

# CSE 515 Multimedia and Web Databases

## Phase #2

(Due October 15 2023, midnight)

**Description:** In this project, you will experiment with

- image features,
- vector models,
- dimensionality curse,
- graph analysis.

### NOTES:

- You can use existing libraries for LDA decomposition.
- You can use existing libraries CP decomposition.
- You can use existing libraries for eigenvector and eigenvalue extraction.
- The tasks in this phase involve the feature models and similarity/distance functions developed in the previous phase.

### PROJECT TASKS:

- **Task 0:**

- **Task 0a:** Using pre-trained RESNET50 neural network model, map even numbered (labeled) images in the Caltec101 data set into 5 different feature spaces and store the resulting data vectors:

- \* Color moments, CM10x10
    - \* Histograms of oriented gradients, HOG
    - \* ResNet-AvgPool-1024
    - \* ResNet-Layer3-1024
    - \* ResNet-FC-1000

In the database, store not only the imageIDs and feature vectors, but also the original image labels.

- **Task 0b:** Implement a program which, given (a) an (even or odd numbered) imageID or an image file, (b) a user selected feature space, and (c) positive integer  $k$ , identifies and visualizes the most similar  $k$  images, along with their scores, under the selected feature space.

- **Task 1:** Implement a program which, given (a) a query label,  $l$  (b) a user selected feature space, and (c) positive integer  $k$ , identifies and visualizes the most relevant  $k$  images for the given label  $l$ , along with their scores, under the selected feature space.

- **Task 2:**

- **Task 2a:** Implement a program which, given (a) a query imageID or image file, (b) a user selected feature space, and (c) positive integer  $k$ , identifies and lists  $k$  most likely matching labels, along with their scores, under the selected feature space.
- **Task 2b:** Implement a program which, given (a) a query imageID or image file and (b) positive integer  $k$ , identifies and lists  $k$  most likely matching labels, along with their scores, under the RESNET50 neural network model.
- **Task 3 (LS1):** Implement a program which (a) given one of the feature models, (b) a user specified value of  $k$ , (c) one of the four dimensionality reduction techniques (SVD, NNMF, LDA, k-means) chosen by the user, reports the top- $k$  latent semantics extracted under the selected feature space.
  - Store the latent semantics in a properly named output file
  - List imageID-weight pairs, ordered in decreasing order of weights
- **Task 4 (LS2):** Implement a program which (a) given one of the feature models, (b) a user specified value of  $k$ , reports the top- $k$  latent semantics extracted using CP-decomposition of a three modal (image-feature-label) tensor under the selected features space. Each latent semantic should be presented in the form of a list of label-weight pairs, ordered in decreasing order of weights.
  - Store the latent semantics in a properly named output file
  - List label-weight pairs, ordered in decreasing order of weights
- **Task 5 (LS3):** Implement a program which, (a) given one of the feature models and (b) a value  $k$ ,
  - creates (and saves) a label-label similarity matrix,
  - performs a user selected dimensionality reduction technique (SVD, NNMF, LDA, k-means) on this label-label similarity matrix,
  - stores the latent semantics in a properly named output file
  - lists label-weight pairs, ordered in decreasing order of weights
- **Task 6 (LS4):** Implement a program which, (a) given one of the feature models and (b) a value  $k$ ,
  - creates (and saves) an image-image similarity matrix,
  - performs a user selected dimensionality reduction technique (SVD, NNMF, LDA, k-means) on this image-image similarity matrix
  - stores the latent semantics in a properly named output file
  - lists image-weight pairs, ordered in decreasing order of weights
- **Task 7:** Implement a program which, given (a) an (even or odd numbered) imageID or an image file name, (b) a user selected latent semantics, and (c) positive integer  $k$ , identifies and visualizes the most similar  $k$  images, along with their scores, under the selected latent space.
- **Task 8:** Implement a program which, given (a) an (even or odd numbered) imageID or an image file name, (b) a user selected latent semantics, and (c) positive integer  $k$ , identifies and lists  $k$  most likely matching labels, along with their scores, under the selected latent space.
- **Task 9:** Implement a program which, given (a) a label  $l$ , (b) a user selected latent semantics, and (c) positive integer  $k$ , identifies and lists  $k$  most likely matching labels, along with their scores, under the selected latent space.
- **Task 10:** Implement a program which, given (a) a label  $l$ , (b) a user selected latent semantics, and (c) positive integer  $k$ , identifies and lists  $k$  most relevant images, along with their scores, under the selected latent space.

- **Task 11** Implement a program which (a) given a feature model or latent space, (b) a value  $n$ , (c) a value  $m$ , and (d) a label  $l$ 
  - creates a similarity graph,  $G(V, E)$ , where  $V$  corresponds to the images in the database and  $E$  contains node pairs  $v_i, v_j$  such that, for each subject  $v_i, v_j$  is one of the  $n$  most similar images in the database in the given space
  - identifies the most significant  $m$  images (relative to the given label  $l$ ) using personalized PageRank measure. See Huang, S., Li, X., Candan, K. S., Sapino, M. L. (2016). Reducing seed noise in personalized PageRank. Social Network Analysis and Mining, 6(1), 1-25.

**Deliverables:**

- Your code (properly commented) and a README file.
- Your outputs for the provided sample inputs.
- A short report describing your work and the results.

Please place your code in a directory titled “Code”, the outputs to a directory called “Outputs”, and your report in a directory called “Report”; zip or tar all off them together and submit it through the digital dropbox.