Finite Math

25 April 2019



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$$S_3 = \{M, D\}$$

as a sample space where *M* stands for match and *D* stands for do not match.

### Now You Try It!

#### Example

An experiment consists of recording the boy-girl composition of a three-child family. What would be an appropriate sample space if:

- (a) we are interested in the genders of the children in the order of their births? (A tree diagram can help.)
- (b) we are interested only in the number of girls in family?
- (c) we are interested only in which gender there are more of?
- (d) we are interested in all three items from (a)-(c)?

### Now You Try It!

#### Solution

(a)

$$S_1 = \{GGG, GGB, GBG, BGG, GBB, BGB, BBG, BBB\}$$

(b)

$$S_2 = \{0, 1, 2, 3\}$$

(c)

$$S_3 = \{G, B\}$$

(d) Use  $S_1$ .

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### **Dice Rolling Outcomes**

#### Example

Suppose we run an experiment by rolling two dice. What is the most fundamental sample space for this experiment? Give the event for each of the following outcomes. Which are simple events?

- (a) A sum of 7 turns up.
- (b) A sum of 11 turns up.
- (c) A sum less than 4 turns up.
- (d) A sum of 12 turns up.
- (e) A sum of 5 turns up.
- (f) A sum which is prime turns up.
- (g) A sum of 2 turns up.

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#### **Definition (Probabilities of Simple Events)**

Given a sample space

$$S = \{e_1, e_2, ..., e_n\}$$

with n simple events, to each simple event  $e_i$  we assign a real number, denoted by  $P(e_i)$ , called the probability of event  $e_i$ . These numbers can be assigned in an arbitrary manner as long as the following two conditions are satisfied:

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Any probability assignment that satisfies Conditions 1 and 2 is said to be an acceptable probability assignment.

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$$P(H) = \frac{1}{2}$$
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Notice that this satisfies the two conditions of the previous definition. Is this a reasonable assignment of probabilities? Ostensibly yes since there are only two outcomes of a coin flip and there is no reason to doubt that the coin is *fair*, i.e., the two outcomes are equally likely.

Suppose we flipped the coin 1000 times and we turn up with 373 heads and 627 tails.

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since that reflects the actual results of a large collection of outcomes. Another assignment we could technically make is

$$P(H) = 1$$
  $P(T) = 0$ .

While this does fit the rules of an acceptable probability assignment, it is not reasonable in this case, unless the coin had two heads.

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since 
$$P(H) + P(T) = 1.4 > 1$$
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Given an acceptable probability assignment for the simple events in a sample space S, we define the probability of an arbitrary event E, denoted P(E), as follows:

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- (b) If E is a simple event, then P(E) has already been assigned.
- (c) If E is a compound event, then P(E) is the sum of the probabilities of all the simple events in E.
- (d) If E is the sample space S, then P(E) = P(S) = 1 (this is a special case of part (c).)

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### Example

#### Example

Refer back to the example where we roll two dice. If we assume that every simple event is equally likely:

- (a) What is the probability of a simple event happening?
- (b) What are the possible numbers that the two dice could add up to?
- (c) What are the probability of each of the events in part (b) happening?

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When we were talking about assigning probabilities of 0.5 to heads and 0.5 to tails for flipping a coin, and a probability of  $\frac{1}{6}$  for any number to come up when rolling a 6-sided die, we are making an assumption on the probabilities of the experiment called an *equally likely assumption*.

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$$S = \{\textbf{e}_1, \textbf{e}_2, ..., \textbf{e}_n\},$$

we assign to each  $e_i$  a probability of  $\frac{1}{n}$  since there are n possible outcomes and we want each of them to be equally likely. This gives us the following theorem...

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#### Theorem (Probability of an Arbitrary Event under an Equally Likely Assumption)

If we assume that each simple event in a sample space S is equally likely to occur, then the probability of an arbitrary event E in S is given by

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We saw this theorem in action when we found the theoretical probabilities for rolling a number on a pair of dice.

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$$P(H) = \frac{373}{1000}$$
  $P(T) = \frac{623}{1000}$ 

since it reflects the results of an extensive experiment.

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Probability

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