Technology for Informatics Kmeans and MapReduce Parallelism

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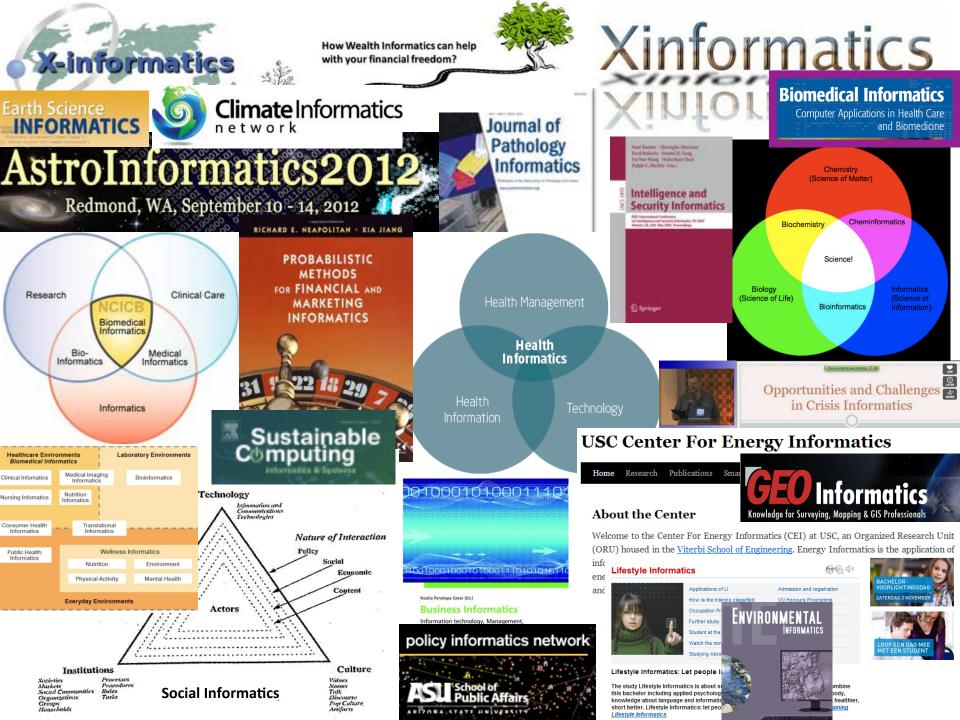
2013

Big Data Ecosystem in One Sentence

Use Clouds running Data Analytics Collaboratively processing Big Data to solve problems in X-Informatics (or e-X)

X = Astronomy, Biology, Biomedicine, Business, Chemistry, Climate, Crisis, Earth Science, Energy, Environment, Finance, Health, Intelligence, Lifestyle, Marketing, Medicine, Pathology, Policy, Radar, Security, Sensor, Social, Sustainability, Wealth and Wellness with more fields (physics) defined implicitly Spans Industry and Science (research)

Education: Data Science see recent New York Times articles http://datascience101.wordpress.com/2013/04/13/new-york-times-data-science-articles/



MapReduce Kmeans in Python

Slightly Changed Sequential version

- We change to allow MapReduce implementation
- Overall iteration over starting positions
 - Initialize Centroids
 - Iterate until converged
 - Call map to find association and do first step in find centroids as sum over point vectors and distortion
 - Reduce step: Divide summed centers by points in center to get centroid
 - Do other book keeping
 - Delete zero size clusters
 - Check convergence
- Return best solution based on Quality criterion

Parallel MapReduce Version

```
Iseven = np.empty([tot], dtype=bool)
for i in arange(tot):
                Iseven[i] = (i\%2 == 0);
obs1 = compress(Iseven, obs, 0)
obs2 = compress(logical not(Iseven), obs, 0)
avg dist = []
diff = thresh+1.
                                                                                                                    Map
while diff > thresh:
                if Parallelism == 1:
                                 code book, NumPointsinClusters, distortsum, distortmax, NumPoints = Kmeans map(obs, code book)
                if Parallelism == 2:
                                 # Can be Parallel Map Operations
                                 code book1, NumPointsinClusters1, distortsum1, distortmax1, NumPoints1 = Kmeans map(obs1, code book)
                                 code book2, NumPointsinClusters2, distortsum2, distortmax2, NumPoints2 = Kmeans map(obs2, code book)
                                                  Following are 4 Reduction Operations
                                                  Note maps include local reductions
                                 code book = np.add( code book1, code book2)
                                 NumPointsinClusters = np.add( NumPointsinClusters1, NumPointsinClusters2)
                                 distortsum = distortsum1 + distortsum2
                                 distortmax = np.maximum(distortmax1, distortmax2)
                                                                                                                    Reduce
                                 NumPoints = NumPoints1 + NumPoints2
                code book = compress(np.greater(NumPointsinClusters, 0), code book, 0)
                                 remove code books that didn't have any members
                i = 0
                nc = code book.shape[0]
                for i in arange(nc):
                                 if NumPointsinClusters[i] > 0:
                                                  code book[j,:] = code book[j,:] / NumPointsinClusters[i]
                                                  i = i + 1
```

Map for Kmeans

```
def Kmeans map(obs, code book):
        No = obs.shape[0]
        nc = code book.shape[0]
        # nc is current number of clusters (may decrease if zero clusters last iteration)
        #compute membership and distances between obs and code book
        obs code, distort = vg(obs, code book)
        distortsum = np.sum(distort)
        distortmax = np.amax(distort)
        # vg returns an indexing array obs code mapping rows of obs (the points) to code book (the centroids)
        # distort is an array of length No that has difference between observation and chosen centroid
        # vg stands for vector quantization and is provided in SciPy
        VectorDimension = obs.shape[1]
        NewCode Book = np.zeros([nc, VectorDimension])
        NumPointsinClusters = np.zeros([nc])
        for i in arange(nc):
                                  Loop over clusters labelled with i
                     cell members = compress(equal(obs code, i), obs, 0)
                     NumPointsinClusters[i] = cell members.shape[0]
                                   Extract Points in this Cluster; extract points whose quantization label is i
                     NewCode Book[i] = np.sum(cell members, 0)
                                   Calculate centroid of i'th cluster
        return NewCode Book, NumPointsinClusters, distortsum, distortmax, No
```

- This routine takes Points in obs and Centroids in code_book and associates each point with nearest centroid
- It calculates Non normalized centroids (sum over vectors in cluster), Number of points in each cluster and convergence measures (mean and max of distance between points and centers)