

# Homework #5 - Theory + SVM

204039453  
205714447

## 1. Kernels and mapping functions (25 pts)

- a. (20 pts) Let  $K(x, y) = (x \cdot y + 1)^3$  be a function over  $\mathbb{R}^2 \times \mathbb{R}^2$  (i.e.,  $x, y \in \mathbb{R}^2$ ).

Find  $\psi$  for which  $K$  is a kernel. (It may help to first expand the above term on the right-hand side).

- b. (2 pts) What did we call the function  $\psi$  in class if we remove all coefficients?

- c. (3 pts) How many multiplication operations do we save by using  $K(x, y)$

versus  $\psi(x) \cdot \psi(y)$ ?

$$*(a+b)^3 = a^3 + 3a^2b + 3ab^2 + b^3$$

$$\checkmark x \cdot y = (x_1y_1 + x_2y_2)$$

$$a. (x \cdot y + 1)^3 = (x \cdot y)^3 + 3(x \cdot y)^2 + 3x \cdot y + 1$$

$$* (x_1y_1 + x_2y_2)^3 = (x_1^3y_1^3 + 3x_1^2y_1^2x_2y_2 + 3x_1y_1x_2^2y_2^2 + x_2^3y_2^3)$$

$$* 3(x_1y_1 + x_2y_2)^2 = 3[x_1^2y_1^2 + 2x_1y_1x_2y_2 + x_2^2y_2^2]$$

$$* 3(x_1y_1 + x_2y_2)$$

↓ plug in the equation:

$$= (x_1^3y_1^3 + 3x_1^2y_1^2x_2y_2 + 3x_1y_1x_2^2y_2^2 + x_2^3y_2^3) + 3[x_1^2y_1^2 + 2x_1y_1x_2y_2 + x_2^2y_2^2] + 3(x_1y_1 + x_2y_2) + 1$$

$$= (x_1^3, 3x_1^2x_2, 3x_1x_2^2, x_2^3, 3x_1^2, 6x_1x_2, 3x_2^2, 3x_1, 3x_2, 1) \cdot$$

$$(y_1^3, 3y_1^2y_2, 3y_1y_2^2, y_2^3, 3y_1^2, 6y_1y_2, 3y_2^2, 3y_1, 3y_2, 1)$$

$$\text{therefore: } \varphi(x) = (x_1^3, 3x_1^2x_2, 3x_1x_2^2, x_2^3, 3x_1^2, 6x_1x_2, 3x_2^2, 3x_1, 3x_2, 1)$$

- b. Removing all the coefficients will generate a full rational variety of order 3

A dimension 2.

- c. # of multiplication operation for  $\varphi(x) \cdot \varphi(y) = 10$  (10 indices,  $\binom{5}{3} = 10$ )

# of multiplication operation for  $K(x, y) = 4$

therefore with the kernel we saved  $(10 - 4) = 6$  multiplication operation

## 2. Lagrange multipliers (25 pts)

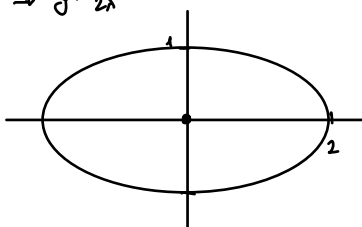
Let  $f(x, y) = 2x - y$ . Find the minimum and the maximum points for  $f$  under the constraint  $g(x, y) = \frac{x^2}{4} + y^2 = 1$ .

2)  $\mathcal{L}(x, y, \lambda) = 2x - y + \lambda(\frac{x^2}{4} + y^2 - 1)$

\*  $\frac{\partial \mathcal{L}}{\partial x} = 2 + \lambda \cdot \frac{2x}{4} = 0 \rightarrow 2 + \frac{1}{2}\lambda x = 0 \quad / : 2 \rightarrow 1 + \frac{1}{4}\lambda x = 0 \rightarrow \lambda x = -4 \quad / : \lambda, \lambda \neq 0 \rightarrow x = -\frac{4}{\lambda}$

\*  $\frac{\partial \mathcal{L}}{\partial y} = -1 + \lambda \cdot 2y = 0 \quad / : 2 \rightarrow -\frac{1}{2} + \lambda y = 0 \rightarrow \lambda y = \frac{1}{2} \quad / : \lambda, \lambda \neq 0 \rightarrow y = \frac{1}{2\lambda}$

\*  $\frac{\partial \mathcal{L}}{\partial \lambda} = g(x, y) = \frac{x^2}{4} + y^2 - 1 = 0$  (ellipse as drawn:  $\rightarrow$



$x = -\frac{4}{\lambda}, y = \frac{1}{2\lambda}, \lambda \neq 0: \quad \frac{x^2}{4} + y^2 - 1 = 0$

$\frac{(-\frac{4}{\lambda})^2}{4} + (\frac{1}{2\lambda})^2 - 1 = 0$

$\frac{16}{\lambda^2} + \frac{1}{4\lambda^2} - 1 = 0$

$\frac{16}{\lambda^2} + \frac{1}{4\lambda^2} - 1 = 0 \quad / \cdot 4\lambda^2, \lambda \neq 0$

$16 + 1 - 4\lambda^2 = 0 \rightarrow 4\lambda^2 = 17 \rightarrow \lambda^2 = \frac{17}{4} \rightarrow \lambda = \pm \frac{\sqrt{17}}{2}$

#1

$\lambda_1 = \frac{\sqrt{17}}{2}: \quad x_1 = -\frac{4}{\lambda} \rightarrow -\frac{4}{\frac{\sqrt{17}}{2}} \rightarrow -\frac{8}{\sqrt{17}} \rightarrow x_1 = -\frac{8}{\sqrt{17}}$

$y_1 = \frac{1}{2\lambda} \rightarrow \frac{1}{2 \cdot \frac{\sqrt{17}}{2}} \rightarrow y_1 = \frac{1}{\sqrt{17}}$

#2

$\lambda_2 = -\frac{\sqrt{17}}{2}: \quad x_2 = -\frac{4}{\lambda} \rightarrow -\frac{4}{-\frac{\sqrt{17}}{2}} \rightarrow \frac{8}{\sqrt{17}} \rightarrow x_2 = \frac{8}{\sqrt{17}}$

$y_2 = \frac{1}{2\lambda} \rightarrow \frac{1}{2 \cdot (-\frac{\sqrt{17}}{2})} \rightarrow y_2 = -\frac{1}{\sqrt{17}}$

$$f(x_1, y_1) = 2 \cdot \left(-\frac{8}{\sqrt{17}}\right) - \frac{1}{\sqrt{17}} = \frac{-17}{\sqrt{17}} = -\sqrt{17}$$

$$f(x_2, y_2) = 2 \cdot \left(\frac{8}{\sqrt{17}}\right) - \left(-\frac{1}{\sqrt{17}}\right) = \frac{17}{\sqrt{17}} = \sqrt{17}$$

therefore:  $f(x_1, y_1) < f(x_2, y_2)$

↓

$(x_1, y_1) = \left(-\frac{8}{\sqrt{17}}, \frac{1}{\sqrt{17}}\right)$  is a minimum point for  $f$  under the constraint  $g(x, y) = x^2 + y^2 = 1$   
with  $\lambda = \frac{\sqrt{17}}{2}$

$(x_2, y_2) = \left(\frac{8}{\sqrt{17}}, -\frac{1}{\sqrt{17}}\right)$  is a maximum point for  $f$  under the constraint  $g(x, y) = x^2 + y^2 = 1$   
with  $\lambda = -\frac{\sqrt{17}}{2}$

### 3. PAC Learning (25 pts)

Let  $X = \mathbb{R}^2$ . Let vectors  $u = \left(\frac{\sqrt{3}}{2}, \frac{1}{2}\right)$ ,  $w = \left(\frac{\sqrt{3}}{2}, -\frac{1}{2}\right)$ ,  $v = (0, -1)$

$$\text{and } C = H = \left\{ h(r) = \left\{ (x_1, x_2) \mid \begin{cases} (x, y) \cdot u \leq r, \\ (x, y) \cdot v \leq r, \\ (x, y) \cdot w \leq r \end{cases} \right\}, \text{ for } r > 0, \right.$$

the set of all origin-centered upright equilateral triangles.

Describe a polynomial sample complexity algorithm  $L$  that learns  $C$  using  $H$ . State the time complexity and the sample complexity of your suggested algorithm. Prove all your steps.

from the piazza:  
 $w = \left(-\frac{\sqrt{3}}{2}, \frac{1}{2}\right)$

Algorithm  $L(\delta)$ :

$$D \in X^m, \forall d \in D \rightarrow d = (x_1, x_2), x_1, x_2 \in \mathbb{R}$$

algo:

$r = 0$  / initialize  $r$  with a number close to zero since  $r > 0$ .

for every  $d$  in  $D$ :

$$r_{\max} = \arg \max \{ d \cdot u, d \cdot v, d \cdot w \}$$

if  $r_{\max} > \delta$ :

$$r = r_{\max}$$

return  $h(r)$ ,  $h \in H$ .

algo Complexity  $O(m)$

Time Complexity: Since we iterate over  $m$  samples and make operations with  $O(1)$

therefore the total time complexity will be  $O(m)$  where  $m$  is polynomial in  $1/\epsilon$  &  $1/\delta$

(Continue to the sample complexity...)

Sample Complexity:  $\delta$ -Confidence level  
 $0 \leq \delta \leq 1$

$$p(\text{err} > \epsilon) \leq \delta$$

$$p(\{D \in X^m : \text{Err}(L(D), c) > \epsilon\}) \leq$$

$$p(X-B)^m \leq (1-\epsilon)^m \leq e^{-m\epsilon}$$

$$e^{-m\epsilon} \leq \delta$$

$$e^{-m\epsilon} \leq \delta$$

$$-m\epsilon \leq \ln(\delta) / \cdot (-1)$$

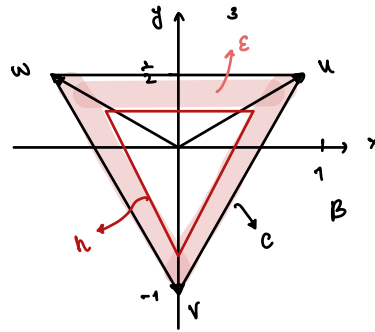
$$m\epsilon \geq -\ln(\delta) / \frac{1}{\epsilon}$$

$$m \geq \frac{1}{\epsilon} \cdot [-\ln(\delta)]$$

For example, with certainty of  $(1-\delta)$  and err of  $(\epsilon)$

$$m \geq \frac{1}{0.1} [-\ln(0.05)] \rightarrow m \geq 30$$

therefore the sample size needs to be at least 30.



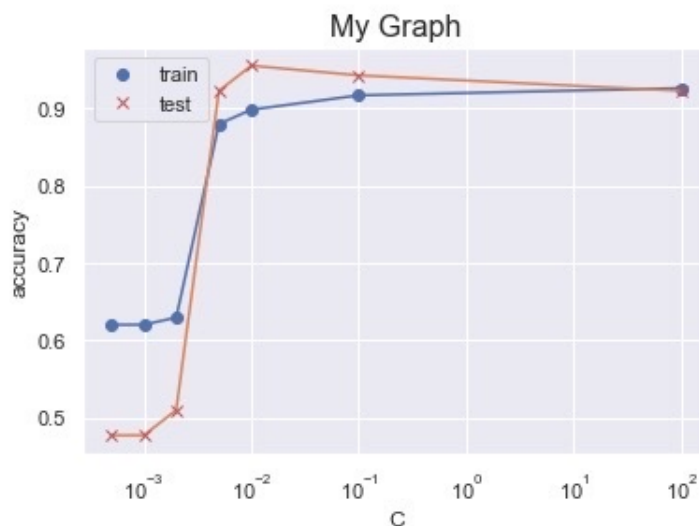
4. (15 pts) A business manager at your ecommerce company asked you to make a model to predict whether a user is going to proceed to checkout or abandon their cart. You created the model using, and reported 20% error on your test set of size 1000 samples. In the business manager's presentation to upper management, he presented your model and stated that the company can expect 20% error when deploying the model live on the website.
- Luckily, you realize that this is a mistaken assumption, and you correct the statement to say that with 95% confidence, the true error they can expect is up to what percentage? (Just state the error percentage).

upper bound:  $\hat{p} + 2SE = \hat{p} + 2\sqrt{\frac{\hat{p}(1-\hat{p})}{n}} = 0.2 + 2\sqrt{\frac{0.2 \cdot 0.8}{1000}} = 0.2 + 2 \cdot 0.0126 = 0.225 = 22.5\%$

5. SVM (10 pts)

See the notebook in the homework files and follow the instructions there.

Take a **screenshot** of your resulting graph near the bottom of the notebook (titled "My Graph") and paste into your submission PDF along with your answers to the theoretical questions. Do **NOT** submit your code.



`Cs = [0.0005, 0.001, 0.002, 0.005, 0.01, 0.1, 100]`