

Stat 134

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Follows Jim Pitman's book: Probability Sections 6.5

Bivariate Normal

Let (X,Y) be independent Normal variables. The joint density: $f(x,y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x^2+y^2)}$

> ρ = 0; 0 X

Bivariate Normal

<u>Question</u>: What is the joint distribution of (X,Y)

Where X = child height and Y = parent height?

We expect X and Y to be Gaussian

•However, X and Y are not independent: p(X,Y) > 0.

Bivariate Normal

Intuitive sample of Gaussian (X,Y) with $\rho = 0.7$



Bivariate Normal Intuitive sample of X,Y with $\rho(X,Y) = 1$.





Standard Bivariate Normal

Def: We say that X and Y have standard bivariate normal distribution with correlation ρ if and only if

 $V = \rho X + \sqrt{1 - \rho^2 Z}$ Where X and Z are independent N(0,1). Claim: If (X,Y) is p correlated bivariate normal then: $f(x,y) = \frac{1}{2\pi\sqrt{1-\rho^2}} \exp\{-\frac{1}{2(1-\rho^2)} (x^2 - 2\rho x y + y^2)\}$ Marginals: $X \sim N(0,1); Y \sim N(0,1)$ Conditionals: $X|Y=y \sim N(\rho y, 1-\rho^2); Y|X=x \sim N(\rho x, 1-\rho^2)$ **Independence:** X,Y are independent if and only if $\rho = 0$

Symmetry: (Y,X) is p correlated bivariate normal.

Bivariate Normal Distribution

<u>Definition</u>: We say that the random variables U and V have bivariate normal distribution with parameters μ_{U} , μ_{V} , σ^{2}_{U} , σ^{2}_{V} and ρ if and only if the standardized variables

 $X=(U - \mu_U)/\sigma_U$ $Y=(V - \mu_V)/\sigma_V$

have standard bivariate normal distribution with correlation ρ .

Errors in Voting Machines



There are two candidates in a certain election: Mr.B and Mr.K. We use a voting machine to count the votes. Suppose that the voting machine at a particular polling station is somewhat capricious – it flips the votes with probability ε .

A voting Problem

•Consider a vote between two candidates where:

At the morning of the vote:
Each voter tosses a coin and votes
according to the outcome.
Assume that the winner of the vote is
the candidate with the majority of votes.

- $\boldsymbol{\cdot}$ In other words let $V_i \in \{\pm \ 1\}$ be the vote of voter i. So
 - if $V = \sum_i V_i > 0$ then +1 wins;
 - otherwise -1 wins.





<u>A mathematical model of voting machines</u>

Which voting schemes are more robust against noise? <u>Simplest model of noise</u>: The voting machine flips each vote independently with probability ε .



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On Errors in voting machines
Question: Let V_i = intended vote i.
                W_i = registered vote i.
What is dist of V = \sum_{i=1}^{n} V_i, W = \sum_{i=1}^{n} W_i for large n?
Answer: V \sim N(0,n); W \sim N(0,n).
Question: What is P[V > 0]? P[W > 0]?
<u>Answer:</u> ~<sup>1</sup>/<sub>2</sub>.
Question: What is the probability that machine errors flipped
the elections outcome?
Answer: P[V > 0, W < 0] + P[V < 0, W > 0]?
         = 1 - 2 P[V > 0, W > 0].
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On Errors in voting machines

Answer continued: Take $(X,Y) = (V,W)/n^{1/2}$. Then (X,Y) is bivarite-normal where SD(X) = SD(Y) = 1 and $\rho = \rho(X,Y) = \rho(V,W)/n = \rho(V_i,W_i) = 1 - 2 \epsilon.$ Need to find 1-2P[X > 0, Y > 0]. <u>Answer continued</u>: Let $\rho = \cos \theta$. Then we need to find: $P[X > 0, Y > 0] = P[X \cos 0 + Z \sin 0 > 0, X \cos \theta + Z \sin \theta > 0] =$ = $P[X > 0, Z > -(ctg \theta) X] = (\pi - \theta)/2\pi =$ $=\frac{1}{2}(1-(\arccos \rho)/\pi).$ So 1-2P[X > 0, Y > 0] = $(\arccos \rho)/\pi$ VOTE HERE = $(arcos(1-2\varepsilon))/\pi$

Majority and Electoral College

- Probability of error ~ $\epsilon^{1/2}$
- Result is essentially due to Sheppard (1899): "On the application of the theory of error to cases of normal distribution and normal correlation".
- + For $n^{1/2} \times n^{1/2}$ electoral college f $\epsilon^{1/4}$



Conditional Expectation given an interval

Suppose that (X,Y) has the Standard Bivariate Normal Distribution with correlation p.

Question: For a < b, what is the E(Y|a < X <b)?



$$E(Y|a < X < b) = \frac{\int_a^b \int_{-\infty}^\infty yf(x,y)dydx}{\int_a^b \int_{-\infty}^\infty f(x,y)dydx}$$
$$= \frac{\int_a^b \int_{-\infty}^\infty yf_y(y|X=x)f_X(x)dydx}{\int_a^b f_X(x)dx}$$
$$= \frac{\int_a^b E(Y|X=x)f_X(x)dydx}{\int_a^b f_X(x)dx}.$$

Conditional Expectation given an interval Solution: $E(Y | a < X < b) = \int_{a}^{b} E(Y | X = x) f_{X}(x) / \int_{a}^{b} f_{X}(x) dx$

We know that $f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$ and $\int_a^b f_X(x) dx = \Phi(b) - \Phi(a)$. Since $Y = \rho X + \sqrt{1 - \rho^2} Z$, where X & Z are independent

We have: $E(Y|X=x) = E(\rho X + \sqrt{1 - \rho^2}Z|X=x) = \rho x$

$$E(Y|a < X < b) = \frac{\rho}{\sqrt{2\pi}} \int_{a}^{b} x e^{-\frac{x^{2}}{2}} dx / (\Phi(b) - \Phi(a))$$
$$= \frac{\rho}{\sqrt{2\pi}} [e^{-\frac{a^{2}}{2}} - e^{-\frac{b^{2}}{2}}] / (\Phi(b) - \Phi(a))$$

Linear Combinations of indep. Normals

Claim: Let $V = \sum_{i=1}^{n} a_i Z_i$, $W = \sum_{i=1}^{n} b_i Z_i$ where Z_i are independent normal variables $N(\mu_i, \sigma_i^2)$. Then (V,U) is bivariate normal.

Problem: calculate $\mu_V, \mu_W, \sigma_V, \sigma_W$ and ρ