Class 5 - Statistics and Probability (part I)

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Probability review

- ✓ Experiment: Any procedure that can be infinitely repeated and has a well-defined set of outcomes.
- ✓ Random variable: A variable that takes on numerical values and the outcome is determined by experiment.
 - Discrete random variable: Number of heads in a coin-flipping experiment, number of times observing
 a 4 when throwing a dice, letter GPA, number of bedrooms in a house, number of stocks in a portfolio
 and ...
 - Continuous random variable: Height, Weight, time, distance, numerical GPA, GDP, int rate, prices and
 ...
- ✓ Probability Density Function (PDF): summarizes the probabilities of outcomes!
- ✓ Cumulative Distribution Function (CDF): summarizes the cumulative probabilities of outcomes!

Probability review

✓ Conditional probabilities: A conditional probability is the probability of an event, given some other event has already occurred

$$P(Y|X) = \frac{P(Y \cap X)}{P(X)}$$

✓ Independence

$$P(Y \cap X) = P(Y)P(X) \rightarrow P(Y|X) = P(Y)$$

If Y and X are independent random variables, then knowledge of the value taken on by X tells us nothing about the probability that Y takes on various values and vice versa.

✓ Measures of central tendency:

1. Expected value

- If X is a discrete random variable: $E(X) = \sum_{i=1}^{k} x_i f(x_i)$
- If X is a continuous random variable: $E(X) = \int_{-\infty}^{\infty} xf(x)$

2. Median

- Median m, is a value such that one half of the area under the pdf is to the left of m and one-half of the area is to the right of m.
- ☐ Median is less sensitive than the average to extreme values (outliers). Ex: median income or median housing prices (why?)
- ☐ If X has a symmetric distribution about the mean, then the expected value and median are equal.
- ☐ Properties of Expected values:
- 1. For any constant c, E(c) = c
- 2. For any constant a and b, E(aX + b) = aE(X) + b
- 3. The Expected value of the sum is equal to the sum of expected values.

✓ Conditional Expectation:

$$E(Y|X) = \sum y_i f(Y|X)$$

- ☐ Properties of Conditional expectation:
- 1. Mean Independence: If X and Y are independent, then E(Y|X) = E(Y)This means that the expected value of Y given X does not depend on X.
- 2. E[f(X)|X] = f(X)Intuitively, this simply means that if we know X, then we also know f(X)
- 3. For functions f(X) and g(X)

$$E[\{f(X)Y + g(X)\} | X] = f(X) E(Y|X) + g(X)$$

- ✓ Measures of dispersion:
 - 1. Variance: How far, on average, is X from its mean!
 - 2. Standard Deviation
- \Box Let $E(X) = \mu$

$$Var(X) = E[(X - \mu)^2] = E(X^2) - \mu^2$$
$$sd(X) = \sqrt{Var(X)}$$

- ☐ Standard deviation is easier to interpret than variance because it has the same unit as the expected value.
- ☐ Properties of variance and standard deviation
- 1. Variance of any constant is zero
- 2. For any constant a and b, $Var(aX + b) = a^2Var(X)$
- 3. The variance of the sum is **NOT** equal to the sum of variances.

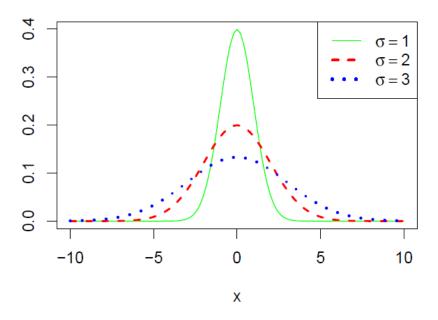


Figure 2: Normal distributions with different Standard deviations

Standardizing a random variable (Z transform of X)

- ✓ Different random variables can have very different units!
 - Number of heads when flipping a coin 3 times (0,1,2,3)
 - Stock price in the past 12 months! (\$120-220)
 - Height of US teenagers at age 11 (110-160 cm)
- ✓ These random variables can be transformed (standardized) such that they have the same mean (0) and the same variance (1)

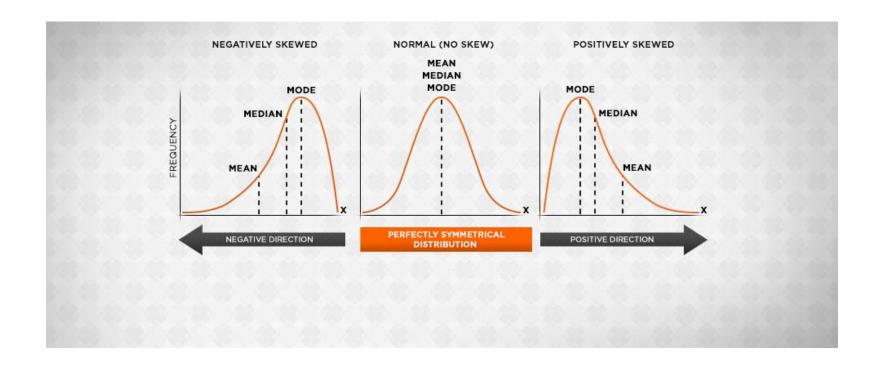
$$Z = \frac{X-\mu}{\sigma}$$

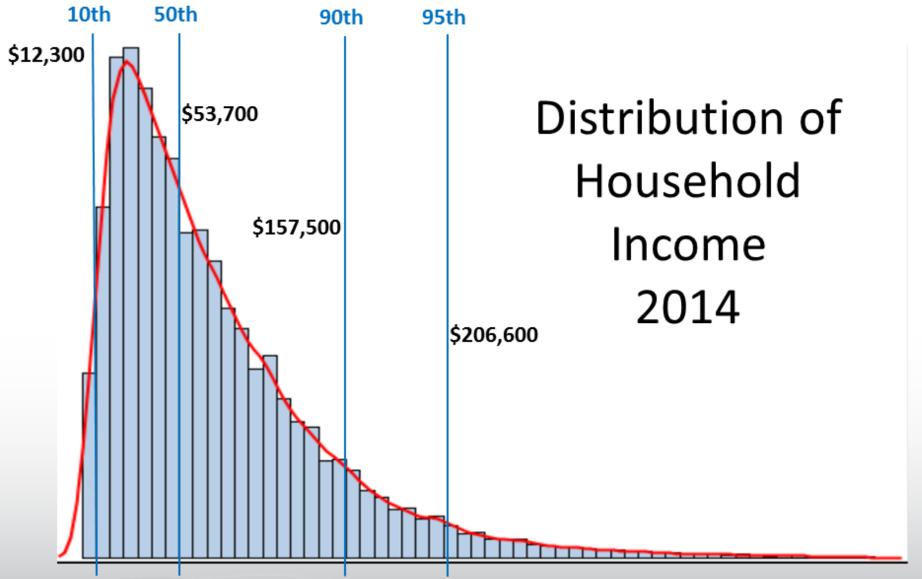
Z measures how many standard deviations X is away from it's mean.

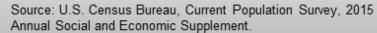
✓ More importantly, the probability distribution of these standardized random variables are often virtually identical!

- ✓ Measures of symmetricity:
 - 1. Skewness: Measures whether a distribution is symmetric about its mean!

$$E(Z^3) = E[(X - \mu)^3]/\sigma^3$$







Features of joint distributions

- ✓ Measures of Association (Joint variability):
 - 1. Covariance: How, on average, two random variables vary with one another.

$$Cov(X,Y) = E[(X - \mu_X)(Y - \mu_Y)] = E(XY) - \mu_X \mu_Y$$

- ☐ Properties of Covariance
- 1. If X and Y are independent, then Cov(X,Y)=0
- 2. For any constant a_1 , a_2 , b_1 , b_2 , $Cov(a_1X+b_1,a_2Y+b_2)=a_1a_2Cov(X,Y)$
- ☐ Variance of sums of random variables

$$Var(aX + bY) = a^2Var(X) + b^2Var(Y) + 2ab\ Cov(X, Y)$$

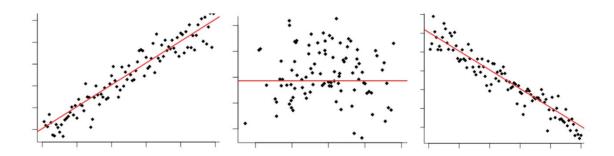
Implication! If X and Y are negatively correlated, then the Var(sum) < Sum(var)

Features of joint distributions

2. Correlation: On average, measures the linear relationship between two random variables Covariance depends on units! Example: Cov(education, earnings (\$ or thousand \$)) Correlation fixes this deficiency of covariance:

$$Corr(X,Y) = \frac{Cov(X,Y)}{sd(X) sd(Y)} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y} = \rho_{XY}$$

- ☐ Properties of Correlations
- 1. $-1 \leq Corr(X,Y) \leq 1$



2. For any constant a_1 , a_2 , b_1 , b_2 , $Corr(a_1X + b_1, a_2Y + b_2) = Corr(X, Y)$

Independence VS. Mean independence VS. Zero Correlation

Independence ⇒ Mean independence ⇒ Zero correlation

$$P(Y|X) = P(Y) \Rightarrow E(Y|X) = E(Y) \Rightarrow Corr(Y,X) = 0$$

BUT

Independence

✓ Mean independence
✓ Zero correlation

- ✓ Conditional expectation captures the nonlinear relationship between X and Y.
- ✓ Correlation captures the linear relationship between X and Y