

Comparison of variational Bayes and Gibbs sampling in reconstruction of missing values with probabilistic principal component analysis

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Collaborative filtering

- Consider a table of ratings of movies (m_1, m_2, \dots) given by different people (p_1, p_2, \dots)
- The task is to make personalized recommendations
- This can be seen as reconstructing missing ratings in the table given the observed ones

	p1	p2	p3	p4	p5	p6
m1	4	5	?	5	2	
m2	5	?		5		3
m3		1	3		5	?

Principal Component Analysis (PCA)

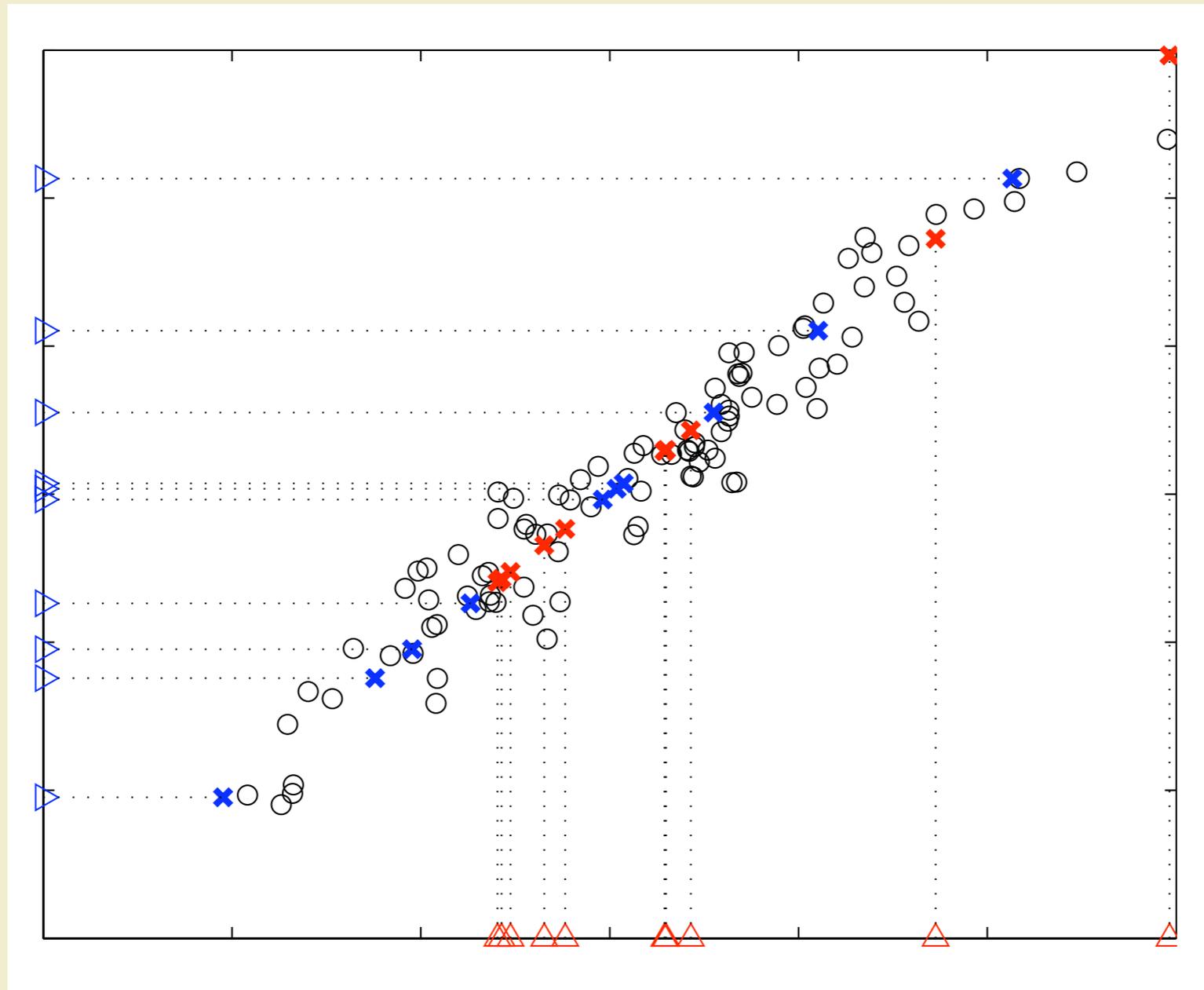
- Data Y consists of n d -dimensional vectors
- Matrix Y is decomposed into a product of smaller matrices such that the square reconstruction error is minimized

$$\mathbf{Y} \approx \mathbf{WX} + \mathbf{m}$$

Algorithms for PCA

- Eigenvalue decomposition (standard approach)
 - Compute the covariance matrix and its eigenvectors
- EM algorithm
 - Iterates between updates of W and X

PCA with missing values

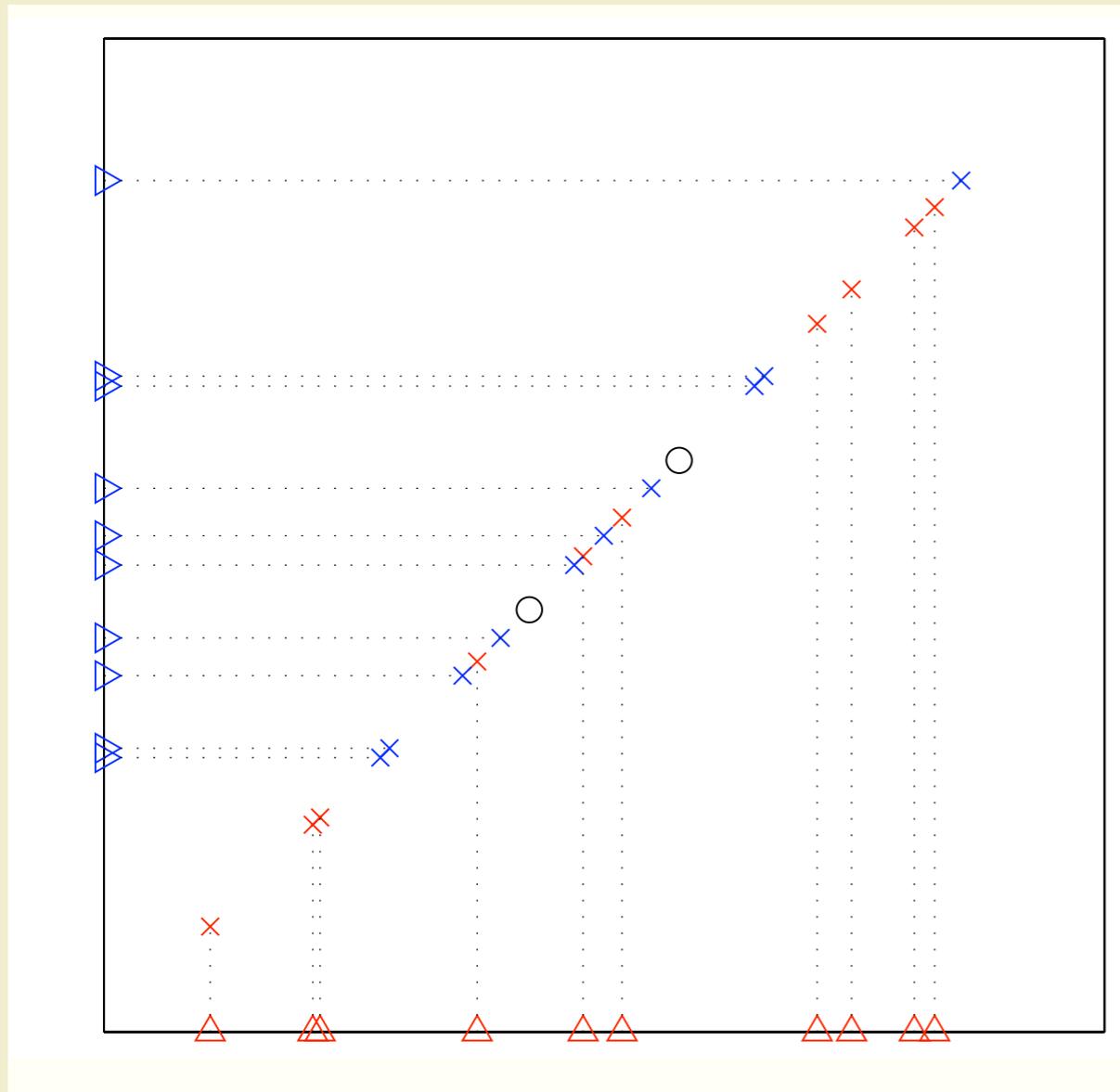


- Red and blue data points are reconstructed based on only one of the two dimensions

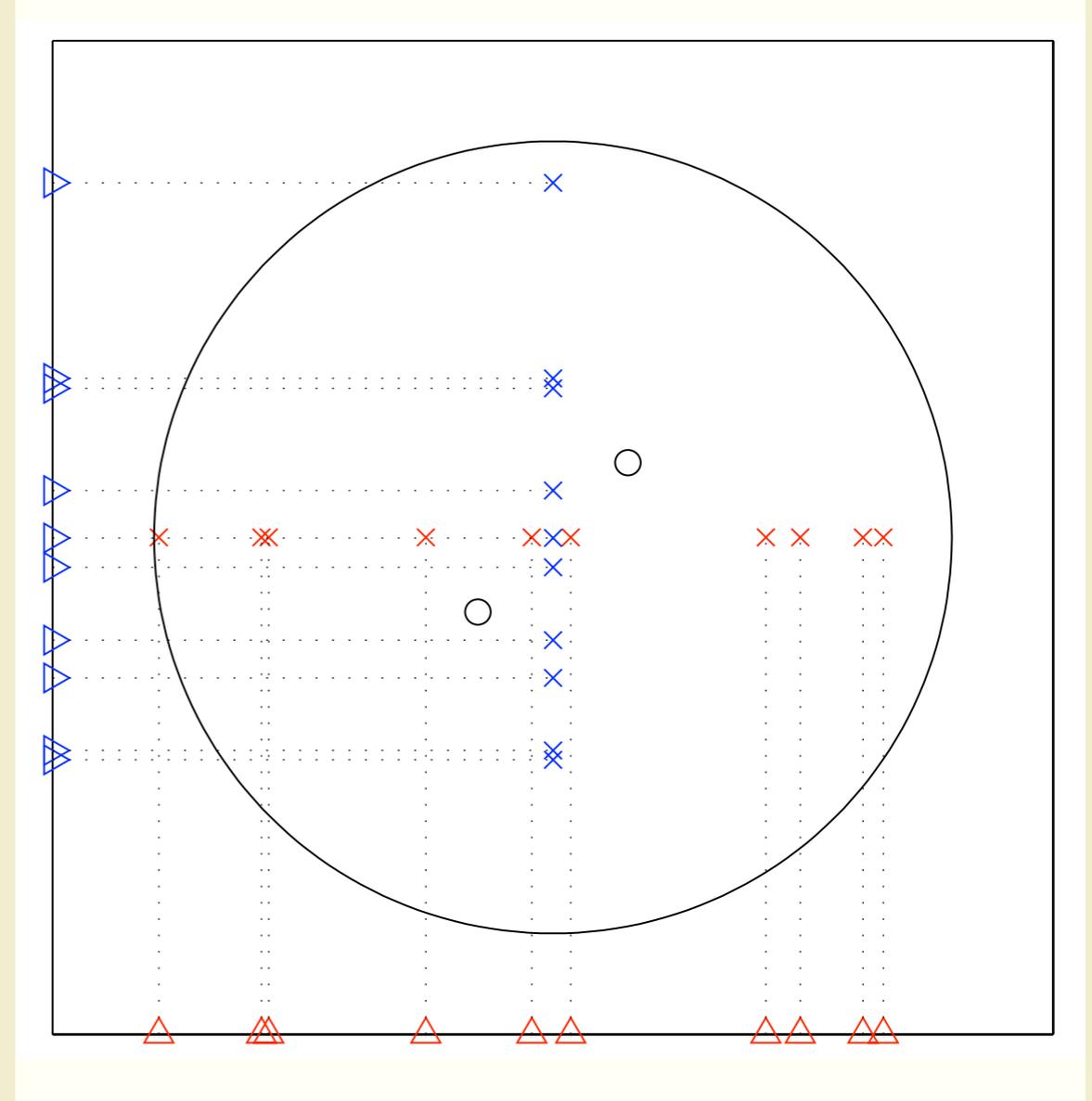
Adapting the algorithms for missing values

- Iterative imputation
 - Alternately 1) fill in missing values and 2) solve normal PCA with the standard approach
- EM algorithm becomes a bit more involved
 - Can be extended, and was thus used here

Overfitting in case of sparse data

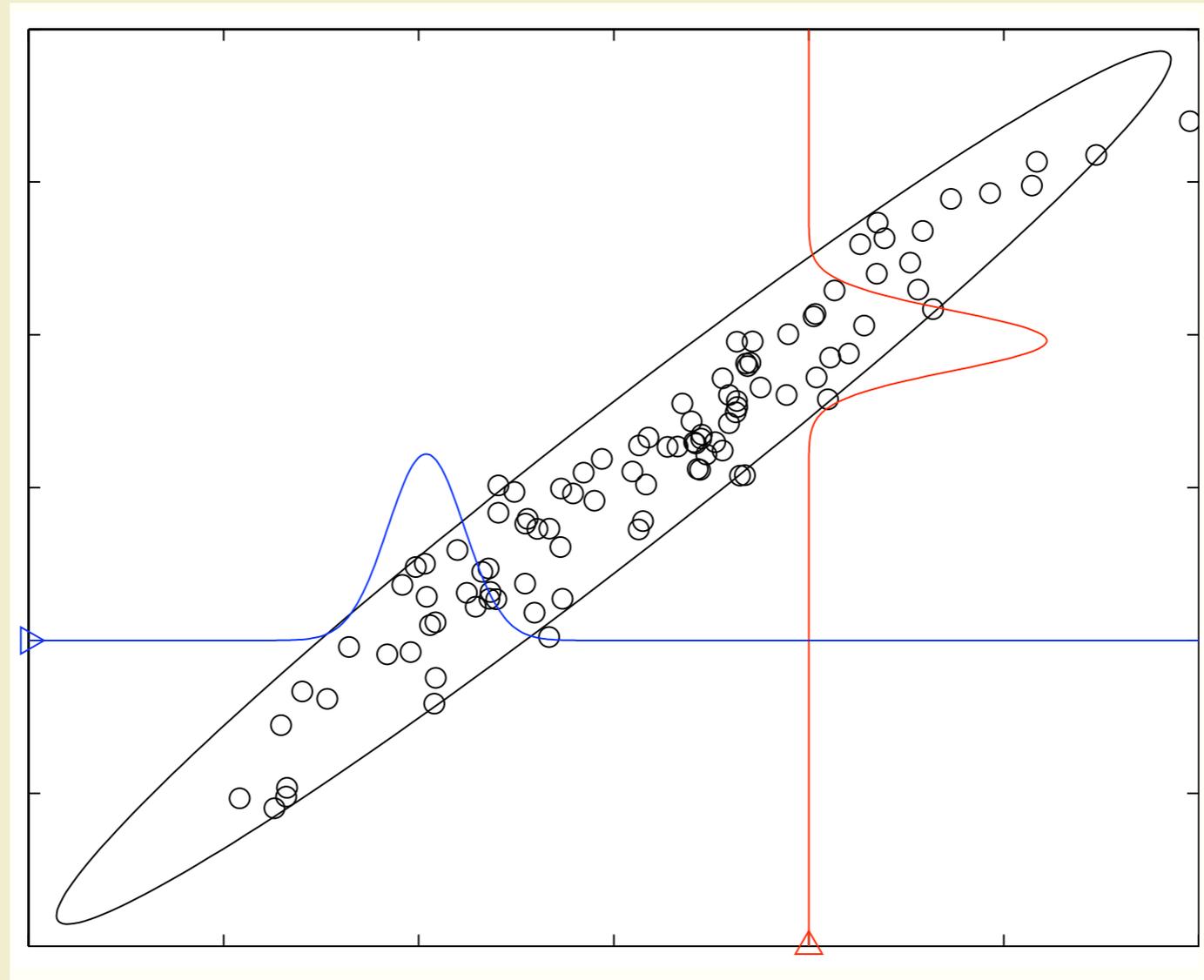


Overfitted solution



Regularized solution

Regularization against Overfitting



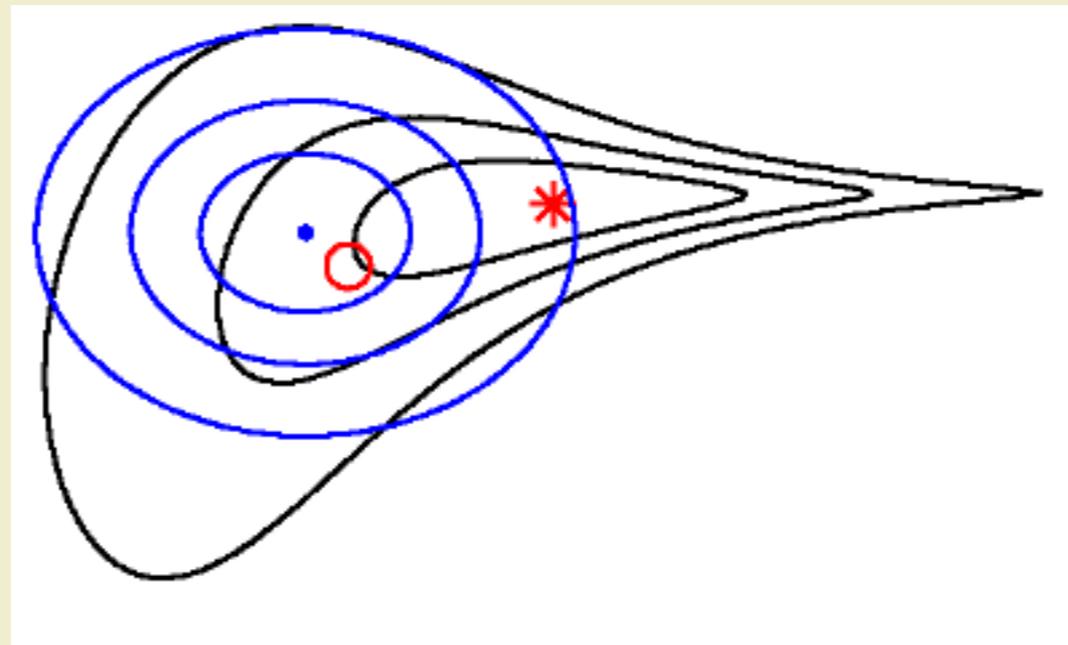
- Penalizing the use of large parameter values
- Estimating the distribution of unknown parameters

Modelling the uncertainty

$$P(\mathbf{Y}_{MIS} | \mathbf{Y}_{OBS}) = \int P(\mathbf{Y}_{MIS} | \mathbf{W}, \mathbf{X}, \mathbf{m}) P(\mathbf{W}, \mathbf{X}, \mathbf{m} | \mathbf{Y}_{OBS}) d\mathbf{W} d\mathbf{X} d\mathbf{m}.$$

- The optimal reconstruction is an integral over the unknown variables of the model
- There is no analytical solution to the integral so approximative methods must be used
- **Variational Bayes** and **Gibbs sampling** are such methods

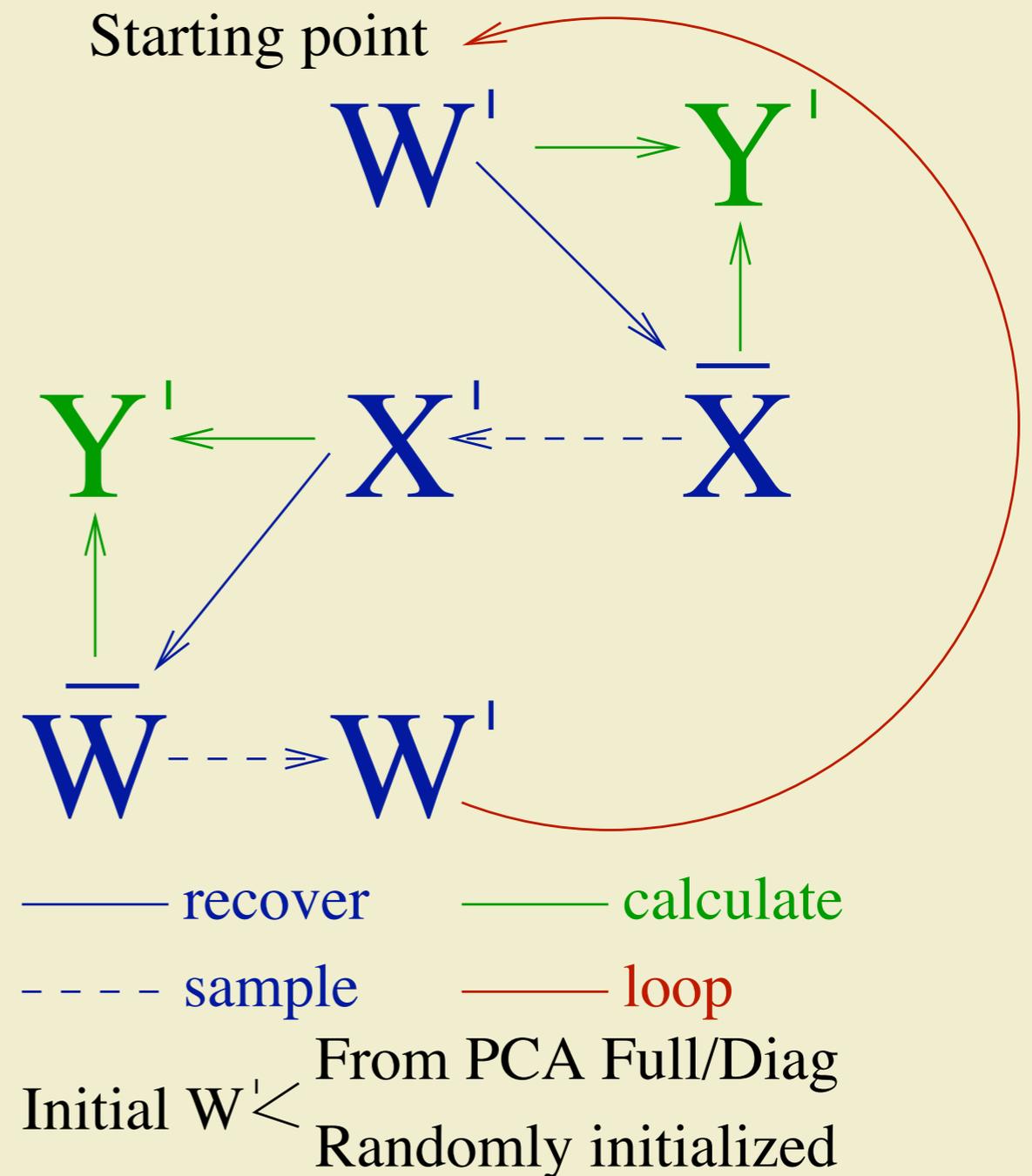
Variational Bayes



- **Distribution with a fixed form** is fitted to the **true distribution**
- The misfit between the two distributions is minimized iteratively

Gibbs sampling

- The distribution of W and m given Y and X has a simple form
- The distribution of X given Y , W and m has a simple form
- Draw random samples alternately from these

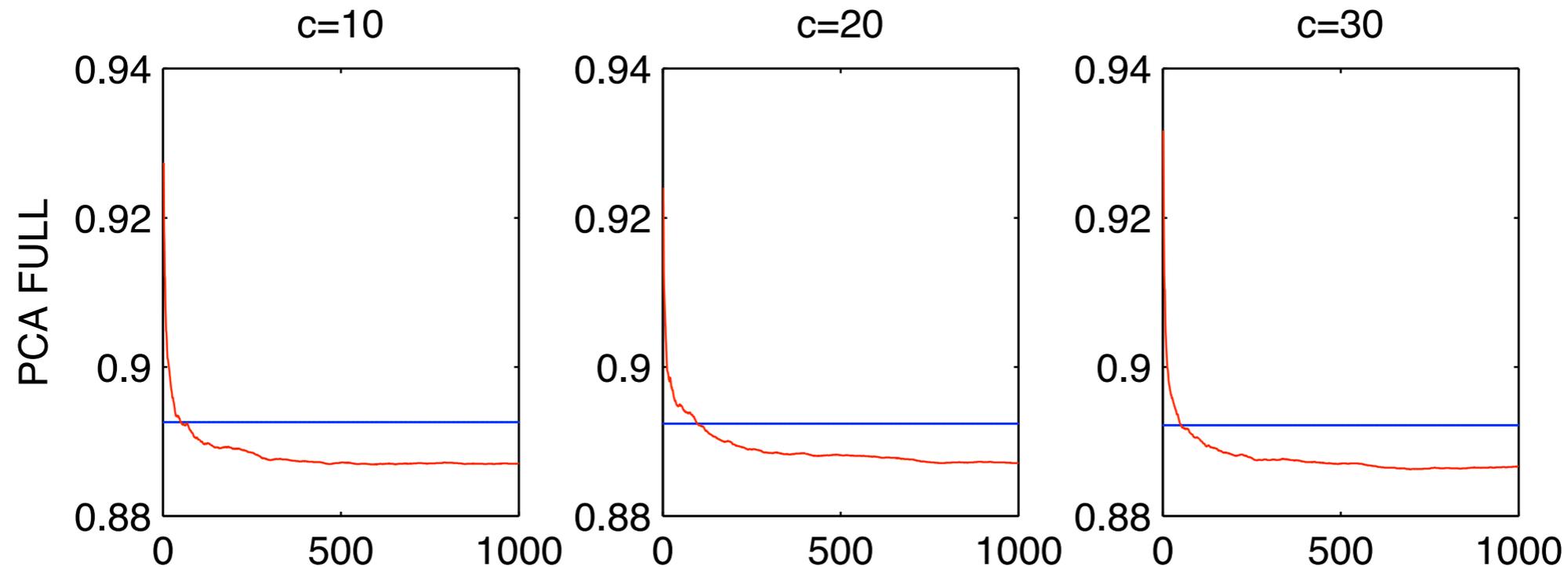


MovieLens data

- 100000 ratings from 1-5 given by 943 people to 1682 movies
- We used 95000 ratings for training and 5000 ratings for testing
- 94% missing values

Results with MovieLens data

- **Sampling** against **variational Bayes** with different number of components $c=10,20,30$
- RMS reconstruction error of test ratings



Conclusions

- When the data becomes very sparse, modelling uncertainty in PCA becomes important
- Sampling required about 100 samples to surpass variational Bayesian reconstructions