


# Probability Distributions

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# Random Variable

- A random variable  $x$  takes on a defined set of values with different probabilities.
  - For example, if you roll a die, the outcome is random (not fixed) and there are 6 possible outcomes, each of which occur with probability one-sixth.
  - For example, if you poll people about their voting preferences, the percentage of the sample that responds “Yes on Proposition 100” is also a random variable (the percentage will be slightly differently every time you poll).
- Roughly, probability is how frequently we expect different outcomes to occur if we repeat the experiment over and over (“frequentist” view)

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## Random variables can be discrete or continuous

- **Discrete** random variables have a countable number of outcomes
  - Examples: Dead/alive, treatment/placebo, dice, counts, etc.
- **Continuous** random variables have an infinite continuum of possible values.
  - Examples: blood pressure, weight, the speed of a car, the real numbers from 1 to 6.

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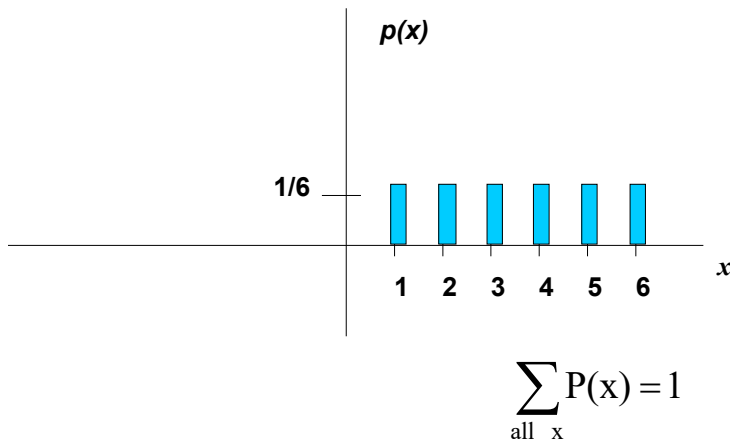


## Probability functions

- A probability function maps the possible values of  $x$  against their respective probabilities of occurrence,  $p(x)$
- $p(x)$  is a number from 0 to 1.0.
- The area under a probability function is always 1.

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## Discrete example: roll of a die



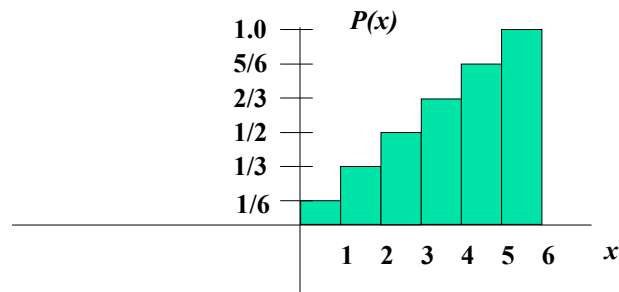
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## Probability mass function (pmf)

$x$	$p(x)$
1	$p(x=1)=1/6$
2	$p(x=2)=1/6$
3	$p(x=3)=1/6$
4	$p(x=4)=1/6$
5	$p(x=5)=1/6$
6	$p(x=6)=1/6$
	1.0

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## Cumulative distribution function (CDF)



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## Cumulative distribution function

x	$P(x \leq A)$
1	$P(x \leq 1) = 1/6$
2	$P(x \leq 2) = 2/6$
3	$P(x \leq 3) = 3/6$
4	$P(x \leq 4) = 4/6$
5	$P(x \leq 5) = 5/6$
6	$P(x \leq 6) = 6/6$

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## Examples

1. What's the probability that you roll a 3 or less?

$$P(x \leq 3) = 1/2$$

2. What's the probability that you roll a 5 or higher?

$$P(x \geq 5) = 1 - P(x \leq 4) = 1 - 2/3 = 1/3$$

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## Practice Problem

Which of the following are probability functions?

- a.  $f(x) = .25$  for  $x = 9, 10, 11, 12$
- b.  $f(x) = (3-x)/2$  for  $x = 1, 2, 3, 4$
- c.  $f(x) = (x^2 + x + 1)/25$  for  $x = 0, 1, 2, 3$

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## Answer (a)

a.  $f(x) = .25$  for  $x = 9, 10, 11, 12$

$x$	$f(x)$
<b>9</b>	<b>.25</b>
<b>10</b>	<b>.25</b>
<b>11</b>	<b>.25</b>
<b>12</b>	<b><u>.25</u></b>

1.0

Yes, probability function!

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## Answer (b)

b.  $f(x) = (3-x)/2$  for  $x = 1, 2, 3, 4$

$x$	$f(x)$
<b>1</b>	<b><math>(3-1)/2 = 1.0</math></b>
<b>2</b>	<b><math>(3-2)/2 = .5</math></b>
<b>3</b>	<b><math>(3-3)/2 = 0</math></b>
<b>4</b>	<b><math>(3-4)/2 = -.5</math></b>

Though this sums to 1, you can't have a negative probability; therefore, it's not a probability function.

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## Answer (c)

c.  $f(x) = (x^2 + x + 1)/25$  for  $x=0, 1, 2, 3$

<b>x</b>	<b>f(x)</b>
<b>0</b>	<b>1/25</b>
<b>1</b>	<b>3/25</b>
<b>2</b>	<b>7/25</b>
<b>3</b>	<b><u>13/25</u></b>

$24/25$

Doesn't sum to 1. Thus, it's not a probability function.

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## Practice Problem:

- The number of ships to arrive at a harbor on any given day is a random variable represented by  $x$ . The probability distribution for  $x$  is:

<b>x</b>	<b>10</b>	<b>11</b>	<b>12</b>	<b>13</b>	<b>14</b>
<b>P(x)</b>	<b>.4</b>	<b>.2</b>	<b>.2</b>	<b>.1</b>	<b>.1</b>

Find the probability that on a given day:

- exactly 14 ships arrive  $p(x=14) = .1$
- At least 12 ships arrive  $p(x \geq 12) = (.2 + .1 + .1) = .4$
- At most 11 ships arrive  $p(x \leq 11) = (.4 + .2) = .6$

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## Practice Problem:

You are lecturing to a group of 1000 students. You ask them to each randomly pick an integer between 1 and 10. Assuming, their picks are truly random:

- What's your best guess for how many students picked the number 9?

Since  $p(x=9) = 1/10$ , we'd expect about  $1/10^{\text{th}}$  of the 1000 students to pick 9. 100 students.

- What percentage of the students would you expect picked a number less than or equal to 6?

Since  $p(x \leq 6) = 1/10 + 1/10 + 1/10 + 1/10 + 1/10 + 1/10 = .6$   
60%

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## Important discrete distributions in epidemiology...

### ■ Binomial

- Yes/no outcomes (dead/alive, treated/untreated, smoker/non-smoker, sick/well, etc.)

### ■ Poisson

- Counts (e.g., how many cases of disease in a given area)



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## Continuous case

- The probability function that accompanies a continuous random variable is a continuous mathematical function that integrates to 1.
- The probabilities associated with continuous functions are just areas under the curve (integrals!).
- Probabilities are given for a range of values, rather than a particular value (e.g., the probability of getting a math SAT score between 700 and 800 is 2%).

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## Continuous case

- For example, recall the negative exponential function (in probability, this is called an “exponential distribution”):  $f(x) = e^{-x}$

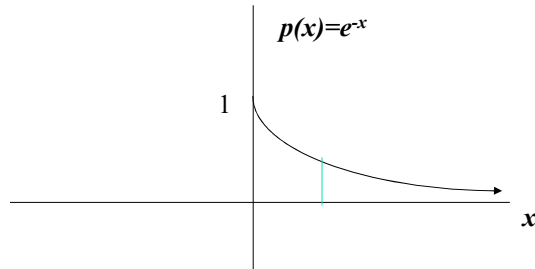
- This function integrates to 1:

$$\int_0^{+\infty} e^{-x} = -e^{-x} \Big|_0^{+\infty} = 0 + 1 = 1$$

$e$  is approximately equal to 2.71828.

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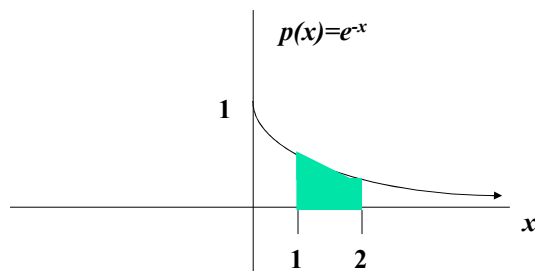
## Continuous case: “probability density function” (pdf)



The probability that  $x$  is any exact particular value (such as 1.9976) is 0; we can only assign probabilities to possible ranges of  $x$ .

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For example, the probability of  $x$  falling within 1 to 2:



$$P(1 \leq x \leq 2) = \int_1^2 e^{-x} = -e^{-x} \Big|_1^2 = -e^{-2} - (-e^{-1}) = -.135 + .368 = .23$$

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## Cumulative distribution function

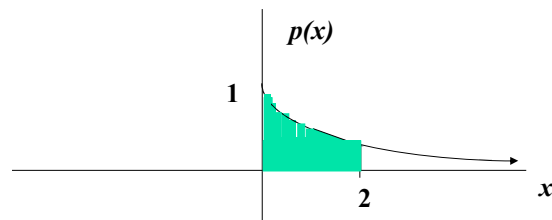
As in the discrete case, we can specify the “cumulative distribution function” (CDF):

The CDF here =  $P(x \leq A) =$

$$\int_0^A e^{-x} = -e^{-x} \Big|_0^A = -e^{-A} - (-e^0) = -e^{-A} + 1 = 1 - e^{-A}$$

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## Example



$$P(x \leq 2) = 1 - e^{-2} = 1 - .135 = .865$$

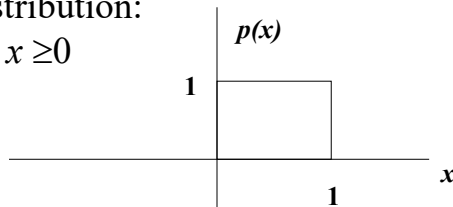
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## Example 2: Uniform distribution

The uniform distribution: all values are equally likely

The uniform distribution:

$$f(x) = 1, \text{ for } 1 \geq x \geq 0$$



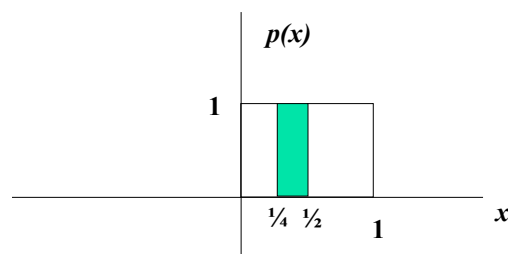
We can see it's a probability distribution because it integrates to 1 (the area under the curve is 1):

$$\int_0^1 1 = x \Big|_0^1 = 1 - 0 = 1$$

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## Example: Uniform distribution

What's the probability that  $x$  is between  $\frac{1}{4}$  and  $\frac{1}{2}$ ?



$$P(\frac{1}{2} \geq x \geq \frac{1}{4}) = \frac{1}{4}$$

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## Practice Problem

4. Suppose that survival drops off rapidly in the year following diagnosis of a certain type of advanced cancer. Suppose that the length of survival (or time-to-death) is a random variable that approximately follows an exponential distribution with parameter 2 (makes it a steeper drop off):

$$\text{probability function: } p(x = T) = 2e^{-2T}$$

$$[\text{note: } \int_0^{+\infty} 2e^{-2x} = -e^{-2x} \Big|_0^{+\infty} = 0 + 1 = 1]$$

**What's the probability that a person who is diagnosed with this illness survives a year?**

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## Answer

The probability of dying within 1 year can be calculated using the cumulative distribution function:

Cumulative distribution function is:

$$P(x \leq T) = -e^{-2x} \Big|_0^T = 1 - e^{-2(T)}$$

The chance of surviving past 1 year is:  $P(x \geq 1) = 1 - P(x \leq 1)$

$$1 - (1 - e^{-2(1)}) = .135$$

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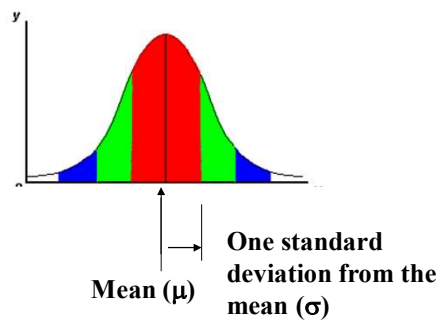
## Expected Value and Variance

- All probability distributions are characterized by an expected value and a variance (standard deviation squared).

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**For example, bell-curve (normal) distribution:**



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## Expected value, or mean

- If we understand the underlying probability function of a certain phenomenon, then we can make informed decisions based on how we expect  $x$  to behave on-average over the long-run...(so called "frequentist" theory of probability).
- Expected value is just the weighted average or mean ( $\mu$ ) of random variable  $x$ . Imagine placing the masses  $p(x)$  at the points  $X$  on a beam; the balance point of the beam is the expected value of  $x$ .

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## Example: expected value

- Recall the following probability distribution of ship arrivals:

<b><math>x</math></b>	<b>10</b>	<b>11</b>	<b>12</b>	<b>13</b>	<b>14</b>
<b><math>P(x)</math></b>	<b>.4</b>	<b>.2</b>	<b>.2</b>	<b>.1</b>	<b>.1</b>



$$\sum_{i=1}^5 x_i p(x) = 10(.4) + 11(.2) + 12(.2) + 13(.1) + 14(.1) = 11.3$$

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# Expected value, formally

**Discrete case:**

$$E(X) = \sum_{\text{all } x} x_i p(x_i)$$

General Formula:

Outcome	$x_1$	$x_2$	$x_3$	$x_4$	...	$x_n$
Probability	$P_1$	$P_2$	$P_3$	$P_4$	...	$P_n$
Expected Value	$= P_1 \cdot x_1 + P_2 \cdot x_2 + P_3 \cdot x_3 + P_4 \cdot x_4 + \dots + P_n \cdot x_n$					

Example: A single fair six-sided die is rolled.

Outcome	1	2	3	4	5	6
Probability	$1/6$	$1/6$	$1/6$	$1/6$	$1/6$	$1/6$
Expected Value	$= 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + 3 \cdot \frac{1}{6} + 4 \cdot \frac{1}{6} + 5 \cdot \frac{1}{6} + 6 \cdot \frac{1}{6} = 3.5$					

**Continuous case:**

$$E(X) = \int_{\text{all } x} x_i p(x_i) dx$$

The symbol  $dx$ , called the differential of the variable  $x$ , indicates that the variable of integration is  $x$ .

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# Empirical Mean is a special case of Expected Value...

Sample mean, for a sample of  $n$  subjects: =

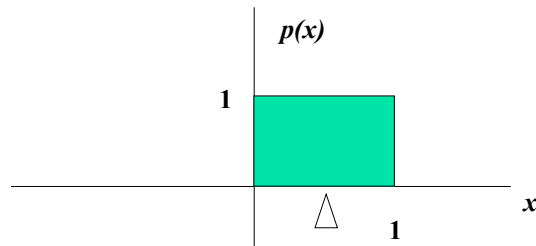
$$\bar{X} = \frac{\sum_{i=1}^n x_i}{n} = \sum_{i=1}^n x_i \left(\frac{1}{n}\right)$$

**The probability (frequency) of each person in the sample is  $1/n$ .**

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## Extension to continuous case: uniform distribution



$$E(X) = \int_0^1 x(1) dx = \frac{x^2}{2} \Big|_0^1 = \frac{1}{2} - 0 = \frac{1}{2}$$

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## Symbol Interlude

- $E(X) = \mu$ 
  - these symbols are used interchangeably
- Expected value is an extremely useful concept for good decision-making!

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## Example: the lottery

- The Lottery (also known as a tax on people who are bad at math...)
- A certain lottery works by picking 6 numbers from 1 to 49. It costs \$1.00 to play the lottery, and if you win, you win \$2 million after taxes.
- *If you play the lottery once, what are your expected winnings or losses?*

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## Lottery

Calculate the probability of winning in 1 try:

$$\frac{1}{\binom{49}{6}} = \frac{1}{49! / (43!6!)} = \frac{1}{13,983,816} = 7.2 \times 10^{-8}$$

“49 choose 6”

Out of 49 numbers, this is the number of distinct combinations of 6.

The probability function (note, sums to 1.0):

x\$	p(x)
-1	.999999928
+ 2 million	7.2 x 10 <sup>-8</sup>

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## Expected Value

### The probability function

$x$	$p(x)$
-1	.999999928
+ 2 million	$7.2 \times 10^{-8}$

### Expected Value

$$E(X) = P(\text{win}) * \$2,000,000 + P(\text{lose}) * -\$1.00 \\ = 2.0 \times 10^6 * 7.2 \times 10^{-8} + .999999928 (-1) = .144 - .999999928 = -$.86$$

Negative expected value is never good!

You shouldn't play if you expect to lose money!

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## Expected Value

**If you play the lottery every week for 10 years, what are your expected winnings or losses?**

$$520 \times (-.86) = -\$447.20$$

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## Gambling (or how casinos can afford to give so many free drinks...)

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A roulette wheel has the numbers 1 through 36, as well as 0 and 00. If you bet \$1 that an odd number comes up, you win or lose \$1 according to whether or not that event occurs. If random variable  $X$  denotes your net gain,  $X=1$  with probability  $18/38$  and  $X=-1$  with probability  $20/38$ .

$$E(X) = 1(18/38) - 1(20/38) = -\$0.053$$

On average, the casino wins (and the player loses) 5 cents per game.

The casino rakes in even more if the stakes are higher:

$$E(X) = 10(18/38) - 10(20/38) = -\$0.53$$

If the cost is \$10 per game, the casino wins an average of 53 cents per game. If 10,000 games are played in a night, that's a cool \$5300.

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## \*\*A few notes about Expected Value as a mathematical operator:

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If  $c =$  a constant number (i.e., not a variable) and  $X$  and  $Y$  are any random variables...

- $E(c) = c$
- $E(cX) = cE(X)$
- $E(c + X) = c + E(X)$
- $E(X + Y) = E(X) + E(Y)$


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$$E(c) = c$$

---

$E(c) = c$   
Example: If you cash in soda cans in CA, you always get 5 cents per can.  
Therefore, there's no randomness. You always expect to (and do) get 5 cents.

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$$E(cX) = cE(X)$$

---

$E(cX) = cE(X)$   
Example: If the casino charges \$10 per game instead of \$1, then the casino expects to make 10 times as much on average from the game (See roulette example above!)

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$$E(c + X) = c + E(X)$$

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$$E(c + X) = c + E(X)$$

Example, if the casino throws in a free drink worth exactly \$5.00 every time you play a game, you always expect to (and do) gain an extra \$5.00 regardless of the outcome of the game.

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$$E(X + Y) = E(X) + E(Y)$$

---

$$E(X + Y) = E(X) + E(Y)$$

Example: If you play the lottery twice, you expect to lose:  $-\$.86 + -\$.86$ .

**NOTE: This works even if X and Y are dependent!! Does not require independence!! Proof left for later...**

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## Practice Problem

If a disease is fairly rare and the antibody test is fairly expensive, in a resource-poor region, one strategy is to take half of the serum from each sample and pool it with  $n$  other halved samples, and test the pooled lot. If the pooled lot is negative, this saves  $n-1$  tests. If it's positive, then you go back and test each sample individually, requiring  $n+1$  tests total.

- Suppose a particular disease has a prevalence of 10% in a third-world population and you have 500 blood samples to screen. If you pool 20 samples at a time (25 lots), how many tests do you expect to have to run (assuming the test is perfect!)?
- What if you pool only 10 samples at a time?
- 5 samples at a time?

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## Answer (a)

- Suppose a particular disease has a prevalence of 10% in a third-world population and you have 500 blood samples to screen. If you pool 20 samples at a time (25 lots), how many tests do you expect to have to run (assuming the test is perfect!)?

Let  $X$  = a random variable that is the number of tests you have to run per lot:

$$E(X) = P(\text{pooled lot is negative})(1) + P(\text{pooled lot is positive})(21)$$

$$E(X) = (.90)^{20}(1) + [1 - (.90)^{20}](21) = 12.2\%(1) + 87.8\%(21) = 18.56$$

$$E(\text{total number of tests}) = 25 * 18.56 = 464$$

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## Answer (b)

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b. What if you pool only 10 samples at a time?

$$E(X) = (.90)^{10} (1) + [1 - .90^{10}] (11) = 35\% (1) + 65\% (11) = 7.5$$

average per lot

$$50 \text{ lots} * 7.5 = 375$$

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## Answer (c)

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c. 5 samples at a time?

$$E(X) = (.90)^5 (1) + [1 - .90^5] (6) = 59\% (1) + 41\% (6) = 3.05$$

average per lot

$$100 \text{ lots} * 3.05 = 305$$

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## Practice Problem

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If  $X$  is a random integer between 1 and 10, what's the expected value of  $X$ ?

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## Answer

---

If  $X$  is a random integer between 1 and 10, what's the expected value of  $X$ ?

$$\mu = E(x) = \sum_{i=1}^{10} i \left(\frac{1}{10}\right) = \frac{1}{10} \sum_{i=1}^{10} i = (.1) \frac{10(10+1)}{2} = 55(.1) = 5.5$$

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## Expected value isn't everything though...

- Take the show "Deal or No Deal"
- Everyone know the rules?
- Let's say you are down to two cases left. \$1 and \$400,000. The banker offers you \$200,000.
- So, Deal or No Deal?

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## Deal or No Deal...

- This could really be represented as a probability distribution and a non-random variable:

$x$ \$	$p(x)$
+1	.50
+\$400,000	.50

$x$ \$	$p(x)$
+\$200,000	1.0

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## Expected value doesn't help...

$x$	$p(x)$
+1	.50
+\$400,000	.50

$$\mu = E(X) = \sum_{\text{all } x} x_i p(x_i) = +1(.50) + 400,000(.50) = 200,000$$

$x$	$p(x)$
+\$200,000	1.0

$$\mu = E(X) = 200,000$$

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## How to decide?

### Variance!

- If you take the deal, the variance/standard deviation is 0.
- If you don't take the deal, what is average deviation from the mean?
- What's your gut guess?

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## Variance/standard deviation

“The average (expected) squared distance (or deviation) from the mean”

$$\sigma^2 = \text{Var}(x) = E[(x - \mu)^2] = \sum_{\text{all } x} (x_i - \mu)^2 p(x_i)$$

*\*\*We square because squaring has better properties than absolute value. Take square root to get back linear average distance from the mean (=“standard deviation”).*

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## Variance, formally

**Discrete case:**

$$\text{Var}(X) = \sigma^2 = \sum_{\text{all } x} (x_i - \mu)^2 p(x_i)$$

**Continuous case:**

$$\text{Var}(X) = \sigma^2 = \int_{-\infty}^{\infty} (x_i - \mu)^2 p(x_i) dx$$

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## Similarity to empirical variance

The variance of a sample:  $s^2 =$

$$\frac{\sum_{i=1}^N (x_i - \bar{x})^2}{n-1} = \sum_{i=1}^N (x_i - \bar{x})^2 \left(\frac{1}{n-1}\right)$$

Division by n-1 reflects the fact that we have lost a "degree of freedom" (piece of information) because we had to estimate the sample mean before we could estimate the sample variance.

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## Symbol Interlude

- $\text{Var}(X) = \sigma^2$ 
  - these symbols are used interchangeably

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## Variance: Deal or No Deal

$$\sigma^2 = \sum_{\text{all } x} (x_i - \mu)^2 p(x_i)$$

$$\begin{aligned}\sigma^2 &= \sum_{\text{all } x} (x_i - \mu)^2 p(x_i) = \\ &= (1 - 200,000)^2 (.5) + (400,000 - 200,000)^2 (.5) = 200,000^2 \\ \sigma &= \sqrt{200,000^2} = 200,000\end{aligned}$$

**Now you examine your personal risk tolerance...**

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## Practice Problem

A roulette wheel has the numbers 1 through 36, as well as 0 and 00. If you bet \$1.00 that an odd number comes up, you win or lose \$1.00 according to whether or not that event occurs. If  $X$  denotes your net gain,  $X=1$  with probability  $18/38$  and  $X=-1$  with probability  $20/38$ .

We already calculated the mean to be =  $-\$.053$ .  
What's the variance of  $X$ ?

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## Answer

$$\begin{aligned}\sigma^2 &= \sum_{\text{all } x} (x_i - \mu)^2 p(x_i) \\ &= (+1 - -.053)^2(18/38) + (-1 - -.053)^2(20/38) \\ &= (1.053)^2(18/38) + (-1 + .053)^2(20/38) \\ &= (1.053)^2(18/38) + (-.947)^2(20/38) \\ &= .997 \\ \sigma &= \sqrt{.997} = .99\end{aligned}$$

Standard deviation is \$.99. Interpretation: On average, you're either 1 dollar above or 1 dollar below the mean, which is just under zero. Makes sense!

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## Handy calculation formula!

**Handy calculation formula (if you ever need to calculate by hand!):**

$$\text{Var}(X) = \sum_{\text{all } x} (x_i - \mu)^2 p(x_i) = \sum_{\text{all } x} x_i^2 p(x_i) - (\mu)^2$$

Intervening algebra!  $= E(x^2) - [E(x)]^2$

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## Var(x) = E(x-μ)<sup>2</sup> = E(x<sup>2</sup>) - [E(x)]<sup>2</sup> (your calculation formula!)

### Proofs (optional!):

$$\begin{aligned}
 E(x-\mu)^2 &= E(x^2 - 2\mu x + \mu^2) \\
 &= E(x^2) - E(2\mu x) + E(\mu^2) \\
 &= E(x^2) - 2\mu E(x) + \mu^2 \\
 &= E(x^2) - 2\mu\mu + \mu^2 \\
 &= E(x^2) - \mu^2 \\
 &= E(x^2) - [E(x)]^2
 \end{aligned}$$

remember "FOIL"?!

Use rules of expected value: E(X+Y) = E(X) + E(Y)

E(c) = c

E(x) = μ

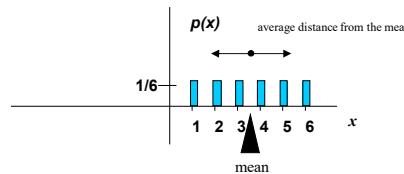
### OR, equivalently:

$$\begin{aligned}
 E(x-\mu)^2 &= \sum_{all\ x} [(x-\mu)^2] p(x) = \sum_{all\ x} [x^2 - 2\mu x + \mu^2] p(x) = \sum_{all\ x} x^2 p(x) - 2\mu \sum_{all\ x} x p(x) + \mu^2 \sum_{all\ x} p(x) = E(x^2) - 2\mu E(x) + \mu^2 (1) = \\
 &E(x^2) - 2\mu^2 + \mu^2 (1) = E(x^2) - \mu^2
 \end{aligned}$$

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## For example, what's the variance and standard deviation of the roll of a die?

x	p(x)
1	p(x=1)=1/6
2	p(x=2)=1/6
3	p(x=3)=1/6
4	p(x=4)=1/6
5	p(x=5)=1/6
6	p(x=6)=1/6
	1.0



$$E(x) = \sum_{all\ x} x_i p(x_i) = (1)\left(\frac{1}{6}\right) + 2\left(\frac{1}{6}\right) + 3\left(\frac{1}{6}\right) + 4\left(\frac{1}{6}\right) + 5\left(\frac{1}{6}\right) + 6\left(\frac{1}{6}\right) = \frac{21}{6} = 3.5$$

$$E(x^2) = \sum_{all\ x} x_i^2 p(x_i) = (1)\left(\frac{1}{6}\right) + 4\left(\frac{1}{6}\right) + 9\left(\frac{1}{6}\right) + 16\left(\frac{1}{6}\right) + 25\left(\frac{1}{6}\right) + 36\left(\frac{1}{6}\right) = 15.17$$

$$\sigma_x^2 = Var(x) = E(x^2) - [E(x)]^2 = 15.17 - 3.5^2 = 2.92$$

$$\sigma_x = \sqrt{2.92} = 1.71$$

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**\*\*A few notes about Variance as a mathematical operator:**

---

If  $c$  = a constant number (i.e., not a variable) and  $X$  and  $Y$  are random variables, then

- $\text{Var}(c) = 0$
- $\text{Var}(c+X) = \text{Var}(X)$
- $\text{Var}(cX) = c^2\text{Var}(X)$
- $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$  **ONLY IF  $X$  and  $Y$  are independent!!!!**
- $\{\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X,Y)$  **IF  $X$  and  $Y$  are not independent}**

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**$\text{Var}(c) = 0$**

---

$\text{Var}(c) = 0$

Constants don't vary!

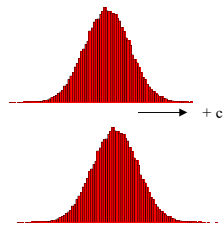
66




## $\text{Var}(c+X) = \text{Var}(X)$

$$\text{Var}(c+X) = \text{Var}(X)$$

Adding a constant to every instance of a random variable doesn't change the variability. It just shifts the whole distribution by  $c$ . If everybody grew 5 inches suddenly, the variability in the population would still be the same.



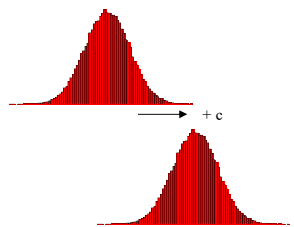
67




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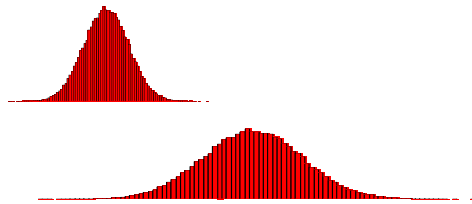


68



$$\text{Var}(cX) = c^2 \text{Var}(X)$$

$$\text{Var}(cX) = c^2 \text{Var}(X)$$

Multiplying each instance of the random variable by  $c$  makes it  $c$ -times as wide of a distribution, which corresponds to  $c^2$  as much variance (deviation squared). For example, if everyone suddenly became twice as tall, there'd be twice the deviation and 4 times the variance in heights in the population.



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$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$$

$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$  **ONLY IF X and Y are independent!!!!!!**

With two random variables, you have more opportunity for variation, unless they vary together (are dependent, or have covariance):  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)$

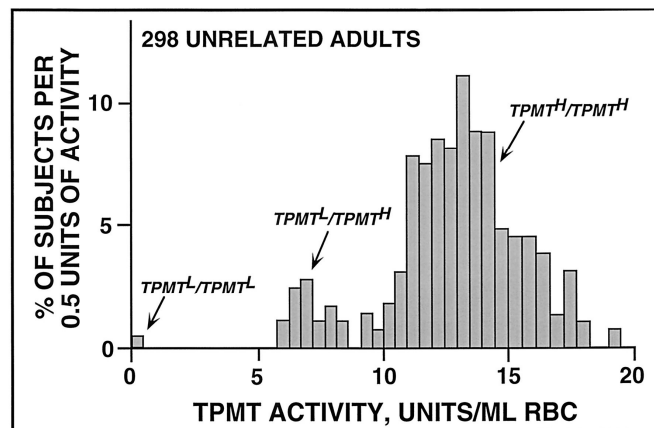
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## Example of $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$ : TPMT

- TPMT metabolizes the drugs 6-mercaptopurine, azathioprine, and 6-thioguanine (chemotherapy drugs)
- People with  $\text{TPMT}^{-}/\text{TPMT}^{+}$  have reduced levels of activity (10% prevalence)
- People with  $\text{TPMT}^{-}/\text{TPMT}^{-}$  have no TPMT activity (prevalence 0.3%).
- They cannot metabolize 6-mercaptopurine, azathioprine, and 6-thioguanine, and risk bone marrow toxicity if given these drugs.

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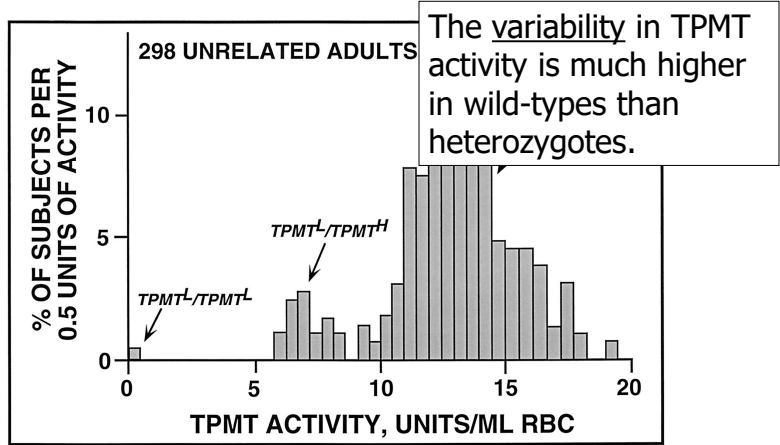
## TPMT activity by genotype



Weinshilboun R. Drug Metab Dispos. 2001 Apr;29(4 Pt 2):601-5

72

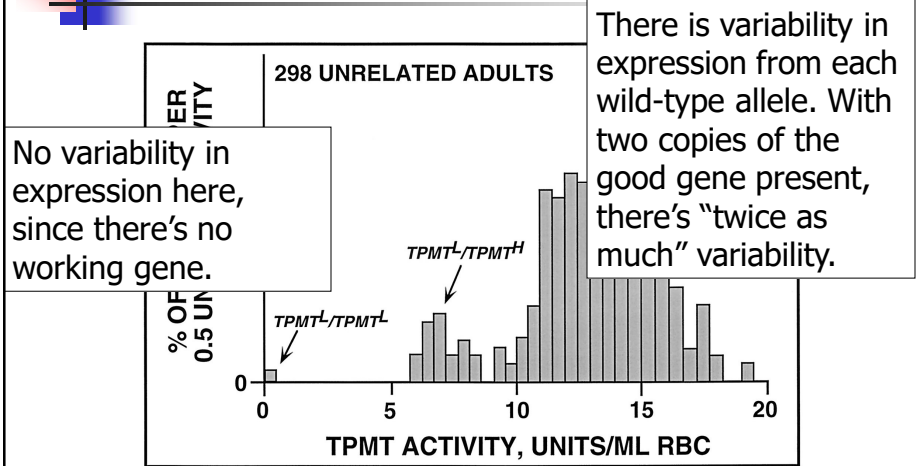
## TPMT activity by genotype



Weinshilboum R. Drug Metab Dispos. 2001 Apr;29(4 Pt 2):601-5

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## TPMT activity by genotype



Weinshilboum R. Drug Metab Dispos. 2001 Apr;29(4 Pt 2):601-5

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## Practice Problem

Find the variance and standard deviation for the number of ships to arrive at the harbor (recall that the mean is 11.3).

<b>x</b>	<b>10</b>	<b>11</b>	<b>12</b>	<b>13</b>	<b>14</b>
<b>P(x)</b>	<b>.4</b>	<b>.2</b>	<b>.2</b>	<b>.1</b>	<b>.1</b>

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## Answer: variance and std dev

<b>x<sup>2</sup></b>	<b>100</b>	<b>121</b>	<b>144</b>	<b>169</b>	<b>196</b>
<b>P(x)</b>	<b>.4</b>	<b>.2</b>	<b>.2</b>	<b>.1</b>	<b>.1</b>

$$E(x^2) = \sum_{i=1}^5 x_i^2 p(x_i) = (100)(.4) + (121)(.2) + 144(.2) + 169(.1) + 196(.1) = 129.5$$

$$Var(x) = E(x^2) - [E(x)]^2 = 129.5 - 11.3^2 = 1.81$$

$$stddev(x) = \sqrt{1.81} = 1.35$$

**Interpretation: On an average day, we expect 11.3 ships to arrive in the harbor, plus or minus 1.35. This gives you a feel for what would be considered a usual day!**

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## Practice Problem

You toss a coin 100 times. What's the expected number of heads? What's the variance of the number of heads?

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## Answer: expected value

Intuitively, we'd probably all agree that we expect around 50 heads, right?

Another way to show this→

Think of tossing 1 coin.  $E(X=\text{number of heads}) = (1)P(\text{heads}) + (0)P(\text{tails})$

$$\therefore E(X=\text{number of heads}) = 1(.5) + 0 = .5$$

If we do this 100 times, we're looking for the sum of 100 tosses, where we assign 1 for a heads and 0 for a tails. (these are 100 "independent, identically distributed (i.i.d)" events)

$$\begin{aligned} E(X_1 + X_2 + X_3 + X_4 + X_5 \dots + X_{100}) &= E(X_1) + E(X_2) + E(X_3) + E(X_4) + E(X_5) \dots + \\ E(X_{100}) &= \\ 100 E(X_1) &= 50 \end{aligned}$$

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## Answer: variance

What's the variability, though? More tricky. But, again, we could do this for 1 coin and then use our rules of variance.

Think of tossing 1 coin.

$$E(X^2 = \text{number of heads squared}) = 1^2 P(\text{heads}) + 0^2 P(\text{tails})$$

$$\therefore E(X^2) = 1(.5) + 0 = .5$$

$$\text{Var}(X) = .5 - .5^2 = .5 - .25 = .25$$

Then, using our rule:  $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$  (coin tosses are independent!)

$$\text{Var}(X_1 + X_2 + X_3 + X_4 + X_5 + \dots + X_{100}) = \text{Var}(X_1) + \text{Var}(X_2) + \text{Var}(X_3) + \text{Var}(X_4) + \text{Var}(X_5) + \dots + \text{Var}(X_{100}) =$$

$$100 \text{Var}(X_1) = 100 (.25) = 25$$

$$\text{SD}(X) = 5$$

Interpretation: When we toss a coin 100 times, we expect to get 50 heads plus or minus 5.

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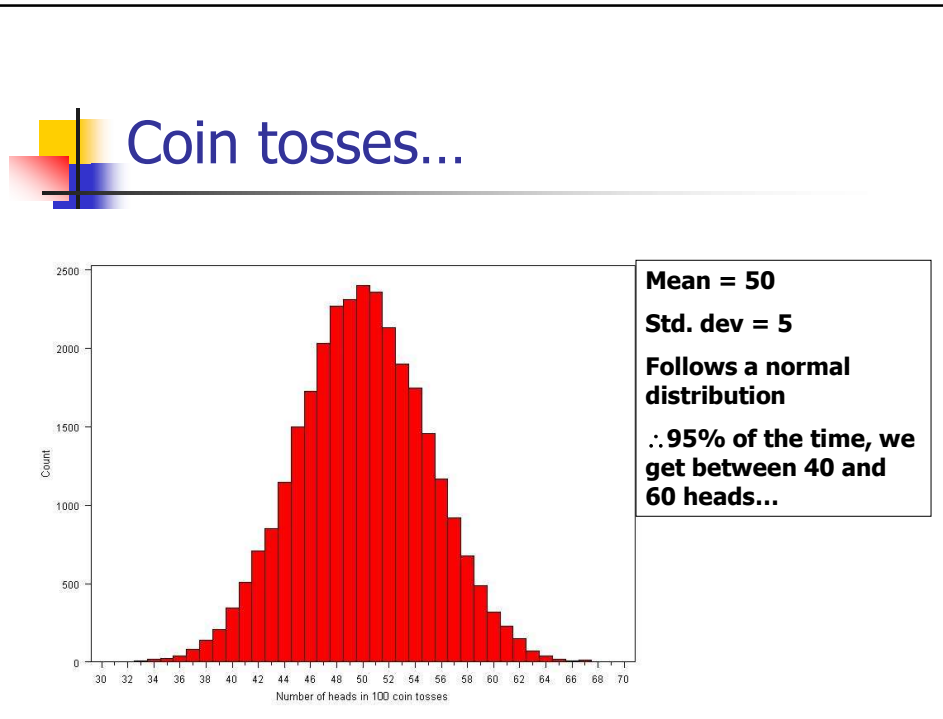


## Or use computer simulation...

- Flip coins virtually!
  - Flip a virtual coin 100 times; count the number of heads.
  - Repeat this over and over again a large number of times (we'll try 30,000 repeats!)
  - Plot the 30,000 results.

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## Covariance: joint probability

- The covariance measures the strength of the linear relationship between two variables
- The covariance:  $E[(x - \mu_x)(y - \mu_y)]$

$$\sigma_{xy} = \sum_{i=1}^N (x_i - \mu_x)(y_i - \mu_y) P(x_i, y_i)$$

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## The Sample Covariance

- The sample covariance:

$$\text{cov}(x, y) = \frac{\sum_{i=1}^n (x_i - \bar{X})(y_i - \bar{Y})}{n-1}$$

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## Interpreting Covariance

- **Covariance** between two random variables:

$\text{cov}(X, Y) > 0 \rightarrow$  X and Y are positively correlated

$\text{cov}(X, Y) < 0 \rightarrow$  X and Y are inversely correlated

$\text{cov}(X, Y) = 0 \rightarrow$  X and Y are independent

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## Covariance example

	S&P 500	ABC Corp.
2013	1,692	68
2014	1,978	102
2015	1,884	110
2016	2,151	112
2017	2,519	154

$$\text{Mean (S\&P 500)} = \frac{1,692 + 1,978 + 1,884 + 2,151 + 2,519}{5} = 2,044.80$$

$$\text{Mean (ABC Corp.)} = \frac{68 + 102 + 110 + 112 + 154}{5} = 109.20$$

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## Covariance example

	S&P 500	ABC Corp.	Step 3		Step 4
			a	b	a x b
2013	1,692	68	-352.80	-41.20	14,535.36
2014	1,978	102	-66.80	-7.20	480.96
2015	1,884	110	-160.80	0.80	-128.64
2016	2,151	112	106.20	2.80	297.36
2017	2,519	154	474.20	44.80	21,244.16
Mean	2,044.80	109.20	Sum		36,429.20

$$\text{Cov(S\&P 500, ABC Corp.)} = \frac{36,429.20}{5 - 1} = 9,107.30$$

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