

Record all answers in the blue books. Show your work.

- From a random sample of n pairs (X_i, Y_i) we compute the sample correlation coefficient $\hat{\rho}_n$, which estimates the population correlation ρ . It is known that $\sqrt{n}(\hat{\rho}_n - \rho) \rightarrow_d N(0, (1 - \rho^2)^2)$ as $n \rightarrow \infty$. With $\phi = g(\rho) = (1/2) \log\left(\frac{1+\rho}{1-\rho}\right)$, and with $\hat{\phi}_n = g(\hat{\rho}_n)$, show that $\sqrt{n}(\hat{\phi}_n - \phi) \rightarrow_d N(0, 1)$. How can this result be used to obtain an approximate 95% confidence interval for ρ ?
- Independent random variables X_1, X_2, \dots, X_n have a Pareto distribution with probability density function, for $x \geq 1$,

$$f(x; \theta) = \frac{\theta}{x^{\theta+1}}$$

where $\theta > 0$ is an unknown parameter. A sample is realized as x_1, x_2, \dots, x_n .

- Write the likelihood, log likelihood, and score functions and find the maximum likelihood estimate $\hat{\theta}$.
 - Describe how to use the nonparametric bootstrap to estimate the standard error of $\hat{\theta}$.
- Counts $X_1, \dots, X_n, X_{n+1}, \dots, X_{2n}$ are independent and Poisson distributed. For $i = 1, 2, \dots, n$, the Poisson mean is $a\theta$ and for $i = n + 1, \dots, 2n$ the Poisson mean is $b\theta$, where $a, b > 0$ are distinct known constants. The counts $\{X_i\}$ are not observed directly. Instead we observe $Y_i = 1[X_i \geq 1]$ for all i .
 - Give a formula for the log likelihood $l_c(\theta)$ which would have been obtained had we been able to observe $\{X_i\}$.
 - For an initial guess θ_1 of θ , evaluate

$$J(\theta, \theta_1) = E_{\theta_1} [l_c(\theta) | Y_i = y_i \text{ for all } i],$$

where randomness is in the latent X_i 's.

- Derive the value θ_2 which maximizes $J(\theta, \theta_1)$ in θ .
- In words, briefly explain the EM algorithm in the present context.