Math 70 Homework 6

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Problem 1

Instructions

(Problem 5.5 pg. 309) Let X and Y be two independent normally distributed random variables with means μ_X and μ_Y and variances σ_X^2 and σ_Y^2 .

- (a) Prove that a necessary condition for $X \prec Y$ is that $\mu_X < \mu_Y$.
- (b) Provide an example when $\mu_X < \mu_Y$ but $X \not\prec Y$.
- (c) Prove that if $\sigma_X = \sigma_Y$, then $\mu_X < \mu_Y$ is a necessary and sufficient condition for $X \prec Y$.

Solution

(a) To prove that a necessary condition for $X \prec Y$ is that $\mu_X \leq \mu_Y$, we'll first consider the definition of uniformly smaller. A random variable X is said to be uniformly smaller than Y, denoted by $X \prec Y$, if for any x:

Let $\Phi(x)$ denote the standard normal cdf. Since X and Y are normally distributed, we can express the above as:

$$\Phi(\frac{x - \mu_X}{\sigma_X}) > \Phi(\frac{x - \mu_Y}{\sigma_Y})$$

Let $z = \frac{x - \mu_X}{\sigma_X}$. Then, we have:

$$\Phi(z) > \Phi(\frac{(\sigma_X z + \mu_X) - \mu_Y}{\sigma_Y}) = \Phi(\frac{\sigma_X}{\sigma_Y} z + \frac{\mu_X - \mu_Y}{\sigma_Y})$$

Since $\Phi(x)$ is an increasing function and the inequality above must hold true $\forall z$ given that $X \prec Y$, we can see that the only way to satisfy this inequality is for $\frac{\sigma_X}{\sigma_Y}z + \frac{\mu_Y - \mu_X}{\sigma_Y} < z$. Because the standard deviations of X and Y are positive, this implies that:

$$\mu_X - \mu_Y < 0 \Rightarrow \mu_X < \mu_Y$$

(b) In this example, let:

$$X \sim \mathcal{N}(\mu_X = 0, \sigma_X^2 = 4)$$

$$Y \sim \mathcal{N}(\mu_Y = 1, \sigma_Y^2 = 1)$$

In this example, $\mu_X < \mu_Y$, but let's check if $X \prec Y$:

$$Pr(X < x) = \Phi(\frac{x - 0}{\sqrt{4}}) = \Phi(\frac{x}{2})$$

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$$Pr(Y < x) = Phi(\frac{x-1}{\sqrt{1}}) = \Phi(x-1)$$

Let x = 6. Then, we have:

$$Pr(X < 6) = \Phi(3)$$

$$Pr(Y < 6) = \Phi(5)$$

Because $\Phi(x)$ is an increasing function, we can see that:

$$\Phi(3) < \Phi(5) \Rightarrow Pr(X < 6) < Pr(Y < 6) \Rightarrow X \not\prec Y$$

And therefore, $X \not\prec Y$.

(c) From part (a), we know that $X \prec Y$ is equivalent to:

$$\Phi(z) > \Phi(\frac{\sigma_X}{\sigma_Y}z + \frac{\mu_X - \mu_Y}{\sigma_Y})$$

If we let $\sigma_X = \sigma_Y$, then we have:

$$\Phi(z) > \Phi(z + \frac{\mu_X - \mu_Y}{\sigma_Y})$$

We know that σ_Y is positive, so therefore if $\mu_X < \mu_Y$, then $\frac{\mu_X - \mu_Y}{\sigma_Y} < 0$, making the above inequality true $\forall z$, which implies that $X \prec Y$, making $\mu_X < \mu_Y$ a necessary and sufficient condition for $X \prec Y$ when $\sigma_X = \sigma_Y$.

Problem 2

Instructions

File bp.csv contains blood pressure (BP) for normal patients (controls, high=0) and hypertension patients (high=1).

- (a) Display two cdfs to demonstrate that BP among normal patients is uniformly smaller than among hypertension patients.
- (b) Display the data-driven ROC curve for the identification of normal patients and the superimposed binormal counterpart.
- (c) Compute and display AUCs using three methods: (1) empirical, as the sum or rectangles, (2) empirical, using vectorized computation, and (3) theoretical, using the formula.
- (d) The cost associated with overlooking a hypertension patient is \$10K and the cost of the false identification of hypertension is \$1K. Display the data-driven total cost and the superimposed continuous counterpart as a function of the threshold along with the respective optimal thresholds.
- (e) Display the optimal threshold on the binormal ROC curve and the respective BP scale using axis(side=3).

Solution

Explanation

- (a)
- (b)
- (c)

(d)

(e)

AUC calculated using sum of rectangles: 0.9019565 AUC calculated using vectorized computation: 0.8970754 AUC calculated using theoretical formula: 0.8936977

 \mathbf{Code}

Plots

Empirical CDF for Blood Pressure of Hypertension & Non-Hypertension Patients

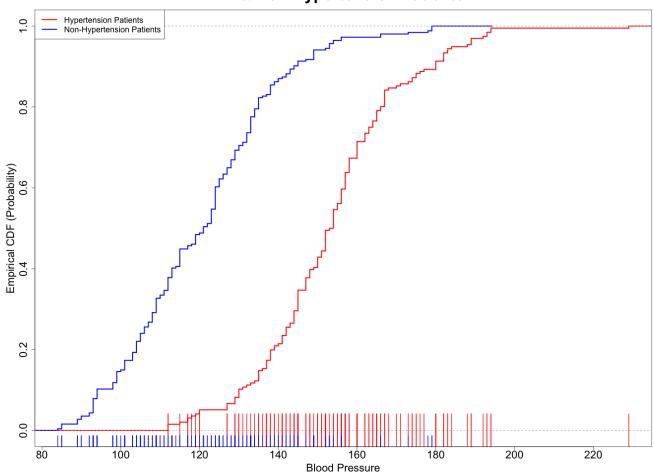


Figure 1: (a) Empirical CDFs

ROC Curve for Blood Pressure

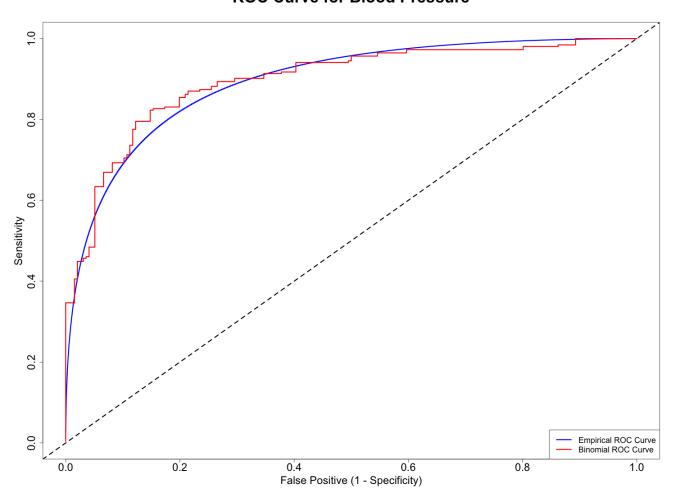


Figure 2: (b) ROC Curve

Total Cost vs Threshold

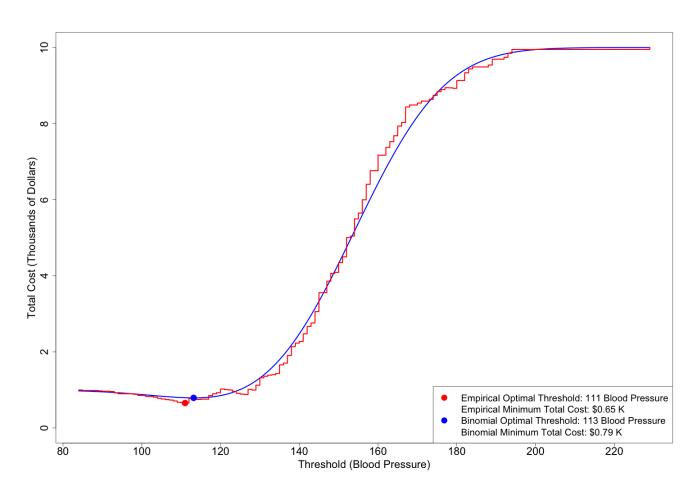


Figure 3: (d) Total Cost

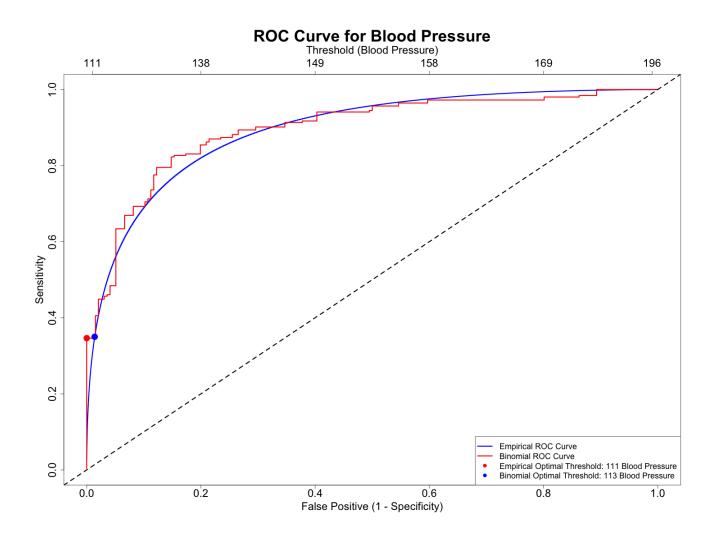


Figure 4: (e) Optimal Threshold & ROC Curve