

## Summary of Results

Two relative errors are evaluated as the metric for comparing various algorithms. They are defined as

$$Error_1 = \frac{\|\mathcal{P}_\Omega(\hat{M} - M^*)\|_F}{\|\mathcal{P}_\Omega(M^*)\|_F},$$

$$Error_2 = \frac{\|\hat{M} - M^*\|_F}{\|M^*\|_F}.$$

Let  $M^*$  be a 10-by-10 matrix of rank 5 with entries generated independently from the standard Normal distribution,  $M_{obs}$  has uniformly selected 20% missing entries. Table 1 summarizes the gap between the two errors as well as the time cost for each algorithm to reach such accuracy.

The hyperparameters of each algorithm are listed below.

- Vanilla GD: learning rate = 0.04, # iteration = 10000;
- Regularized GD: learning rate = 0.04, # iteration = 10000,  $C_t = 5, C_d = 6500, C_1 = 0.4, C_2 = 1$ ;
- Projected GD: learning rate = 0.04, # iteration = 10000,  $c = 1, \mu = 1$ ;
- GD on Manifold: # iteration = 10000,
  - Optimization over  $S$ : learning rate = 1, # iteration = 50;
- AltMin: learning rate = 0.06, # iteration = 100,
  - Optimization over  $L$  and  $R$ : # iteration = 100.

From the table, it is showed that (i) Regularized GD and Projected GD need well-selected tuning parameters in order to perform well as stated in the theorems, (ii) GD on manifold is computationally expensive due to the SVD step and the optimization over  $S$ , (iii) there is a gap of accuracy between the observed matrix and the true matrix.

Table 1: Comparison of Algorithms

	Vanilla GD	Regularized GD	Projected GD	GD on Manifold	AltMin
Error1	$4.24 \cdot 10^{-5}$	$4.24 \cdot 10^{-5}$	$4.24 \cdot 10^{-5}$	$1.71 \cdot 10^{-5}$	$1.27 \cdot 10^{-5}$
Error2	0.0010	0.0010	0.0010	0.0004	0.0002
Time(s)	0.27	4.79	7.81	68.72	0.27

However, when the missing fraction increases to 30%, none of the algorithms is able to recover the true matrix (Table 2). Note that

$$\frac{\|M_{obs} - M^*\|_F}{\|M^*\|_F} = 0.57.$$

Table 2: Results of Algorithms with missing fraction 30%

	Vanilla GD	Regularized GD	Projected GD	GD on Manifold	AltMin
Error1	0.00041	0.00041	0.00041	0.00840	0.00236
Error2	0.77	0.77	0.77	0.85	0.74
Time(s)	0.27	4.85	8.24	68.72	0.25

Moreover, Figure 1 shows that as iteration number grows, the predictor is closer and closer to  $M_{obs}$  but further and further away from  $M^*$ .

Figure 1: Change of error1 using Vanilla GD

