

HW9 solution

1. In machine learning, manifold is simply a curve surface.

A geodesic is a curve representing the shortest path between two points in a surface.

Geodesic distance is the shortest distance along such a curve in the manifold.

Here we give an example of a unit sphere ($x^2+y^2+z^2=1$) as 2D manifold embedded in 3D Euclid space.

(a) For two point (1,0,0) and (0,1,0), find the geodesic distance.

(b) For two point (1,0,0) and $(1/\sqrt{3}, 1/\sqrt{3}, 1/\sqrt{3})$

answer:

(a) the angle between (1,0,0) and (0,1,0) is $\pi/4$

the geodesic distance is there $2 \cdot \pi/4 = \pi/2$

(b) you can take inner product of (1,0,0) and $(1/\sqrt{3}, 1/\sqrt{3}, 1/\sqrt{3})$

$\cos \theta = 1/\sqrt{3}$

$\theta = 0.9553$

radius =1 , geodesic distance = 0.9553

cross entropy

1. First I would like you to evaluate entropy for a discrete probability distribution $p(x)$

Given a discrete random variable X , which takes values in the X and is distributed according to probability distribution $p(x)$

$$H(p) = - \sum_{x \in X} p(x) \log p(x)$$

Here in machine learning log refers to natural logarithm.

(a) For a Bernouli distribution with two probability

$p(x=0) = 0.3$

$p(x=1) = 0.7$

Evaluate the entropy.

Answer: simply take the log and multiply by p

$H(p) = 0.610$

(b) Now we want to generalize to a continuous random variable and hence $p(x)$ is the probability density function and summation is replaced with integral.

$$H(p) = \int p(x) \log p(x) dx$$

Evaluate the entropy for a Gaussian distribution (for $\sigma=1$, and $\mu=0$)

$$p(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp \left[-\frac{(x - \mu)^2}{2\sigma^2} \right]$$

Answer:

$$H(p) = \int \frac{1}{\sqrt{2\pi}} \exp \left[-\frac{(x)^2}{2} \right] \left[\left(-\frac{(x)^2}{2} \right) + \log \frac{1}{\sqrt{2\pi}} \right] dx = \left[-\frac{1}{2} - \log(\sqrt{2\pi}) \right]$$

(the correct definition is $H(p) = - \int p(x) \log p(x) dx$. From this definition,)

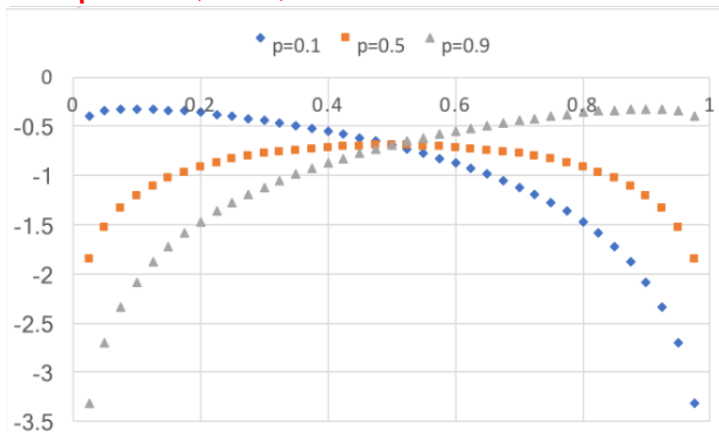
$$H(p) = \frac{1}{2} + \log(\sqrt{2\pi})$$

(c) cross entropy plot

Answer: cross entropy is defined as

$$H(p, q) = p \log q + (1-p) \log(1-q)$$

Fix $p = 0.1, 0.5$, and 0.9



You can see that $p=0.1, 0.5$, and 0.9 the cross entropy diverges at $q=0$ and $q=1$

(d) Kullback Liebler divergence

$$\begin{aligned} D_{\text{KL}}(P \parallel Q) &= \sum_{x \in \mathcal{X}} P(x) \ln \left(\frac{P(x)}{Q(x)} \right) \\ &= \frac{9}{25} \ln \left(\frac{9/25}{1/3} \right) + \frac{12}{25} \ln \left(\frac{12/25}{1/3} \right) + \frac{4}{25} \ln \left(\frac{4/25}{1/3} \right) \\ &= \frac{1}{25} (32 \ln(2) + 55 \ln(3) - 50 \ln(5)) \approx 0.0852996 \end{aligned}$$