

Problem Set 1

Due: October 28th, 2016

Problem sets are due at 5PM in the GSIs mailbox. You may work in groups, but each student should turn in their own write-up (including a printout of a narrated/commented and executed iPython Notebook if applicable). Please also e-mail a copy of any iPython Notebook to the GSI (if applicable).

1 Binomial distribution

You are conducting a survey of the presidential voting intentions of Cal undergraduates. Let $Y = 1$ if a randomly sampled Cal undergraduate plans to vote for Hillary Clinton, and zero if they plan to vote for an alternative candidate. Among the population of Cal undergrads $\theta = \Pr(Y = 1)$ is the true population frequency of individuals who intend to vote for Clinton. You take a random sample of size N from the Cal student body. Let $Z_N = \sum_{i=1}^N Y_i$ equal the total number of sampled students who indicate their intention to vote for Obama.

1. Derive a formula that can be used to calculate the ex ante (i.e., pre-sample) probability of the event that $Z_N < z$ for any $z \in \{1, 2, \dots, N\}$. Provide a 3 - 4 sentence written description of your reasoning.
2. Let $\bar{Y}_N = \frac{1}{N} \sum_{i=1}^N Y_i$. Using your answer above, provide an expression that can be used to calculate the ex ante probability of the event $\frac{\sqrt{N}(\bar{Y}_N - \theta)}{\sqrt{\theta(1-\theta)}} < c$.
3. Using your formula plot, in an iPython Notebook, $\Pr\left(\frac{\sqrt{N}(\bar{Y}_N - \theta)}{\sqrt{\theta(1-\theta)}} < c\right)$ as a function of c for $N = 5, 10, 100, 1000$ and $\theta = 1/2$. Make a single figure with 4 subplots arrayed 2×2 . Title each figure and label all axes.
4. Let $X \sim \mathcal{N}(0, 1)$. Plot $\Pr(X < c)$ as a function of c on *each* of the four plots created in the previous problem.
5. Repeat questions 3 and 4 with $\theta = 1/20$. Comment on your figures (4 - 6 sentences).

2 Binomial-Beta learning

Let θ , as before, denote the probability than a randomly sampled Cal undergraduate intends to vote for Hillary Clinton. Assume that your beliefs about θ are summarized by a prior distribution. In particular the probability that you assign to different possible values of θ is given by a beta(a, b) distribution (i.e., $\theta \sim \text{beta}(a, b)$). Let Z_N , also as before, equal the number of Cal students, out of a random sample of size N , who say they intend to vote for Clinton.

1. What is the conditional distribution of Z_N given θ ?
2. Calculate the joint distribution of Z_N and θ .

3. Calculate the conditional distribution of θ given Z_N . What is the mean of this distribution? Why might posterior be a good name for this distribution? (5 - 6 sentences)
4. Assume that $a = b = 1/2$. Comment on this prior (2 to 3 sentences).

3 Multivariate normal distribution

Let $\mathbf{Y} = (Y_1, \dots, Y_K)'$ be a $K \times 1$ random vector with density function

$$f(y_1, \dots, y_K) = (2\pi)^{-K/2} |\Sigma|^{-1/2} \exp\left(-\frac{1}{2} (\mathbf{y} - \mu)' \Sigma^{-1} (\mathbf{y} - \mu)\right),$$

for Σ a symmetric positive definite $K \times K$ matrix and μ a $K \times 1$ vector. We say that \mathbf{Y} is a multivariate normal random variable with mean μ and covariance Σ or

$$\mathbf{Y} \sim \mathcal{N}(\mu, \Sigma).$$

The multivariate normal distribution arises frequently in econometrics and a mastery of its basic properties is essential for both applied and theoretical work in econometrics. This problem provides an opportunity for you to review and/or learn some of these properties. There are many useful references on the multivariate normal distribution, for example, T. W. Anderson's *An Introduction to Multivariate Statistical Analysis*.

1. Let C be a $K \times K$ nonsingular matrix. Show that $\mathbf{Z} = C\mathbf{Y}$ is distributed according to $\mathcal{N}(C\mu, C\Sigma C')$.
2. Partition $\mathbf{Y} = (\mathbf{Y}'_1, \mathbf{Y}'_2)'$ into $K_1 \times 1$ and $K_2 \times 1$ sub-vectors with $K_1 + K_2 = K$. Let $\mu = (\mu'_1, \mu'_2)'$ and

$$\Sigma = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}$$

be conformable partitions of μ and Σ (note that symmetry implies $\Sigma_{12} = \Sigma'_{21}$). Show that \mathbf{Y}_1 and \mathbf{Y}_2 are independent random vectors if $\Sigma_{12} = \Sigma'_{21} = \mathbf{0}\mathbf{0}'$ (i.e., a matrix of zeros).

3. Let

$$C = \begin{pmatrix} I_{K_1} & -\Sigma_{12}\Sigma_{22}^{-1} \\ 0 & I_{K_2} \end{pmatrix}$$

for I_P a $P \times P$ identity matrix. Derive the distribution of $\mathbf{Z} = C\mathbf{Y}$. Are the first K_1 elements of \mathbf{Z} independent from the second K_2 ? Interpret your result?

4. Let D be a $P \times K$ ($P \leq K$) matrix of rank P . Arrange the first P columns of D , denoted by D_{11} , such that they are non-singular. Denote the remaining $K - P$ columns by D_{12} . Find a $(K - P) \times K$ matrix E such that

$$\begin{pmatrix} \mathbf{Z} \\ \mathbf{X} \end{pmatrix} = \begin{pmatrix} D \\ E \end{pmatrix} \mathbf{Y}$$

is a non-singular transformation of \mathbf{Y} . Finally show that \mathbf{Z} is distributed according to $\mathcal{N}(D\mu, D\Sigma D')$.

5. Consider the partition of \mathbf{Y} introduced in Problem 2 above. Derive the conditional distribution of \mathbf{Y}_1 given $\mathbf{Y}_2 = \mathbf{y}_2$.

6. Let $\{\mathbf{Y}_i\}_{i=1}^N$ be a random sample of size N drawn from the multivariate normal population described above. Show that $\sqrt{N}(\bar{\mathbf{Y}} - \mu)$ is a $\mathcal{N}(0, \Sigma)$ random variable for $\bar{\mathbf{Y}} = \frac{1}{N} \sum_{i=1}^N \mathbf{Y}_i$, the sample mean (HINT: Use independence of the $i = 1, \dots, N$ draws and your result in Problem 4 above).
7. Let $\mathbf{W} = N \cdot (\bar{\mathbf{Y}} - \mu)' \Sigma^{-1} (\bar{\mathbf{Y}} - \mu)$. Show that $\mathbf{W} \sim \chi_K^2$ (i.e., \mathbf{W} is a chi-square random variable with K degrees of freedom).
8. Let $\chi_K^{2, 1-\alpha}$ be the $(1 - \alpha)^{th}$ quantile of the χ_K^2 distribution (i.e., the number satisfying the equality $\Pr(\mathbf{W} \leq \chi_K^{2, 1-\alpha}) = 1 - \alpha$ with \mathbf{W} a chi-square random variable with K degrees of freedom). Let D be a $P \times K$ ($P \leq K$) matrix of rank P and d a $P \times 1$ vector of constants. Consider the hypothesis

$$\begin{aligned} H_0 : D\mu &= d \\ H_1 : D\mu &\neq d. \end{aligned}$$

Maintaining H_0 derive the sampling distribution of $D\bar{\mathbf{Y}}$ as well as that of

$$\mathbf{W} = N \cdot (D\bar{\mathbf{Y}} - d)' (D\Sigma D)^{-1} (D\bar{\mathbf{Y}} - d).$$

You observe that, for the sample in hand, $\mathbf{W} > \chi_P^{2, 1-\alpha}$ for $\alpha = 0.05$. Assuming H_0 is true, what is the ex ante (i.e., pre-sample) probability of this event? What are you inclined to conclude after observing \mathbf{W} in the sample in hand?

4 Normal learning

Let θ be some parameter of interest. For example the average number of hours per week a graduate student in economics spends studying. Upon arriving in graduate school you summarize your beliefs/uncertainty about θ by assuming that $\theta \sim \mathcal{N}(\bar{\theta}, \frac{1}{\rho_\theta})$.

1. If $\bar{\theta} = 10$ and $\rho_\theta = 1/10$, then what is the probability that you assign to the possibility that θ exceeds 40 hours a week? Is less than 10 hours per week?
2. Let $S_t = \theta + \epsilon_t$ with $\epsilon_t \sim \mathcal{N}(0, \frac{1}{\rho_\epsilon})$. Compute the conditional distribution of θ given S_1 .
3. You observe the additional signals S_2, \dots, S_T . Compute the conditional distribution of θ given all T signals.