

## MULTI-CLASS LOGISTIC REGRESSION

CONSIDER A MULTICLASS LOGISTIC REGRESSION PROBLEM OF THE FORM

$$(*) \left[ \min_{X \in \mathbb{R}^{d \times K}} \sum_{i=1}^m \left[ -x_{b_i}^T a_i + \log \left( \sum_{c=1}^K \exp(x_c^T a_i) \right) \right] \right]$$

LIKELIHOOD FOR SINGLE TRAINING EXAMPLE  $i$  WITH FEATURES  $a_i \in \mathbb{R}^d$  AND LABEL  $b_i \in \{1, 2, \dots, K\}$  IS GIVEN BY

$$P(b_i | a_i, X) = \frac{\exp(x_{b_i}^T a_i)}{\sum_{c=1}^K \exp(x_c^T a_i)}$$

PARAMETERS  
↓  
 $d \times K$

WHERE  $x_c$  IS COLUMN  $c$  OF MATRIX PARAMETER  $X \in \mathbb{R}^{d \times K}$

TO MAXIMIZE LIKELIHOOD OVER  $m$  INDEPENDENT IDENTICALLY-DISTRIBUTED TRAINING SAMPLES, WE MINIMIZE NEGATIVE LOG-LIKELIHOOD:

$$\hat{f}(X) = \sum_{i=1}^m \left[ -x_{b_i}^T a_i + \log \left( \sum_{c=1}^K \exp(x_c^T a_i) \right) \right]$$

### FREE TIP

PARTIAL DERIVATIVE IS:

$$\frac{\partial f(X)}{\partial x_{jc}} = - \sum_{i=1}^m a_{ij} \left[ \overset{\text{INDICATOR VARIABLE}}{\uparrow} I(b_i = c) - \frac{\exp(x_c^T a_i)}{\sum_{c'=1}^K \exp(x_{c'}^T a_i)} \right]$$

## HOMEWORK (DEADLINE 21ST OF MAY)

1. RANDOMLY GENERATE A  $1000 \times 1000$  MATRIX WITH ENTRIES FROM A  $N(0, 1)$  DISTRIBUTION
2. GEN.  $b_i \in \{1, 2, \dots, K\}$  ( $K=50$ ) BY DRAWING

$$AX + E$$

WITH  $X, E$  SAMPLED FROM NORMAL DISTRIBUTION  
 $X \in \mathbb{R}^{d \times K}$        $E \in \mathbb{R}^{m \times K}$

CONSIDER MAX INDEX IN THE ROW AS CLASS LABEL!

3. SOLVE PROBLEM ~~\*~~ WITH

(A) GRADIENT DESCENT

(B) BCGD WITH RANDOMIZED ROWS

(C) BCGD WITH GAUSS-SOUTHWELL RULE

USE BLOCKS  $X_{j:c'}$   $c' \in \{1, \dots, K\}$

!!  
[EACH ROW OF  $X$  IS ONE BLOCK!!!]

4. CHOOSE A PUBLICLY AVAILABLE DATASET AND TEST METHODS ON THIS.

5. ANALYSE ACCURACY VS CPU TIME

6. DESCRIBE WHAT YOU DID ON A PDF FILE

7. SUBMIT PROJECT