamic scaling is more complex
 - Fault-tolerance is even harder

Performance test to find configuration sweet spots

Agenda



Future work



- A compromise between worker pools and data affinity that retains scalability and fault-tolerance
 - Buddy system
- Improved caching using blocks
- Fundamental improvements to the k-means algorithm
 - More efficient stopping condition
 - "Lazy" processing that eliminates synchronization barriers
- Further optimizations to multicore parallelism

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