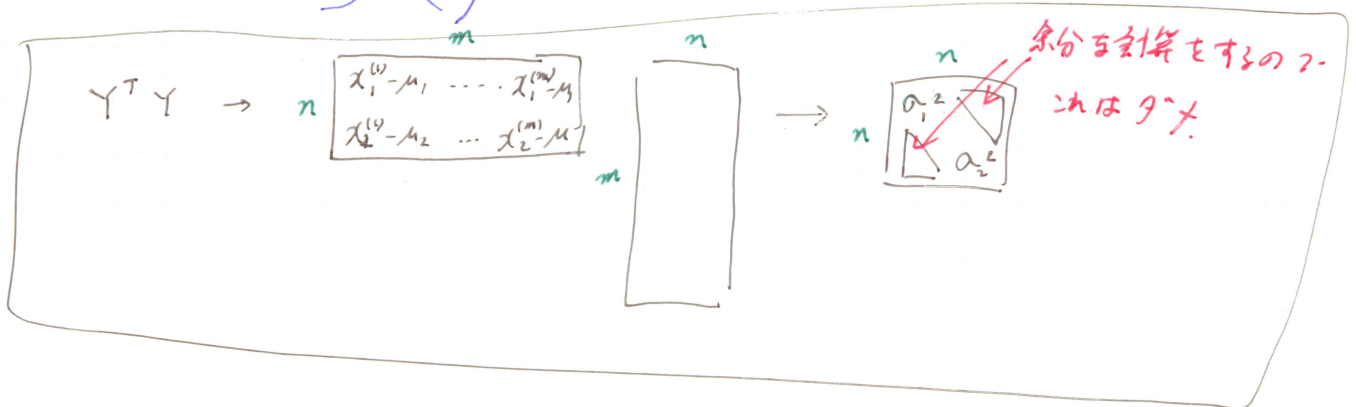
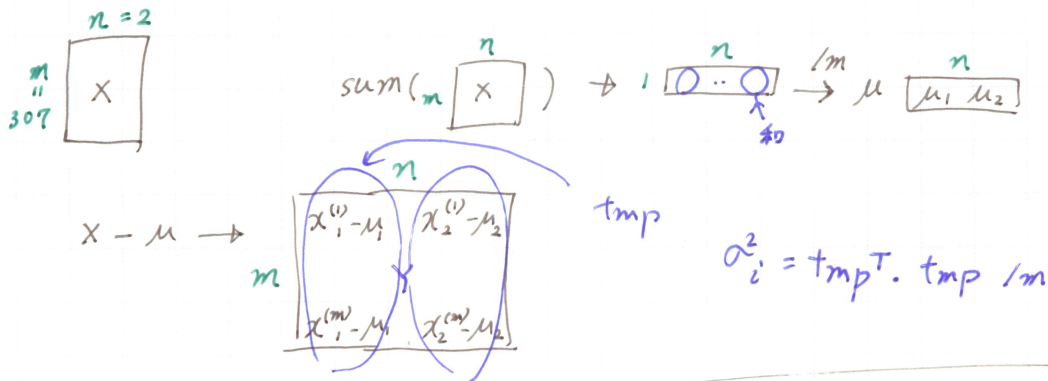


1. 1-1の制作 feature: through-put, latency,  $m=307$   
 最初は2次元で。後は many dimensions へ。

ガウス分布 (正規分布)  $P(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$

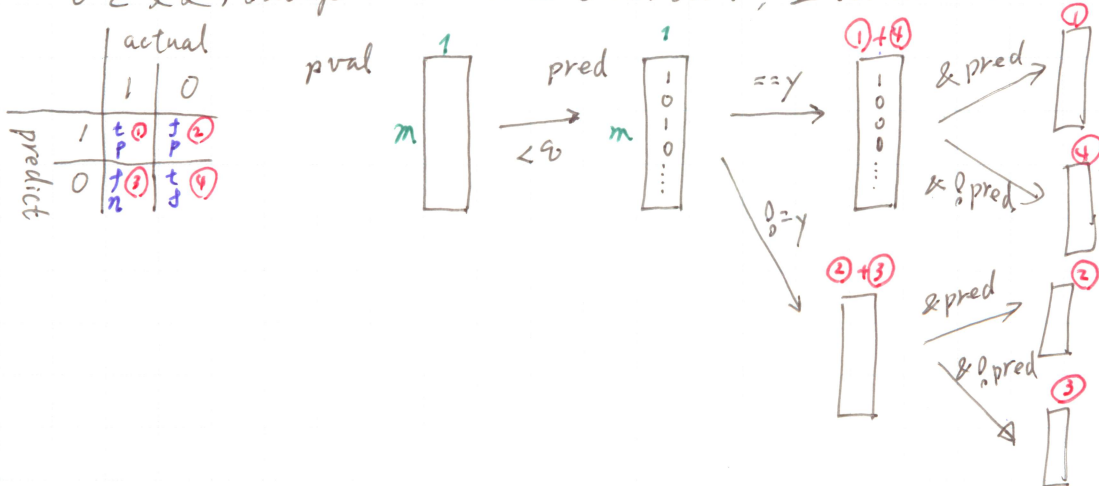
1.2 estimateGaussian.m



1.3 selectThreshold.m

$\epsilon$  と, F1 score を return する  
 $P(x^{(i)}) < \epsilon$  とすれば "anomaly"

$\epsilon$  を変更する loop は既に記述してあるので, 全ての cost を試す



1.4 High dimensional dataset

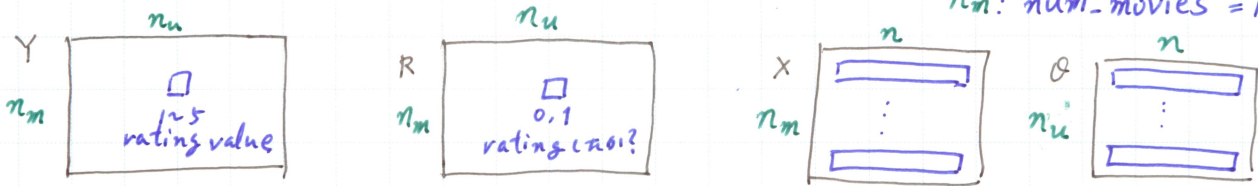
11 features

2. Recommender System

rating = 1 ~ 5,  $n_u = 943$ ,  $n_m = 1682$

$n$ : feature の数 = 100  
 $n_u$ : num-users = 943  
 $n_m$ : num-movies = 1682

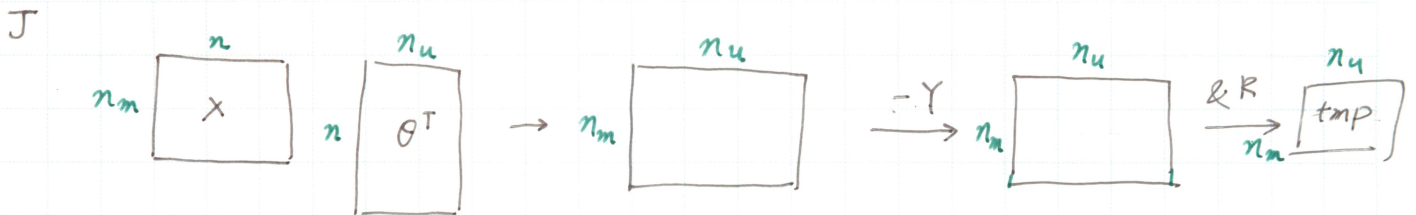
2.1 ex8-cofi.m  $\rightarrow$



2.2 regularization & L cost function

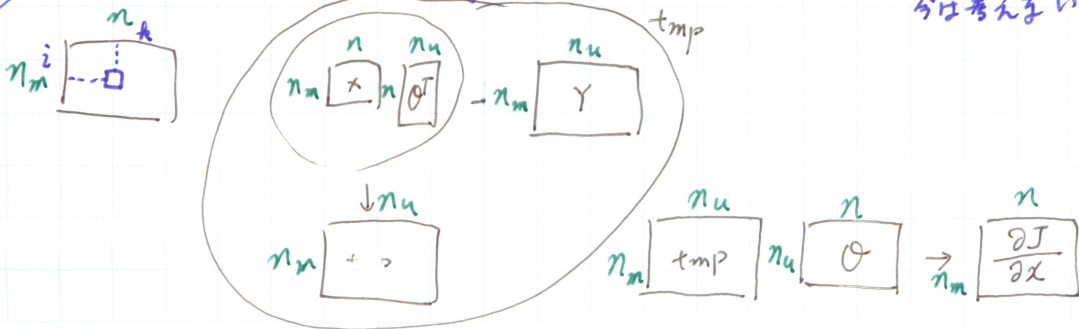
$y^{(i,j)} = (\theta^{(j)})^T x^{(i)}$  movie  $i$  に対する user  $j$  の rating を predict

cofi Cost Func. m



$\text{sum}(\text{sum}(n_m \text{ tmp} \cdot n_m \text{ tmp}^T)) \rightarrow \text{sum}(\text{sum}(n_m \text{ tmp})) \rightarrow \text{sum}(1) \rightarrow 1$

$\frac{\partial J}{\partial x_k} = x_k^{(i)} - \alpha \left( \sum_j ((\theta^{(j)})^T x^{(i)} - y^{(i,j)}) \theta_k^{(j)} + \lambda x_k^{(i)} \right)$   
 今は考えない



$$\frac{\partial J}{\partial \theta} \quad \theta_k^{(j)} := \theta_k^{(j)} - \alpha \left( \sum_i ((\theta^{(i)})^T x^{(i)} - y^{(i,j)}) x_k^{(i)} + \lambda \theta_k^{(j)} \right)$$

今は考えない

