

Expected values

Statistical Inference

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Expected values

- The **expected value** or **mean** of a random variable is the center of its distribution
- For discrete random variable X with PMF p(x), it is defined as follows

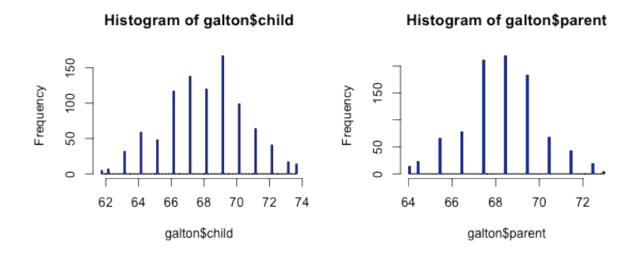
$$E[X] = \sum_x x p(x).$$

where the sum is taken over the possible values of x

 \cdot E[X] represents the center of mass of a collection of locations and weights, $\{x,p(x)\}$

Find the center of mass of the bars

Loading required package: MASS



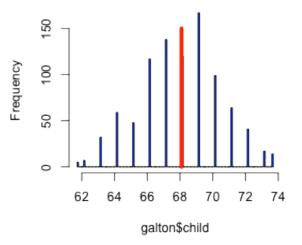
Using manipulate

```
library(manipulate)
myHist <- function(mu){
  hist(galton$child,col="blue",breaks=100)
  lines(c(mu, mu), c(0, 150),col="red",lwd=5)
  mse <- mean((galton$child - mu)^2)
  text(63, 150, paste("mu = ", mu))
  text(63, 140, paste("Imbalance = ", round(mse, 2)))
}
manipulate(myHist(mu), mu = slider(62, 74, step = 0.5))</pre>
```

The center of mass is the empirical mean

```
hist(galton$child, col = "blue", breaks = 100)
meanChild <- mean(galton$child)
lines(rep(meanChild, 100), seq(0, 150, length = 100), col = "red", lwd = 5)</pre>
```

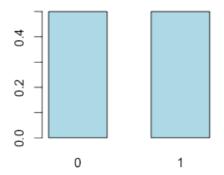
Histogram of galton\$child



- Suppose a coin is flipped and X is declared 0 or 1 corresponding to a head or a tail, respectively
- What is the expected value of X?

$$E[X] = .5 \times 0 + .5 \times 1 = .5$$

Note, if thought about geometrically, this answer is obvious; if two equal weights are spaced at 0 and 1, the center of mass will be .5



- \cdot Suppose that a die is rolled and X is the number face up
- What is the expected value of X?

$$E[X] = 1 imes rac{1}{6} + 2 imes rac{1}{6} + 3 imes rac{1}{6} + 4 imes rac{1}{6} + 5 imes rac{1}{6} + 6 imes rac{1}{6} = 3.5$$

· Again, the geometric argument makes this answer obvious without calculation.

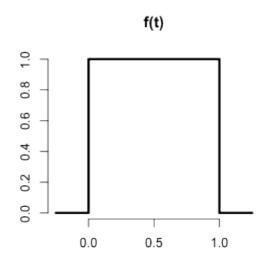
Continuous random variables

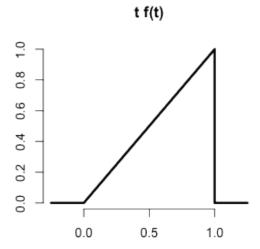
 \cdot For a continuous random variable, X, with density, f, the expected value is defined as follows

$$E[X]$$
 = the area under the function $tf(t)$

· This definition borrows from the definition of center of mass for a continuous body

- Consider a density where f(x) = 1 for x between zero and one
- · (Is this a valid density?)
- \cdot Suppose that X follows this density; what is its expected value?





Rules about expected values

- The expected value is a linear operator
- If a and b are not random and X and Y are two random variables then
 - E[aX + b] = aE[X] + b
 - E[X + Y] = E[X] + E[Y]

 You flip a coin, X and simulate a uniform random number Y, what is the expected value of their sum?

$$E[X + Y] = E[X] + E[Y] = .5 + .5 = 1$$

- Another example, you roll a die twice. What is the expected value of the average?
- Let X_1 and X_2 be the results of the two rolls

$$E[(X_1+X_2)/2]=rac{1}{2}\left(E[X_1]+E[X_2]
ight)=rac{1}{2}\left(3.5+3.5
ight)=3.5$$

- 1. Let X_i for $i=1,\ldots,n$ be a collection of random variables, each from a distribution with mean μ
- 2. Calculate the expected value of the sample average of the X_i

$$egin{aligned} Eiggl[rac{1}{n}\sum_{i=1}^n X_iiggr] &=rac{1}{n}Eiggl[\sum_{i=1}^n X_iiggr] \ &=rac{1}{n}\sum_{i=1}^n E[X_i] \ &=rac{1}{n}\sum_{i=1}^n \mu = \mu. \end{aligned}$$

Remark

- Therefore, the expected value of the **sample mean** is the population mean that it's trying to estimate
- When the expected value of an estimator is what its trying to estimate, we say that the estimator is **unbiased**

The variance

- · The variance of a random variable is a measure of spread
- If X is a random variable with mean μ , the variance of X is defined as

$$Var(X) = E[(X - \mu)^2]$$

the expected (squared) distance from the mean

· Densities with a higher variance are more spread out than densities with a lower variance

Convenient computational form

$$Var(X) = E[X^2] - E[X]^2$$

- If a is constant then $Var(aX) = a^2 Var(X)$
- The square root of the variance is called the **standard deviation**
- \cdot The standard deviation has the same units as X

- · What's the sample variance from the result of a toss of a die?
 - E[X] = 3.5

-
$$E[X^2]=1^2 imes rac{1}{6}+2^2 imes rac{1}{6}+3^2 imes rac{1}{6}+4^2 imes rac{1}{6}+5^2 imes rac{1}{6}+6^2 imes rac{1}{6}=15.17$$

•
$$Var(X) = E[X^2] - E[X]^2 \approx 2.92$$

- What's the sample variance from the result of the toss of a coin with probability of heads (1) of p?
 - $E[X] = 0 \times (1-p) + 1 \times p = p$
 - $E[X^2] = E[X] = p$
- $Var(X) = E[X^2] E[X]^2 = p p^2 = p(1-p)$

Interpreting variances

- Chebyshev's inequality is useful for interpreting variances
- This inequality states that

$$P(|X - \mu| \ge k\sigma) \le \frac{1}{k^2}$$

 \cdot For example, the probability that a random variable lies beyond k standard deviations from its mean is less than $1/k^2$

$$egin{aligned} 2\sigma &
ightarrow 25\% \ 3\sigma &
ightarrow 11\% \ 4\sigma &
ightarrow 6\% \end{aligned}$$

· Note this is only a bound; the actual probability might be quite a bit smaller

- IQs are often said to be distributed with a mean of 100 and a sd of 15
- What is the probability of a randomly drawn person having an IQ higher than 160 or below 40?
- Thus we want to know the probability of a person being more than 4 standard deviations from the mean
- Thus Chebyshev's inequality suggests that this will be no larger than 6\%
- IQs distributions are often cited as being bell shaped, in which case this bound is very conservative
- The probability of a random draw from a bell curve being 4 standard deviations from the mean is on the order of 10^{-5} (one thousandth of one percent)

- · A former buzz phrase in industrial quality control is Motorola's "Six Sigma" whereby businesses are suggested to control extreme events or rare defective parts
- \cdot Chebyshev's inequality states that the probability of a "Six Sigma" event is less than $1/6^2 pprox 3\%$
- · If a bell curve is assumed, the probability of a "six sigma" event is on the order of 10^{-9} (one ten millionth of a percent)