

Semidefinite Programming

Applications in approximating NP-Complete problems & Matrix Completion

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Presentation Overview

- 1 Semidefinite Programming
- 2 Travelling Salesman
- 3 Matrix Completion
Overview
Relaxation
Fashion-MNIST
- 4 Referencing

Low rank matrices

Given an incomplete matrix, can we recover the missing values?

		-1		
			1	
1	1	-1	1	-1
1				-1
		-1		

1	1	-1	1	-1
1	1	-1	1	-1
1	1	-1	1	-1
1	1	-1	1	-1
1	1	-1	1	-1

Yes! Given:

- The matrix is low rank
- We have enough sample data
- and more ... (needle in haystack)

Why is this useful

- 1 **Netflix** has an incomplete set of user preferences based off their past watch history. They
- 2 **Recommendation Engine**
- 3 **Images**

Relaxing Matrix Completion to SDP

Suppose we have a low rank matrix \mathbf{M} . We have a set of location Ω describing our sampling. That is, if $(i, j) \in \Omega$, we observe entry M_{ij} . Given \mathbf{M} is low rank, it seems resonable that we would like to solve the following optimization problem

$$\begin{aligned} & \text{minimize} && \text{rank}(\mathbf{X}) \\ & \text{subject to} && X_{ij} = M_{ij} \quad (i, j) \in \Omega \\ & && \mathbf{X} \in \mathbb{R}^{n \times n} \end{aligned}$$

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But...

Rank is not a convex. This turns out to be an NP-Hard Problem.

Introduce the nuclear norm

Nuclear Norm

The nuclear norm is a close approximation of the rank.

The nuclear norm of a matrix \mathbf{X} is defined as the sum of the eigenvalues.

$$\|\mathbf{X}\|_* = \sum_{k=1}^n \sigma_k \mathbf{X}$$

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For a symmetric positive semi-definite (SPSD) matrices, the nuclear norm is equal to the trace.

A better relaxation

What if our matrix is not SPSD

- We introduce two matrices \mathbf{W}_1 and \mathbf{W}_2

A better relaxation

minimize $\text{trace}(\mathbf{W}_1) + \text{trace}(\mathbf{W}_2)$

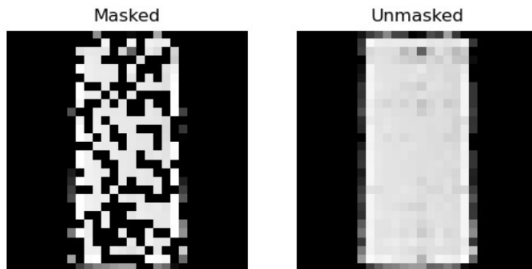
subject to $X_{ij} = M_{ij} \quad (i, j) \in \Omega$

$$\begin{bmatrix} \mathbf{W}_1 & \mathbf{X} \\ \mathbf{X}^\top & \mathbf{W}_2 \end{bmatrix} \succeq 0$$

Masked

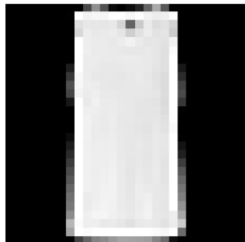


Fashion-MNIST

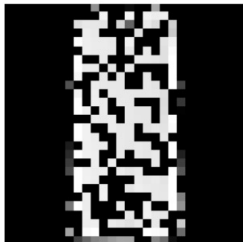


Fashion-MNIST

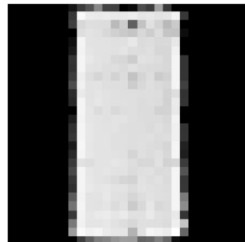
Original (rank=14)



Masked



Unmasked



Masked



Fashion-MNIST

Masked



Unmasked



Fashion-MNIST

Original (rank=20)



Masked



Unmasked



Lists

Bullet Points and Numbered Lists

- Lorem ipsum dolor sit amet, consectetur adipiscing elit
 - Aliquam blandit faucibus nisi, sit amet dapibus enim tempus
 - Lorem ipsum dolor sit amet, consectetur adipiscing elit
 - Nam cursus est eget velit posuere pellentesque
 - Nulla commodo, erat quis gravida posuere, elit lacus lobortis est, quis porttitor odio mauris at libero
-
- 1 Nam cursus est eget velit posuere pellentesque
 - 2 Vestibulum faucibus velit a augue condimentum quis convallis nulla gravida

Citing References

An example of the `\cite` command to cite within the presentation:

This statement requires citation [Smith, 2022, Kennedy, 2023].

References



John Smith (2022)

Publication title

Journal Name 12(3), 45 – 678.



Annabelle Kennedy (2023)

Publication title

Journal Name 12(3), 45 – 678.

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