

# Accelerated proximal gradient for robust PCA

(Due on Sep.-27-2019)

**Instruction:** Each student needs to submit the source code file and a report by Latex. In the report file, you can summarize what you observe or difficulties you have. **You will be evaluated based on correctness of your algorithms, accuracy and efficiency of algorithms, performance on applications, and also the report.** Compress your files into a single .zip file, name it as “MATP6960\_Assignment1\_YourName”, and send it to `optimization.rpi@gmail.com`

## 1 Robust principal component analysis

Let  $X$  be composed of a sparse matrix  $S$  and a low-rank matrix  $L$ . The robust PCA aims at finding  $S$  and  $L$ , given  $X$ . Using  $\ell_1$  norm to promote sparsity and nuclear norm to promote low-rankness, robust PCA can be modeled as

$$\min_{L,S} \|L\|_* + \lambda \|S\|_1, \text{ s.t. } L + S = X. \quad (1)$$

Here,  $\|L\|_*$  denotes the nuclear norm of  $L$  and equals the summation of its singular values, and  $\|S\|_1 = \sum_{i,j} |s_{ij}|$ . Penalizing the constraint into the objective, one can solve the penalty problem

$$\min_{L,S} \|L\|_* + \lambda \|S\|_1 + \frac{\beta}{2} \|L + S - X\|_F^2, \quad (2)$$

or a sequence of the above problem, where  $\beta > 0$  is the penalty parameter.

## 2 Requirements

Include every item below in a single report and attach your code. Use the provided datasets to test your code.

1. Develop three solvers for (2): one is by the proximal gradient, and another is by the accelerated proximal gradient with two versions discussed in class. Treat  $\lambda$  and  $\beta$  as an input of your solvers.

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2. Compare the two solvers by using the provided synthetic data and also Escalator video Dataset. Note that the video dataset is in 3D format. It contains 200 frames of size  $130 \times 160$ . You will need first reshape each frame into a column vector and form a  $20800 \times 200$  matrix  $X$ . Report how these solvers decrease the objective values in terms of iteration number of also actual running time. Reshape each column of  $L$  and  $S$  into a  $130 \times 160$  image and use `imshow` to show a few selected columns of  $L$  and  $S$ . You can also use `implay` to see how the foreground and background of the video are separated. Report what you observe.