

Machine Learning - Homework 0

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Problem 1

I am Computer Vision student currently working on research in the area of Computational Imaging. I believe that Machine Learning provides a mathematical tool to channel human intuition to computation. While being widely applicable in Vision and Imaging techniques, I hope to have fun learning the math behind Machine Learning which helps me understand how and why the techniques work with more certainty.

Problem 2

- (a) False
- (b) False

Problem 3

- (a) Rank of A is 1
- (b)

$$Au + Bv = \begin{bmatrix} 8 \\ 16 \end{bmatrix}$$

- (c)

$$u^T Av = \begin{bmatrix} 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 20$$

- (d) $\|u\|_2 = \sqrt{2^2 + 1^2} = \sqrt{5}$

(e)

$$\nabla f(x) = 2(A + B)x = 2 \begin{bmatrix} 5 & 2 \\ 2 & 8 \end{bmatrix} x$$

$$\nabla f(v) = 2 \begin{bmatrix} 5 & 2 \\ 2 & 8 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 18 \\ 36 \end{bmatrix}$$

(f) Minimize $x^T(A + B)x$ such that $\|x\|_2 = 1$

$$\text{Minimize } L(x, \lambda) = x^T(A + B)x - \lambda(x^T x - 1)$$

$$\frac{\partial L}{\partial x} = (A + B)x - \lambda x = 0$$

$$(A + B)x = \lambda x$$

Eigenvector for $(A + B)$ gives the value of x which minimizes the function

$f(x)$

Eigenvalues for $(A + B)$ are 4 and 9. The lowest eigenvalue minimizes the function. The Eigenvector of unit length corresponding the eigenvalue 4

$$\text{is } \begin{bmatrix} \frac{-2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{bmatrix}$$

The value of f evaluated at this vector is 4.

Problem 4

(a) A and B are independent. The rest of the pairs are dependent.

(b)

$$\frac{P(CU \cap TU)}{P(TU)} = \frac{0.2}{0.5} = 0.4$$

(c) $P(X = 1) = 11/36$

$$P(X = 2) = 9/36$$

$$P(X = 3) = 7/36$$

$$P(X = 4) = 5/36$$

$$P(X = 5) = 3/36$$

$$P(X = 6) = 1/36$$

(d) $E(X) = 91/36 = 2.528$

(e) $E(\text{number of tosses until head come up}) = \sum_{n=1}^{\infty} n \frac{1}{5} \left(\frac{4}{5}\right)^{n-1} = 5$

(f) $E(\text{number of times the phrase appears in the sentence}) = P(\text{phrase matches starting at position 1}) + P(\text{phrase matches starting at position 2}) + \dots$
 $+ P(\text{phrase matches starting at position } n - 4)$
 $= (n - 4) P(\text{phrase matching at position 1})$
 $= (n - 4) \left(\frac{1}{4}\right)^5$

Problem 5

(a) $\lambda = 6.93 \times 10^{-7}$

(b) $E(X) = \text{Mean of the Gaussian Probability Distribution} = 0$
 $E(Y) = \text{Variance of the Gaussian Probability Distribution} = 1$

(c)

$$\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} p(x_1, x_2) dx_1 dx_2 = 1$$

$$\int_0^{0.5} \int_0^1 c dx_1 dx_2 = 1$$

$$c = 2$$

(d)

$$P(X_2 \geq X_1) = \int_0^{0.5} \int_{x_1}^1 c dx_1 dx_2 = \int_0^{0.5} 2(1-x_1) dx_1 = (2x_1 - x_1^2) \Big|_0^{0.5} = \frac{3}{4}$$

(e) X_1 and Y are not independent.

The probability space for Y (same as $p(x_1, x_2)$) is divided by the line $X_1 - 2X_2 = 0$.

$$E(Y) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} y \cdot p(x_1, x_2) dx_1 dx_2$$

$$E(Y) = \int_0^{0.5} \int_0^{\frac{x_1}{2}} 2 dx_1 dx_2 - \int_0^{0.5} \int_{\frac{x_1}{2}}^1 2 dx_1 dx_2$$

$$E(Y) = \frac{1}{8} - \frac{7}{8} = -\frac{3}{4}$$

(f) X_1 and Z are independent. $E(X_1 Z) = E(X_1)E(Z)$

$$E(Z) = 1/2 - 1/2 = 0$$

Thus, $E(X_1 Z) = 0$

References

[1] <http://cs229.stanford.edu/section/cs229-linalg.pdf>

[2] <http://cs229.stanford.edu/section/cs229-prob.pdf>

[3] Nayar, Shree. "Camera Calibration" Computer Vision(COMSW4731). Columbia University, New York. 10-30-2015. Lecture.