

S1: Principles of Data Science

Problem Sheet 2

MPhil in Data Intensive Science

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Lectures 7 – 12

Topics covered: More distributions, generating from distributions, limit theorems, propagation of errors, estimates, minimum variance bound, the likelihood, maximum likelihood estimation, profile likelihood, extended likelihood, binned likelihood, least-squares, Wilks' theorem

1. For a two-dimensional normal distribution with parameters, $\mu_1 = 1$, $\mu_2 = 4$, $\sigma_1 = 3$, $\sigma_2 = 2$, $\rho = 0.5$ make plots of the conditional and marginal probabilities. Extension (don't waste too much time on this), what if this is now a 3D Gaussian? Can you think of ways of presenting the equivalent information?
2. Show that the mean and variance of the exponential distribution with p.d.f.

$$p(X; \lambda) = \lambda e^{-\lambda X} \quad (1)$$

are given by

$$\text{Mean:} \quad \mu = \frac{1}{\lambda} \quad (2)$$

$$\text{Variance:} \quad V(X) = \frac{1}{\lambda^2} \quad (3)$$

3. Show that the mean and variance of the χ^2 distribution with p.d.f.

$$p(X; k) = \frac{1}{2^{k/2} \Gamma(\frac{k}{2})} X^{k/2-1} e^{-X/2} \quad (4)$$

are given by

$$\text{Mean:} \quad \mu = k \quad (5)$$

$$\text{Variance:} \quad V(X) = 2k. \quad (6)$$

4. Write a simple accept-reject generator in python which can generate from an arbitrary function of a random variable, $f(X)$. Generate some samples for the following distributions:

(a) $f(x) = \cos^2(x)$

(b) $f(x) = \sin(x) + \cos(x) + 2$

(c) $f(x) = \frac{\sin(x) + \cos(x)}{\sinh(x) + \cosh(x)} + 25$

Think about how you might adapt this to work for two-dimensions and then n -dimensions. Can you think about ways which would speed up the generation?

Can you now make a comparison of generation efficiencies with a 1D normal distributions. Let's take a standard normal distribution (*i.e.* $\mu = 0$, $\sigma = 1$). Make a plot of the acceptance efficiency as a function of the generation range in terms of standard deviations. What is the approximate accept efficiency if you generate all the way out to 8 standard deviations?

5. Using the Jacobian matrix derive the standard error propagation formula for σ_f given σ_x and σ_y for the following transformations

(a) $f = x + y$

(b) $f = xy$

(c) $f = x/y$

(d) $f = \sin(x)$

(e) $f = \cos(x)$

assuming that x and y are independent.

6. Using the minimum variance bound show that the sample mean, \bar{X} , is an efficient estimate of the distribution mean, μ .
7. Use the maximum likelihood method to mathematically show that an estimate for the lifetime, τ , of a decay-time distribution, $(1/\tau)e^{-t/\tau}$, is given by the average of the observed decay times, t .
8. Show that when fitting a straight line to pairs of points (x_i, y_i) with the least squares method that estimates of the slope and intercept are given by $\hat{c} = \bar{y} - \hat{m}\bar{x}$ and $m = \text{cov}(x, y)/V(x)$.