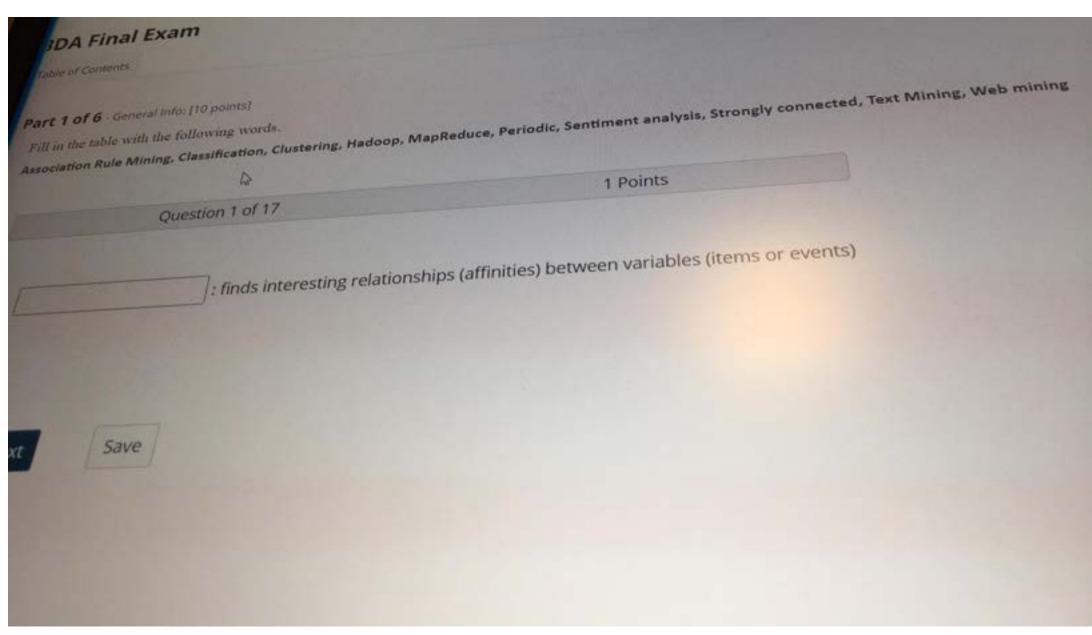
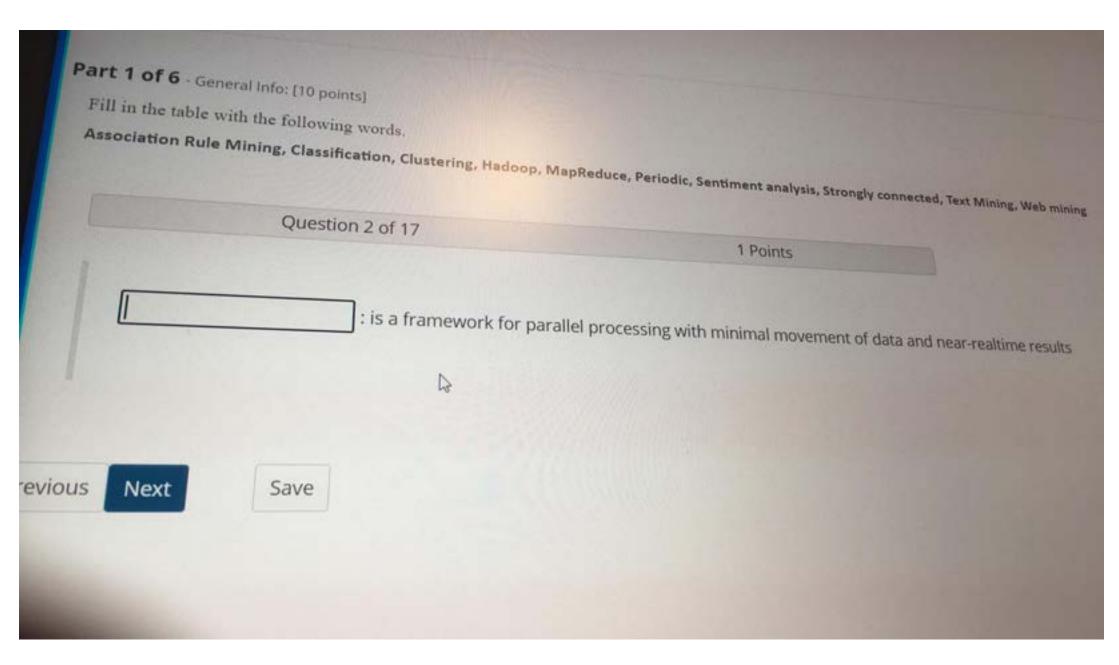
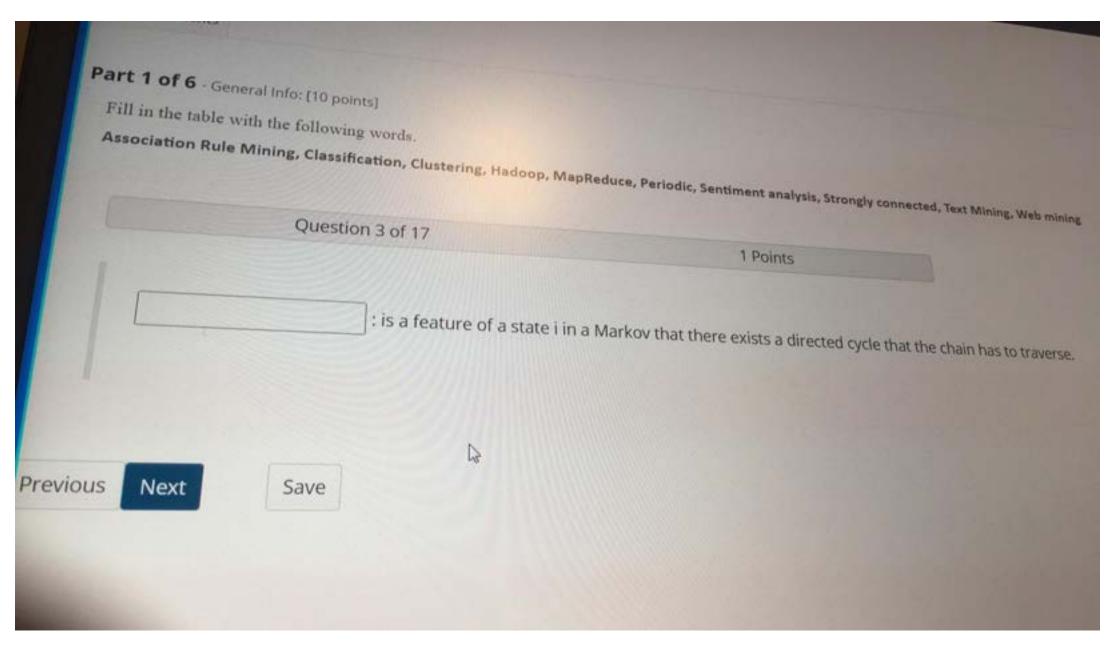
Question 1. General Info - Fill in blank

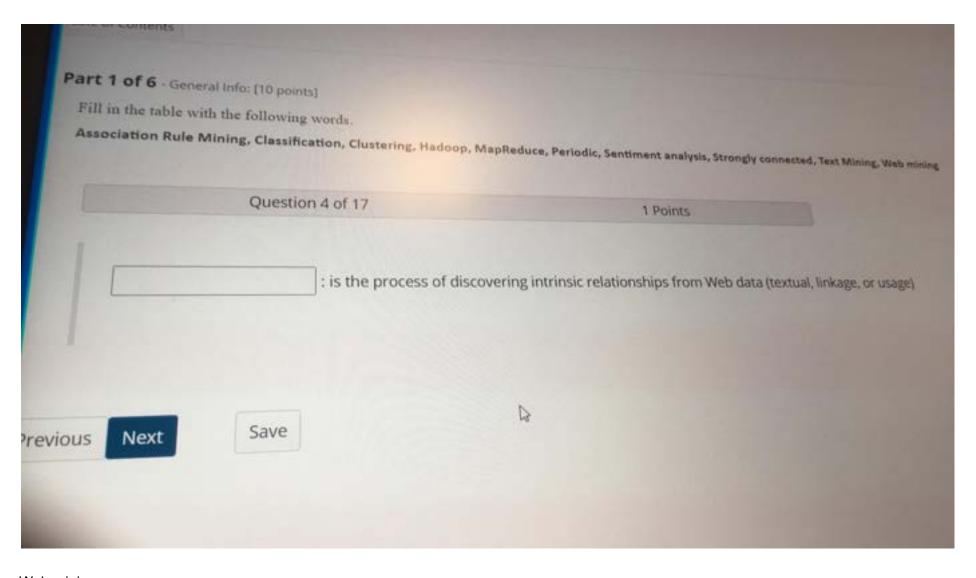


Association rule mining

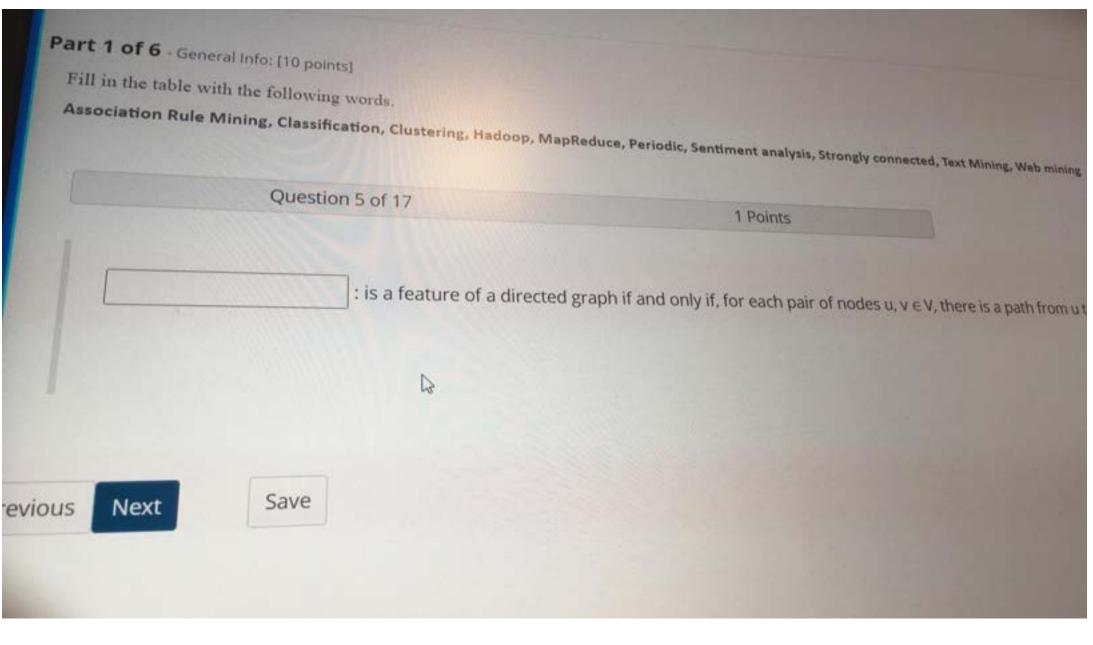




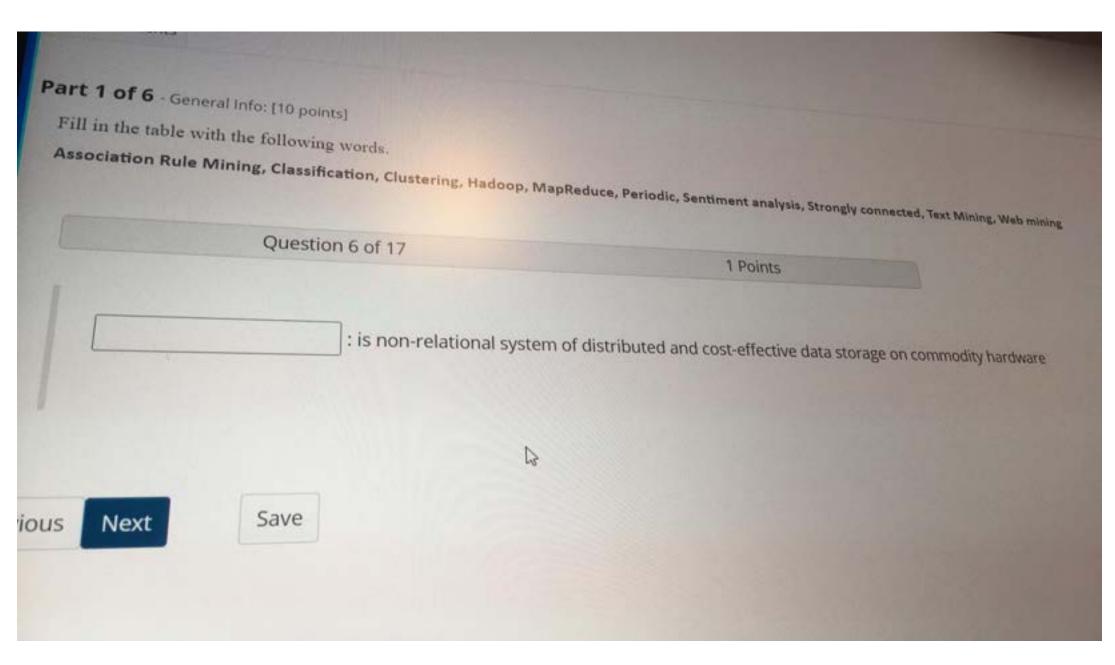
periodic



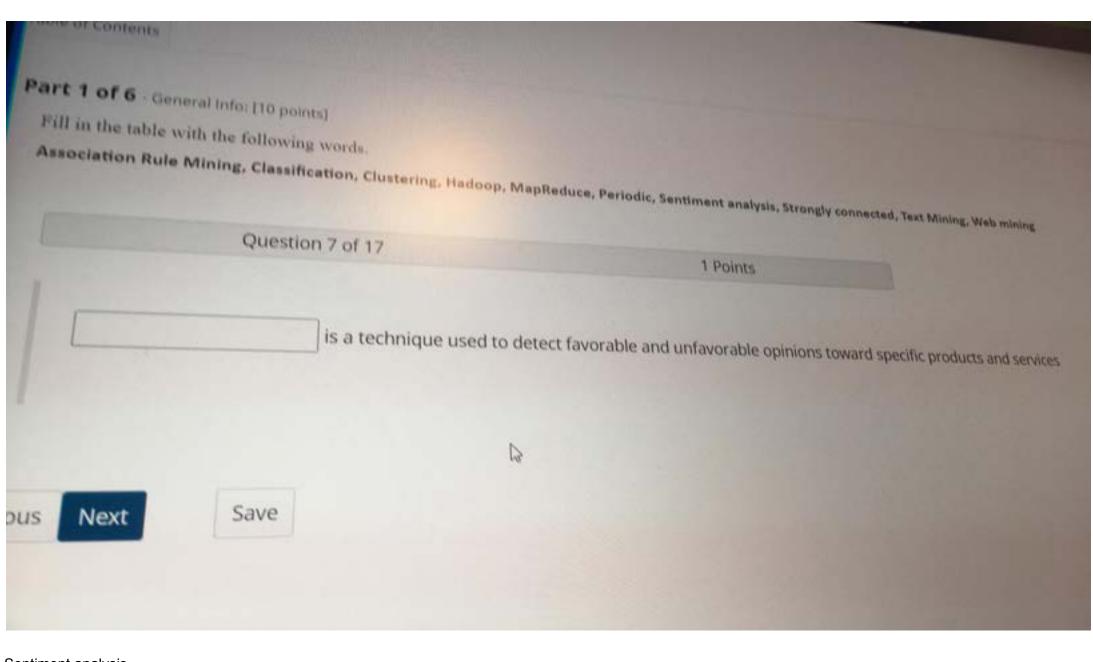
Web mining



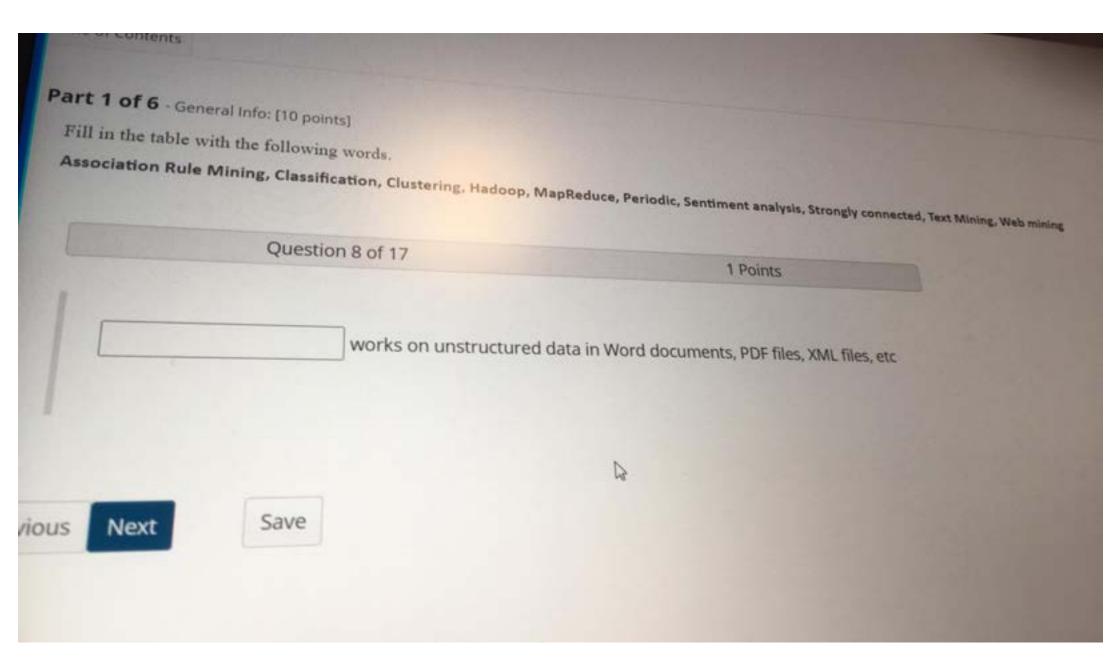
Strongly connected



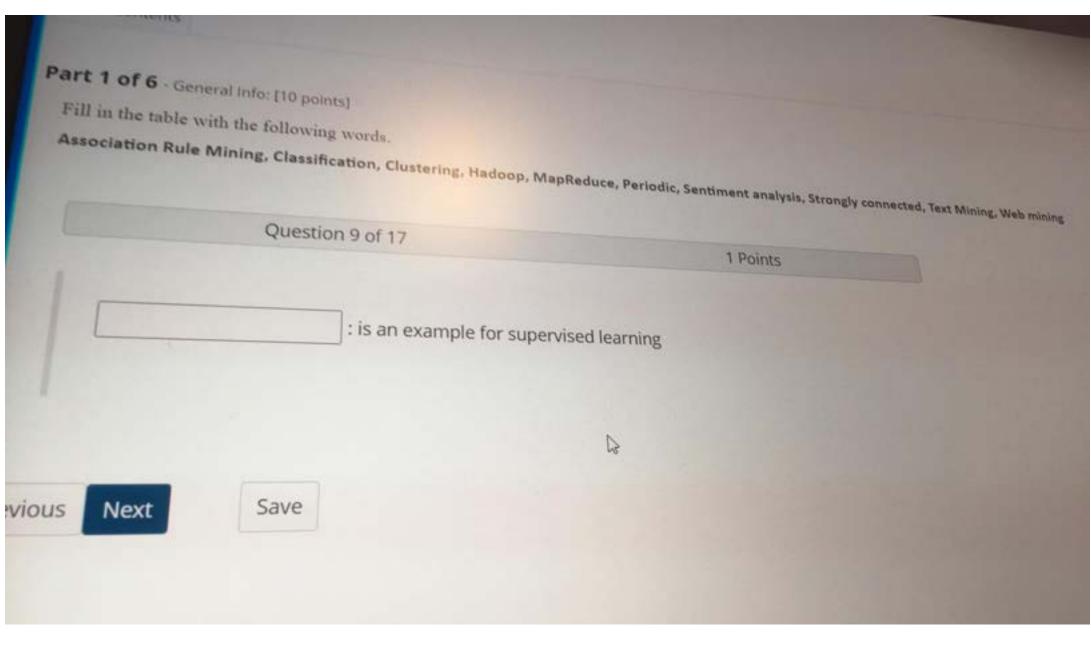
Hadoop



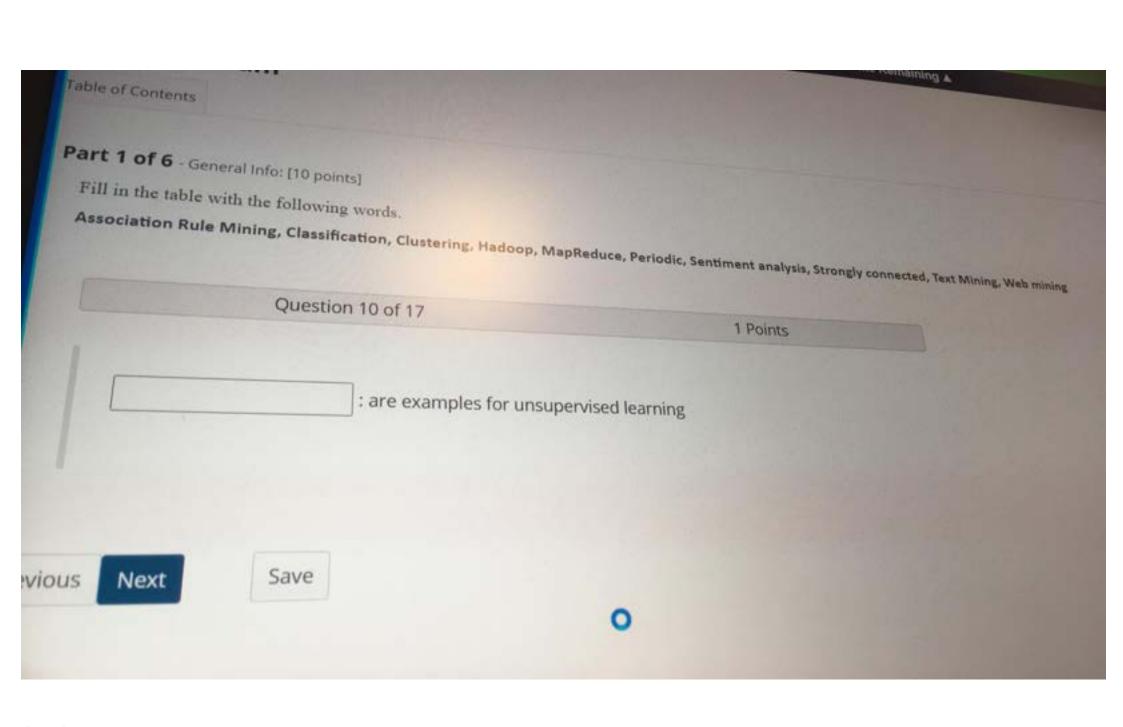
Sentiment analysis



Text mining



Classification



Question 2. Hadoop - Map Reduce

User ID	Restaurant ID	Rating	City ID	ive the name of each stage and list all elements in each stage,
124	294	2	985	
349	827	4	998	
725	751	4	982	
346	294	2	985	
578	827	3	998	
124	934	4	051	
725	294	3	985	
766	751	5	982	
725	294	2	985	
766	294	1	985	13

```
Input --> Splitting --> Mapping --> Shuffling --> Reducing --> Final result
=== Input Rate occurrence
Rate2 Rate4 Rate4 Rate2 Rate3 Rate4 Rate3 Rate5 Rate2 Rate1
=== Splitting phase
Splitter1: Rate2 Rate4 Rate4 Rate2 Rate3 Rate4 Rate3 Rate5 Rate2 Rate1
=== Mapping phase
Mapper1: ('Rate1', 1), ('Rate2', 3), ('Rate3', 2), ('Rate4', 3), ('Rate5', 1)
=== Shuffling phase
Shuffler1: ('Rate1', 1)
Shuffler2: ('Rate2', 3)
Shuffler3: ('Rate3', 2)
Shuffler4: ('Rate4', 3)
Shuffler5: ('Rate5', 1)
=== Reducing phase
Reducer1: Rate1,1
Reducer2: Rate2,3
Reducer3: Rate3,2
Reducer4: Rate4,3
Reducer5: Rate5,1
=== Final result
('Rate1', 1), ('Rate2', 3), ('Rate3', 2), ('Rate4', 3), ('Rate5', 1)
```

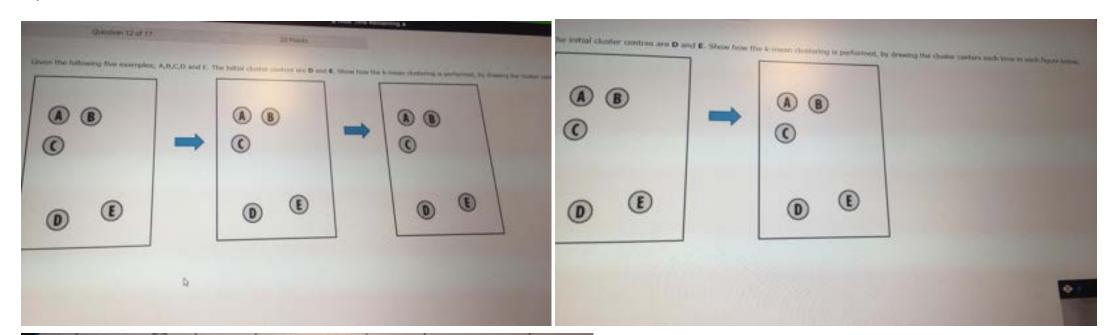
Input split>an input to a MapReduce is split into fixed-size parts called input splits. The term "input split" refers to a portion of the input that is processed by a single map.

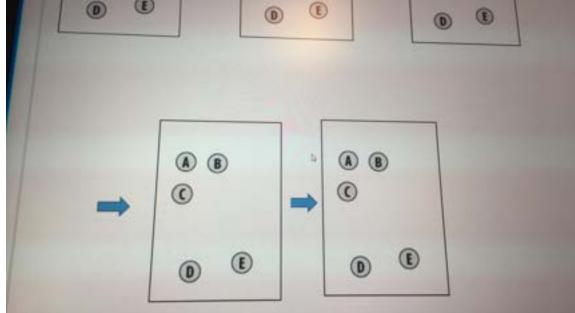
Mapping>In this step, each split's data is handed to a mapping function, which generates output values.

Shuffling>This step consumes the Mapping phase's output. Its job is to bring together all of the relevant records from the Mapping phase's output.

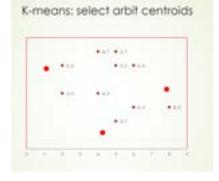
Reducing>The output values from the Shuffling step are consolidated in this phase. This phase takes the values from the Shuffling phase and merges them into a single output value. In a nutshell, this stage summarizes the whole dataset.

Question 3. Cluster - Kmean

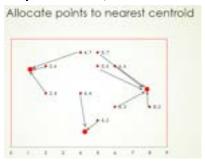




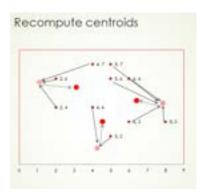
Step 1: centrioids D, E, calculate distance from A,B, C to divide the clusters



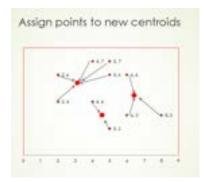
Step 2 AC to D, B to E because the AC closer to D than E, B closer to E than D



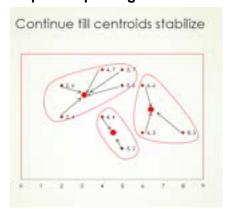
Step 3 Calculate new D' E'



Step 4 - Reassigning points to new centroids D' E'



Step 5 - Repeating until controids stabilize



=== Input

hairbrush shampoo

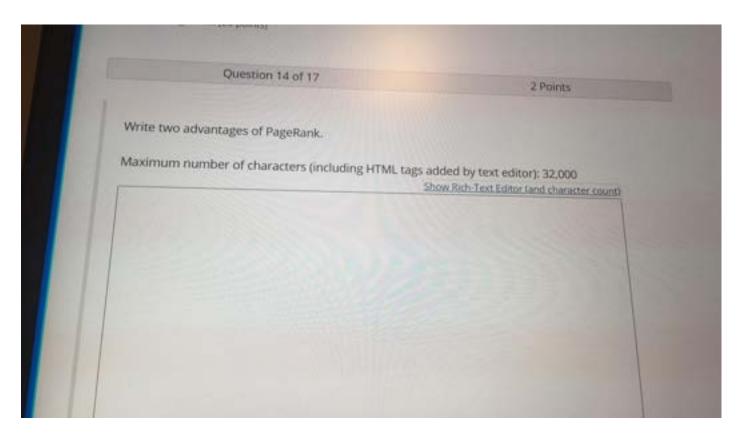
Question 4. Association Rule - Apriori

```
hairdryer shampoo haircond
mirror hairdryer shampoo
hairbrush hairdryer shampoo
mirror hairdryer shampoo haircond
shampoo hairdryer
hairbrush shampoo haircond hairdryer
shampoo mirror haircond
mirror shampoo haircond hairbrush
hairbrush hairdryer mirror
=== raw items: ['hairbrush', 'haircond', 'hairdryer', 'mirror', 'shampoo']
total trans=10, support*trans =2.5
=== 1-item itemsets
hairbrush 5
haircond
hairdryer 7
mirror
shampoo
=== 2-item itemsets
hairbrush, hairdryer
hairbrush, shampoo
haircond, hairdryer
haircond, mirror 3
haircond, shampoo5
hairdryer, mirror3
hairdryer, shampoo
mirror, shampoo 4
---> drop itemsets
      hairbrush, haircond
      hairbrush, mirror 2
=== 3-item itemsets
haircond, hairdryer, shampoo
haircond, mirror, shampoo
---> drop itemsets
      hairbrush, haircond, hairdryer
      hairbrush, haircond, mirror
      hairbrush, haircond, shampoo
      hairbrush, hairdryer, mirror
      hairbrush, hairdryer, shampoo 2
      hairbrush, mirror, shampoo
      haircond, hairdryer, mirror
                                   1
      hairdryer, mirror, shampoo
=== 4-item itemsets
NO 4-item itemsets
---> drop itemsets
      hairbrush, haircond, hairdryer, mirror
      hairbrush, haircond, hairdryer, shampoo
      hairbrush, haircond, mirror, shampoo 1
      hairbrush, hairdryer, mirror, shampoo
      haircond, hairdryer, mirror, shampoo 1
```

ose Aprior	Algorithm to find	all frequent itemsets with minimum support	percentage of 25%. List the itemsets together with	their support
	Transaction ID	Items		
	1	Hairbrush, Shampoo		
	2	Hair dryer, Shampoo, Hair conditioner		
	3	Mirror, Hair dryer, Shampoo, Hair conditioner		
	4	Hairbrush, Hair dryer, Shampoo		
	5	Mirror, Hair dryer, Shampoo, Hair conditioner		
	6	Shampoo, Hair dryer		
	7	Hairbrush, Shampoo, Hair conditioner, Hair dryer	D	
	8	Shampoo, Mirror, Hair conditioner		
	9	Mirror, Shampoo, Hair conditioner, Hairbrush		
-	10	Hairbrush, Hair dryer, Mirror		

Question 5. PageRank

https://colab.research.google.com/drive/1ZE0SMPDozimHKoEDImH60DJ8yYA5peDS?usp=sharing

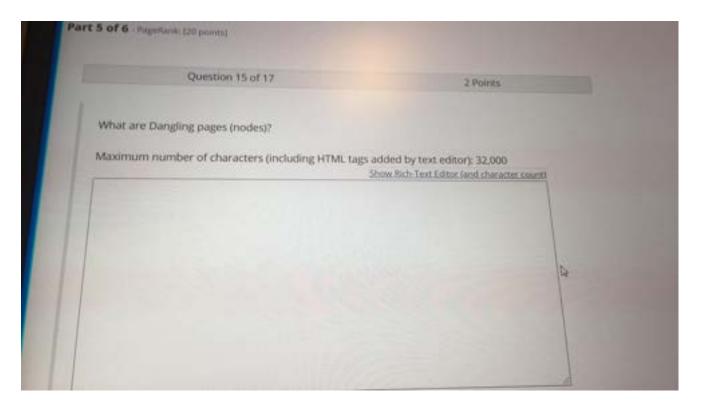


Choose 2 of 5 - dont write out all

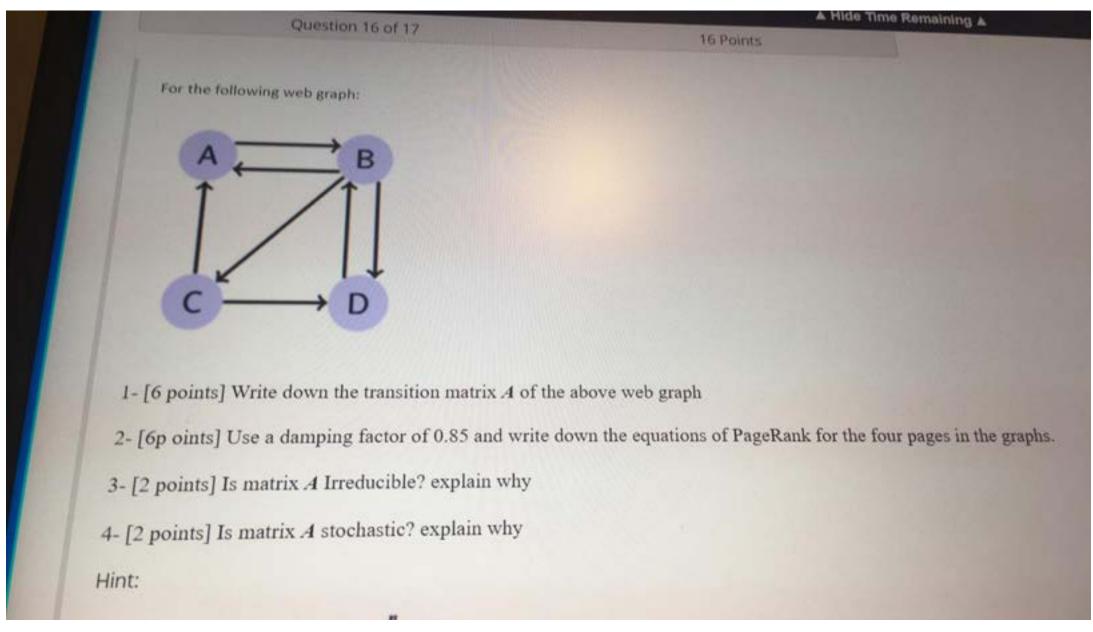
- 1. Since it pre-computes the rank score it takes less time and hence it is fast.
- 2. It is more feasible as it computes rank score at indexing time not at query time
- 3. It returns important pages as Rank is calculated on the basis of the popularity of a page

Hey guys this is for weighted page rank algorithm, not the page rank algorithm, so we probably should not include it

- 4. Quality of the pages returned by this algorithm is high as compared to PageRank algorithm.
- 5. It is more efficient than PageRank because rank value of a page is divided among it's outlink pages according to importance of that page.



Dangling links are simply links that point to any page with no outgoing links. They affect the model because it is not clear where their weight should be distributed, and there are a large number of them.



```
Input
B -> C
B -> D
C -> A
C \rightarrow D
D -> B
Influence relationship
Ra = 0.33Rb + 0.5Rc
Rb = Ra + Rd
Rc = 0.33Rb
Rd = 0.33Rb + 0.5Rc
===>> Transition Matrix A
[[0. 0.33333333 0.5
                                  0.
                                          ]
            0. 0.
[1.
                                  1.
                                            ]
[0.
            0.33333333 0.
                                  0.
                                            ]
 [0.
            0.33333333 0.5
                                  0.
                                            ]]
=== Page Rank with damping factor = 0.85, init value = 1
PRa = 0.15 + 0.85*(0.33Rb + 0.5Rc)
PRb = 0.15 + 0.85*(Ra + Rd)
PRc = 0.15 + 0.85*(0.33Rb)
PRd = 0.15 + 0.85*(0.33Rb + 0.5Rc)
```

```
Khong can ghi tinh toan iteration
iter1 = [0.858, 1.85, 0.433, 0.858]
iter2 = [0.858, 1.609, 0.674, 0.858]
iter3 = [0.892, 1.609, 0.606, 0.892]
iter4 = [0.863, 1.667, 0.606, 0.863]
iter5 = [0.88, 1.618, 0.622, 0.88]
iter6 = [0.873, 1.646, 0.608, 0.873]
iter7 = [0.875, 1.634, 0.616, 0.875]
iter8 = [0.875, 1.637, 0.613, 0.875]
```

A **matrix** is **reducible** if and only if it can be placed into block upper-triangular form by simultaneous row/column permutations. In addition, a **matrix** is **reducible** if and only if its associated digraph is not strongly connected. A square **matrix** that is not **reducible** is said to be **irreducible**.

In mathematics, a **stochastic matrix** is a square matrix used to describe the transitions of a Markov chain. Each of its entries is a nonnegative real number representing a probability. It is also called a probability matrix, transition matrix, substitution matrix, or Markov matrix.

