

Machine Learning*

Homework L^AT_EX

* Teacher: ... TA: ...

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1. HOMEWORK I

A. [20pts] Basic Probability and Statistics

The probability distribution of random variable X follows:

$$f_X(x) = \begin{cases} \frac{1}{2} & 0 < x < 1; \\ \frac{1}{6} & 2 < x < 5; \\ 0 & \text{otherwise.} \end{cases} \quad (1)$$

(1) [5pts] Please give the cumulative distribution function $F_X(x)$ for X ;

(2) [5pts] Define random variable Y as $Y = 1/(X^2)$, please give the probability density function $f_Y(y)$ for Y ;

(3) [10pts] For some random non-negative random variable Z , please prove the following two formulations are equivalent:

$$\mathbb{E}[Z] = \int_{z=0}^{\infty} z f(z) dz, \quad (2)$$

$$\mathbb{E}[Z] = \int_{z=0}^{\infty} \Pr[Z \geq z] dz, \quad (3)$$

Meantime, please calculate the expectation of random variable X and Y by these two expectation formulations to verify your proof.

*Thank you teacher and TAs for your correction.

My solution:

(1):

$$F_X(x) = \begin{cases} 0 & x \leq 0; \\ \int_{-\infty}^x f_X(x) = \frac{x}{2} & 0 < x < 1; \\ \int_{-\infty}^x f_X(x) = \int_0^1 f_X(x) = \frac{1}{2} & 1 \leq x \leq 2 \\ \frac{1}{2} + \frac{x}{6} - \frac{1}{3} = \frac{x+1}{6} & 2 < x < 5; \\ 1 & x \geq 0 \end{cases}$$

(2):

Y 的可能取值范围: $(0, +\infty]$;

$y \leq 0$ 时,

$$F_Y(y) = P(Y \leq y) = 0;$$

$y > 0$ 时,

$$F_Y(y) = P(Y \leq y) = P\left(\frac{1}{y} \leq x^2\right) = 1 - F_X\left(\frac{1}{\sqrt{y}}\right) + F_X\left(-\frac{1}{\sqrt{y}}\right)$$

求导得:

$$p_Y(y) = F'_Y(y) = \begin{cases} \frac{1}{2y^{\frac{3}{2}}} \left(p_X\left(\frac{1}{\sqrt{y}}\right) - p_X\left(-\frac{1}{\sqrt{y}}\right) \right) & y > 0 \\ 0 & y \leq 0 \end{cases}$$

(3):

i. 由定义

$$\begin{aligned} \mathbb{E}[Z] &= \sum_{i=1}^{+\infty} z_i P(Z = z_i) \\ &= \lim_{\Delta z \rightarrow 0} \sum_{i=1}^{+\infty} z_i (F_Z(z_i + \Delta z) - F_Z(z_i)) \\ &= \lim_{\Delta z \rightarrow 0} \sum_{i=1}^{+\infty} z_i \left(\int_{z_i}^{z_i + \Delta z} f(z) dz \right) \\ &= \lim_{\Delta z \rightarrow 0} \sum_{i=1}^{+\infty} \Delta z z_i f(z_i) \\ &= \int_{-\infty}^{\infty} z f(z) dz \\ &= \int_{z=0}^{\infty} z f(z) dz \end{aligned}$$

ii.

$$\begin{aligned}
E(Z) &= \int_0^{+\infty} (1-F(z)) dz \\
0 \leq z F(-z) &\leq \int_{-\infty}^{-z} |x| dF(x) \quad (\forall z > 0) \\
\Rightarrow \lim_{z \rightarrow +\infty} z F(-z) &\rightarrow 0 \Rightarrow \lim_{z \rightarrow -\infty} z F(z) = 0 \\
\text{[2] 证} \quad 0 \leq z [1-F(z)] &\leq \int_0^{+\infty} x d(1-F(x)) \\
\Rightarrow \lim_{z \rightarrow +\infty} z [1-F(z)] &= 0 \\
E(Z) &= \int_0^{+\infty} z dF(z) \\
&= - \int_0^{+\infty} z d[1-F(z)] \\
&= -z(1-F(z)) \Big|_0^{+\infty} + \int_0^{+\infty} (1-F(z)) dz \\
&= \int_0^{+\infty} (1-F(z)) dz \\
&\text{得证.}
\end{aligned}$$

2. [20PTS] STRONG CONVEXITY

Let $D \in \mathbb{R}^2$ be a finite set. Define a function $E : \mathbb{R}^3 \rightarrow \mathbb{R}$ by

$$E(a, b, c) = \sum_{x \in D} (ax_1^2 + bx_1 + c - x_2)^2. \quad (4)$$

(1) [10pts] Show that E is convex.

(2) [10pts] Does there exist a set D such that E is strongly convex? Proof or a counterexample.

A. [20pts] Transition Probability Matrix

Suppose x_k is the fraction of NJU students who prefer course A at year k . The remaining fraction $y_k = 1 - x_k$ prefers course B.

At year $k+1$, $\frac{1}{5}$ of those who prefer course A change their mind. Also at the same year, $\frac{1}{10}$ of those who prefer course B change their mind (possibly after taking the problem 3 last year).

Create the matrix P to give $[x_{k+1} \ y_{k+1}]^\top = P[x_k \ y_k]^\top$ and find the limit of $P^k[1 \ 0]^\top$ as $k \rightarrow \infty$.

B. [20pts] Hypothesis Testing

Yesterday, a student was caught by the teacher when tossing a coin in class. The teacher is very nice and did not want to make things difficult. S(he) wished the student to determine *if the coin is biased for heads* with $\alpha = 0.05$.

Also, according to the student's desk mate, the coin was tossed for 50 times and it got 35 heads.

(1) [10pts] Show all calculate and rules (hint: using z-test).

(2) [10pts] Calculate the p-value and interpret it.

C. [20pts] Performance Measures

We have a set of samples that we wish to classify in one of two classes and a ground truth class of each sample (denoted as 0 and 1). For each example a classifier gives us a score (score closer to 0 means class 0, score closer to 1 means class 1). Below are the results of two classifiers (C_1 and C_2) for 8 samples, their ground truth values (y) and the score values for both classifiers (y_{C_1} and y_{C_2}).

y	1	0	1	1	1	0	0	0
y_{C_1}	0.5	0.3	0.6	0.22	0.4	0.51	0.2	0.33
y_{C_2}	0.04	0.1	0.68	0.22	0.4	0.11	0.8	0.53

(1) [8pts] For the example above calculate and draw the ROC curves for classifier C_1 and C_2 . Also calculate the area under the curve (AUC) for both classifiers.

(2) [8pts] For the classifier C_1 select a decision threshold $th_1 = 0.33$ which means that C_1 classifies a sample as class 1, if its score $y_{C_1} > th_1$, otherwise it classifies it as class 0. Use it to calculate the confusion matrix and the F_1 score. Do the same thing for the classifier C_2 using a threshold value $th_2 = 0.1$.

(3) [4pts] Prove Eq.(2.22) in Page 35. ($AUC = 1 - \ell_{rank}$).

D. [Bonus 10pts]Expected Prediction Error

For least squares linear regression problem, we assume our linear model as:

$$y = x^T \beta + \epsilon, \quad (5)$$

where ϵ is noise and follows $\epsilon \sim N(0, \sigma^2)$. Note the instance feature of training data \mathcal{D} as $\mathbf{X} \in \mathbb{R}^{p \times m}$ and note the label as $\mathbf{Y} \in \mathbb{R}^n$, where n is the number of instance and p is the feature dimension. So the estimation of model parameter is:

$$\hat{\beta} = (\mathbf{X}\mathbf{X}^T)^{-1}\mathbf{X}\mathbf{Y}. \quad (6)$$

For some given test instance x_0 , please proof the expected prediction error $\mathbf{EPE}(x_0)$ follows:

$$\mathbf{EPE}(x_0) = \sigma^2 + \mathbb{E}_{\mathcal{D}}[x_0^T (\mathbf{X}\mathbf{X}^T)^{-1} x_0 \sigma^2]. \quad (7)$$

Please give the steps and details of your proof.(Hint: $\mathbf{EPE}(x_0) = \mathbb{E}_{y_0|x_0} \mathbb{E}_{\mathcal{D}}[(y_0 - \hat{y}_0)^2]$, you can also refer to the proof progress of variance-bias decomposition on the page 45 of our reference book)