

COL774 Minor solutions

Q.1. (a) The general form of the degree k classifier
can be written as:- (assume binary class labels):

$$\sum_{t_1, \dots, t_n} \prod_{j=1}^n x_j^{t_j} = 0$$

t_1, \dots, t_n
s.t. $t_1 + t_2 + \dots + t_n \leq k$

Note:- this is the most general form of a polynomial
with degree k (each term (monomial) in the

expression is degree k or less and all such possible terms are constructed

$$(b) \quad \phi(x) = \sum_{k=0}^k a_k x^k \equiv \sum_{k=0}^k a_k (x_0, x_1, \dots, x_n)^k$$

which will generate all the monomial terms of degree k or less features.

It is given that $\phi(x)^T \phi(z) = K(x, z)$

$$= (x^T z + C)^d$$

missing in
but question
not really
wup for solution

Now, let us look at the w parameter update equation:-

[Note:- entire algorithm is now run in the transformed feature space i.e. represented by $\phi(x)$]

Training Data:-

$$\{x^{(i)}, y^{(i)}\}_{i=1}^m$$

$$J(\theta) = \frac{1}{2} \sum_{i=1}^m (y^{(i)} - h_{\theta}(\phi(x^{(i)})))^2$$

in transformed space

$$\theta^{(0)} \leftarrow \vec{0} \quad [\text{initialization to 0 vector}]$$

$$t = 0$$

$$\text{do } \theta^{(t+1)} \leftarrow \theta^{(t)} - \eta \frac{1}{m} \sum_{i=1}^m (y^{(i)} - h_{\theta}(\phi(x^{(i)}))) \phi(x^{(i)})$$

$$t \leftarrow t+1$$

3 (until convergence);

[replaced occurrence of $x^{(i)}$ by $\phi(x^{(i)})$]

Now:-
$$h_{\theta}(\phi(x^{(i)})) = \frac{1}{1 + e^{-\theta^T \cdot \phi(x^{(i)})}}$$

We will assume that θ can be represented as coefficients $\alpha_1 \dots \alpha_m$ of data points $x^{(1)} \dots x^{(m)}$. In particular:-

$$\theta = \sum_{i=1}^m \alpha_i \phi(x^{(i)})$$

At time t ,
$$\theta(t) = \sum_{i=1}^m \alpha_i(t) \phi(x^{(i)})$$

We will show that this assumption holds inductively

At $t=0$, this assumption is true. Since,
 at $t=0$, $Q=0 \Rightarrow \alpha_i t = 0 \forall i$.
 Next assume it holds for ~~the~~ step t .
 Then let us see if we can prove for
 step $t+1$.

$$Q(t+1) \leftarrow Q(t) - \eta \cdot \frac{1}{n} \sum_{i=1}^n \left[y^{(i)} - h_{Q(t)}(x^{(i)}) \right]$$

$$\text{Then: } h_{Q(t)}(x^{(i)}) = \frac{1}{1 + e^{-Q(t)^T \cdot \phi(x^{(i)})}}$$

Now, by inductive assumption, we have $Q(t) = \sum_{i=1}^m \alpha_i^t \phi(x^{(i)})$

$$\Rightarrow \theta(t)^T \cdot \phi(x^{(i)}) = \left[\sum_{l=1}^m d_l e^t \phi(x^{(l)}) \right]^T \cdot \phi(x^{(i)})$$

→ change of variable from l to i

Computation only in terms of $\phi(x^{(l)})$

terms. $\phi(x^{(i)})$ involvement

No other $\phi(x^{(i)})$ in the expression

$$= \sum_{l=1}^m d_l e^t [\phi(x^{(l)})^T \phi(x^{(i)})]$$

$$= \sum_{l=1}^m d_l e^t K(x^{(l)}, x^{(i)}) = \beta_i^t$$

$$\theta(t+1) \leftarrow \theta(t) - \eta \cdot \frac{1}{m} \sum_{l=1}^m \left[y^{(l)} - \frac{1}{1 + e^{-\beta_i^t}} \right]$$

β_i^t \swarrow $\phi(x^{(i)})$

$$\Rightarrow \theta^{(t+1)} \leftarrow \theta^{(t)} - \eta \cdot \frac{1}{m} \sum_{i=1}^m \gamma_i^t \phi(x^{(i)})$$

$$\Rightarrow \theta^{(t+1)} \leftarrow \theta^{(t)} - \sum_{i=1}^m \gamma_i^t \cdot \phi(x^{(i)})$$

$$[\gamma_i^t = \eta/m \gamma_i^t]$$

\Rightarrow Using inductive assumption

$$\theta^{(t)} = \sum_{i=1}^m \alpha_i^t \phi(x^{(i)})$$

$$\Rightarrow \theta^{(t+1)} = \sum_{i=1}^m \alpha_i^t \phi(x^{(i)}) - \sum_{i=1}^m \gamma_i^t \phi(x^{(i)})$$

$$\Rightarrow \theta^{(t+1)} = \sum_{i=1}^m \underbrace{(d_i^t - \gamma_i^t)}_{\text{Define } d_i^{t+1}} \phi(x^{(i)})$$

$$\Rightarrow \theta^{(t+1)} = \sum_{i=1}^m d_i^{(t+1)} \phi(x^{(i)})$$

\Rightarrow Hence, inductive assumption is satisfied.
 Further, \nexists entire computation can be done in
 form of $[\phi(x)^T \phi(z)]$ expressions
 only. since we did not use
 $\phi(x^{(i)})$ in any other form

on above computation.

(C) Computational complexity:

$\phi(x)$ will have exponential (in k) number of terms. ϕ features [All possible monomials in x_1, \dots, x_n of degree $\leq k$]

Therefore, if we directly represent $\phi(x)$ it will consume exponential in k time/memory.

But if we only need to compute

$$K(x, z) = \phi(x)^T \phi(z) = (x^T z + c)^d$$

this takes only $O(n+d)$ time. (if memory is $O(n)$)

