## COS 435, Spring 2017 - Problem Set 5

Due 11:59 pm Wednesday April 12, 2017 by DropBox submission

## Collaboration and Reference Policy

You may discuss the general methods of solving the problems with other students in the class. However, each student must work out the details and write up his or her own solution to each problem independently. For each problem, list the students with whom you discussed general methods of solving the problem.

Some problems have been used in previous offerings of COS 435. You are NOT allowed to use any solutions posted for previous offerings of COS 435 or any solutions produced by anyone else for the assigned problems. You may use other reference materials; you must give citations to all reference materials that you use.

## Lateness Policy

A late penalty will be applied, unless there are extraordinary circumstances and/or prior arrangements:

- Penalized $10 \%$ of the earned score if submitted by 10am Thursday $(4 / 13 / 17)$.
- Penalized $25 \%$ of the earned score if submitted by $4: 30 \mathrm{pm}$ Friday $(4 / 14 / 17)$.
- Penalized $50 \%$ if submitted later than $4: 30$ pm Friday (4/14/17).


## Submission

Submit your solutions as a PDF file using the Computer Science Department DropBox submission system for COS435 at https://dropbox.cs.princeton.edu/COS435_S2017/HW5 Name your file HW5.pdf. If you have not used this facility before, consult the instructions at https://csguide.cs.princeton.edu/academic/csdropbox - student Note that you are automatically enrolled in CS DropBox using the registrar's COS435 enrollment list.

Problem 1 (similar to an old exam problem)
On the next page is the $5 \times 7$ term-document matrix $C$ for a set of documents under the "set of terms" model (contains/doesn't contain) and the matrices $U, \Sigma$, and $V^{\mathrm{T}}$ that make up the singular value decomposition of C .

Part a. Give the matrices of the rank-three approximation of C. That is, give $\mathrm{U}_{3}^{\prime}, \Sigma^{\prime}{ }_{3}$, and $\mathrm{V}_{3}^{\prime}{ }^{\mathrm{T}}$.

Part b. What is the 3-dimensional representation of Doc 5 for the rank-three approximation?

Part c. What is the 3-dimensional representation of term "cat" for the rank-three approximation?

Part d. In the 3-dimensional representation, what is the similarity of "cat" and "cow"? of "cat" and "dog"? What are the dot product similarities of the original representations of these terms as given in matrix C?

Note: on slide 4 of the slides on Latent Semantic Indexing, I state U is an M×r matrix, V is an $N \times r$ matrix and $\Sigma$ is an $r \times r$ diagonal matrix. In this problem statement, $U$ is given as an $\mathrm{M} \times \mathrm{M}$ matrix, V as an $\mathrm{N} \times \mathrm{N}$ matrix and $\Sigma$ as an $\mathrm{M} \times \mathrm{N}$ diagonal matrix. The larger dimension matrices are actually what is produced in computing SVD. They are what is shown in the on the next page. In the lecture notes, I truncate the matrices in the same way we do to get $\mathrm{M}_{\mathrm{k}}, \mathrm{N}_{\mathrm{k}}$, and $\Sigma_{\mathrm{k}}$ (slide 7) because I know only the first r singular values are non-zero (as indicated in the diagram on slide 4). For the matrix C below, r=5.

## For Problem 1:

$C=$

|  | Doc 1 | Doc 2 | Doc 3 | Doc 4 | Doc 5 | Doc 6 | Doc 7 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| cat | 1 | 0 | 1 | 1 | 0 | 0 | 1 |
| cow | 0 | 1 | 0 | 0 | 1 | 1 | 0 |
| dog | 1 | 0 | 1 | 1 | 0 | 1 | 1 |
| pig | 0 | 1 | 1 | 0 | 1 | 1 | 1 |
| rabbit | 1 | 1 | 1 | 0 | 0 | 1 | 0 |


| -0.40972 | 0.54966 | -0.19439 | -0.19354 | 0.67436 |
| :---: | :---: | :---: | :---: | :---: |
| -0.27231 | -0.59693 | -0.05582 | 0.58538 | 0.47301 |
| -0.53695 | 0.39476 | -0.03386 | 0.55283 | -0.49909 |
| -0.51397 | -0.40542 | -0.51446 | -0.48414 | -0.26907 |
| -0.45332 | -0.14614 | 0.83263 | -0.28260 | 0.00260 |


$\Sigma=$|  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
|  |  |  |  |  |  |  |
| 3.73682 | 0.00000 | 0.00000 | 0.00000 | 0.00000 | 0.00000 | 0.00000 |
| 0.00000 | 2.20586 | 0.00000 | 0.00000 | 0.00000 | 0.00000 | 0.00000 |
| 0.00000 | 0.00000 | 1.19304 | 0.00000 | 0.00000 | 0.00000 | 0.00000 |
| 0.00000 | 0.00000 | 0.00000 | 0.70548 | 0.00000 | 0.00000 | 0.00000 |
| 0.00000 | 0.00000 | 0.00000 | 0.00000 | 0.49931 | 0.00000 | 0.00000 |


| $\mathrm{V}^{\mathrm{T}}=$ |  |  |  |  |  |  |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |  |
| -0.37465 | -0.33173 | -0.51219 | -0.25333 | -0.21041 | -0.47542 | -0.39088 |
| 0.36189 | -0.52065 | 0.17810 | 0.42814 | -0.45440 | -0.34169 | 0.24435 |
| 0.50659 | 0.21990 | 0.07536 | -0.19132 | -0.47801 | 0.19151 | -0.62254 |
| 0.10871 | -0.25707 | -0.57755 | 0.50928 | 0.14350 | 0.52655 | -0.17698 |
| 0.35622 | 0.41365 | -0.18266 | 0.35102 | 0.40844 | -0.58592 | -0.18787 |
| 0.50226 | -0.49771 | -0.00455 | -0.50226 | 0.49771 | 0.00000 | 0.00455 |
| 0.28473 | 0.29260 | -0.57733 | -0.28473 | -0.29260 | 0.00000 | 0.57733 |

computed with www.dotnumerics.com/MatrixCalculator verified using http://comnuan.com/cmnn01004/

## Problem 2:

Slide \#26 of Part 2 of the slides for clustering, posted under April 3, presents an iterative improvement algorithm for divisive partitioning. This problem addresses recalculating the total relative cut cost (slides \#20 and \#21) incrementally for use with that algorithm.

Let U denote the set of objects to be clustered. Assume that for any objects v and w , $\operatorname{sim}(\mathrm{v}, \mathrm{w})=\operatorname{sim}(\mathrm{w}, \mathrm{v})$ (we have been assuming this in class). Also assume that for any object $\mathrm{v}, \operatorname{sim}(\mathrm{v}, \mathrm{v})=0$. Let $\mathrm{C}_{\mathrm{p}}$ be an arbitrary cluster containing object $\mathrm{x}, \mathrm{C}_{\mathrm{q}}$ be an arbitrary cluster that does not contain $x$. (The set notation $C_{p}-\{x\}$ denotes $C_{p}$ with $x$ removed, and $\mathrm{C}_{\mathrm{q}} \mathrm{U}\{\mathrm{x}\}$ denotes $\mathrm{C}_{\mathrm{q}}$ with x added.)

The following relationship holds for incremental changes to the intracost of a cluster when removing or adding an object x .

$$
\begin{array}{rlr}
\operatorname{intracost}\left(C_{p}\right)-\operatorname{intracost}\left(C_{p}-\{x\}\right) & =\sum_{\mathrm{v}_{\mathrm{i}} \text { in } C_{p}-\{x\}} \operatorname{sim}\left(\mathrm{v}_{\mathrm{i}}, x\right) \\
& =\sum_{\mathrm{v}_{\mathrm{i}} \text { in } C_{p}} \operatorname{sim}\left(\mathrm{v}_{\mathrm{i}}, x\right) \quad \text { since } \operatorname{sim}(\mathrm{x}, \mathrm{x})=0
\end{array}
$$

From this relationship we derive the incremental cost changes for intracost:

$$
\begin{aligned}
& \operatorname{intracost}\left(C_{p}-\{x\}\right)=\operatorname{intracost}\left(C_{p}\right)-\sum_{v_{i} \text { in }} \operatorname{Cim}\left(\mathrm{V}_{\mathrm{p}}, x\right) \\
& \operatorname{intracost}\left(\mathrm{C}_{\mathrm{q}} \cup\{\mathrm{x}\}\right)=\operatorname{intracost}\left(\mathrm{C}_{\mathrm{q}}\right)+\sum_{\mathrm{v}_{\mathrm{i}} \text { in } C_{q}} \operatorname{sim}\left(\mathrm{v}_{\mathrm{i}}, x\right)
\end{aligned}
$$

Your task is to derive incremental cost equations for cutcost. The ultimate goal is to minimize the computation time used by the iterative improvement algorithm.

## Part a:

i. Give an equation for $\operatorname{cutcost}\left(\mathbf{C}_{\mathbf{p}}\right)$-cutcost $\left(\mathbf{C}_{\mathbf{p}}-\{\mathbf{x}\}\right)$ when x is an object in $\mathrm{C}_{\mathrm{p}}$. Your equation should be in terms of similarities between $x$ and other objects.

Using your equation, derive equations for
ii. $\quad \operatorname{cutcost}\left(\mathrm{C}_{\mathrm{p}}-\{\mathrm{x}\}\right)$ as an incremental change to $\operatorname{cutcost}\left(\mathrm{C}_{\mathrm{p}}\right)$;
iii. $\operatorname{cutcost}\left(\mathrm{C}_{\mathrm{q}} \mathrm{U}\{\mathrm{x}\}\right)$ as an incremental change to $\operatorname{cutcost}\left(\mathrm{C}_{\mathrm{q}}\right)$.

Part b: Given the equations for the incremental changes in intracost and cutcost, what is the computational time complexity of the step:
move $v_{j}$ to that cluster, if any, such that move gives maximum decrease in cost of the iterative improvement algorithm on slide \#25? Specify the data structures you are using and how they are used to achieve the time complexity.

