

Probability - II

Addition rules

Addition rules deal with unions (\cup) of events. These can be thought of OR probabilities.

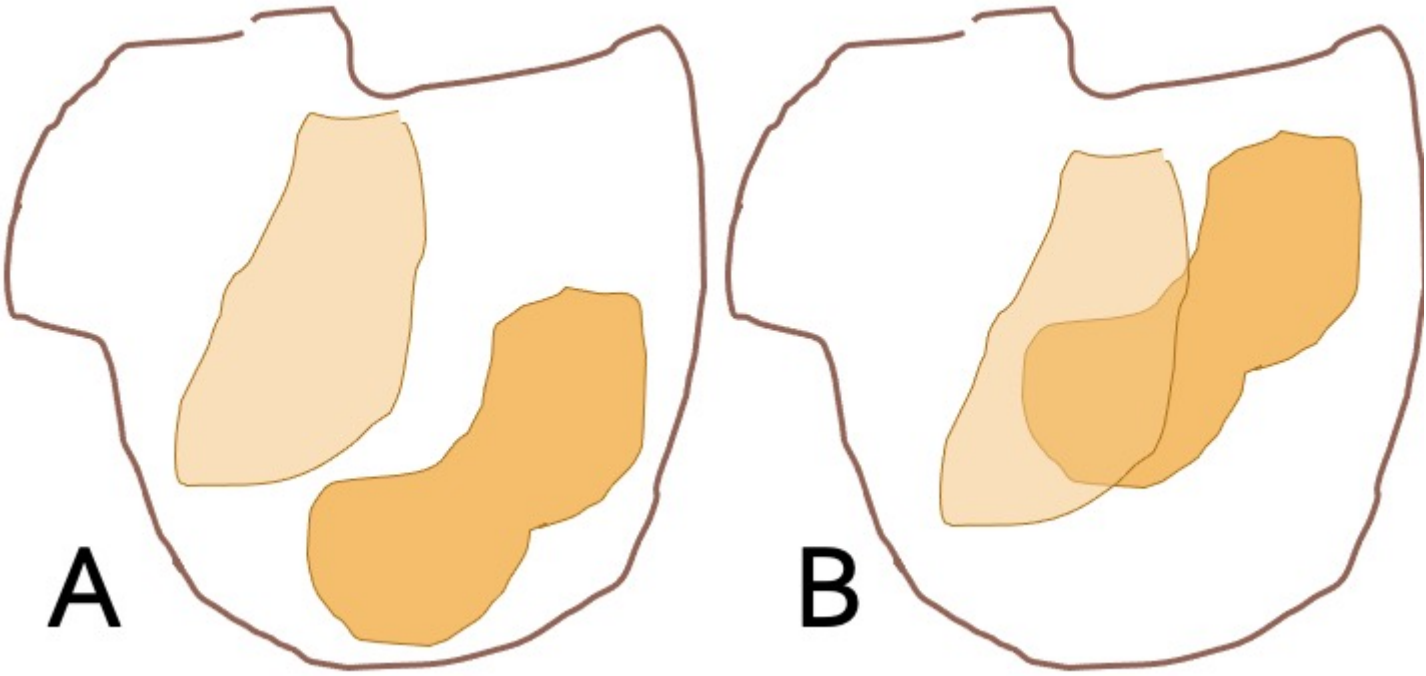
(\cup - non-repeating superset, \cap - common elements)

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

or when A and B are mutually exclusive

$$P(A \cup B) = P(A) + P(B)$$

In the figure below, A represents a case where events in the two shaded regions are mutually exclusive, ie there are no overlapping areas. In B, the overlapping area is the intersection (\cap) of the two areas.



Multiplication rules

Multiplication rules deal with intersections (\cap) of events. These can be thought of AND probabilities.

(\cup - non-repeating superset, \cap - common elements)

Dependent

$$P(A \cap B) = P(A) * P(B|A)$$

Independent

$$P(A \cap B) = P(A) * P(B)$$

Mutually exclusive

$$P(A \cap B) = 0$$

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Conditional, independent, disjoint, exchangeable

Conditional

$P(A|B) = \frac{P(A,B)}{P(B)}$

Independent

$P(A, B) = P(A) * P(B)$

Disjoint

$P(A, B) = 0$

Exchangeable

$P(AthenB) = P(BthenA)$

Factoring joint probabilities

Given a joint probability, we often need to simplify to get to some tractable or known distribution. We can do this via the rules given above and the chain rule.

$$P(A, B) = P(A|B) * P(B)$$

$$P(A, B, C) = P(A|B, C) * P(B, C) = P(A|B, C) * P(B|C) * P(C)$$

Random variable

Random variables are (often) real-valued functions mapping outcomes to a measureable space. In set notation, this looks like:

$$X : \Omega \rightarrow \mathbb{R}$$

This mapping defines the probability giving by X in a measureable set as:

$$P(X \in S) = P(x \in \Omega | X(x) \in S)$$

Random sample, experiment, trial

Random sample = unbiased realization of outcomes

Random experiment = an activity with an observable result

Trial = 1 or more experiments

In the figure and table below, we are plotting and tabulating the results from the roll of two dice. We did not define the experiment up front, but will do so now for clarity.

Experiment = throw 2 dice 500 times

Trial = in this case, there was a single trial of the experiment

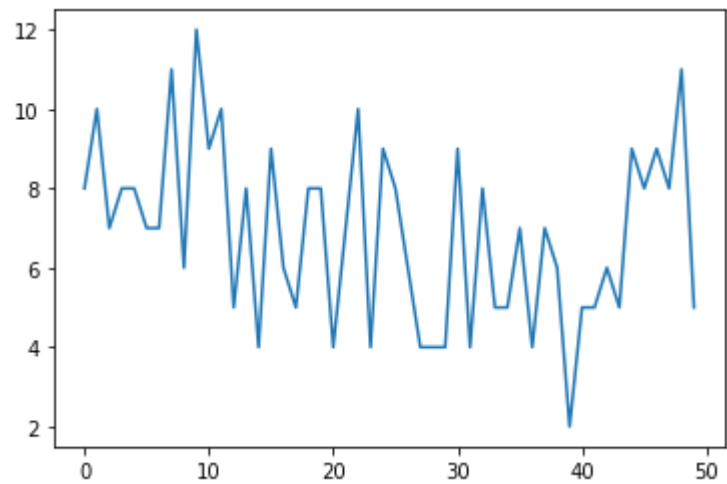
Random sample = we have 500 realizations of two dice being thrown

Note, due to how random numbers are generated and used to get the die rolls, we sampled with replacement into a long 1000 long list. We then pulled the list into two sequences with every other 1 being die1 vs die2. We could have done this in a for loop rolling both dice in each iteration. Based on how randon numbers are generated, this would have resulted in the same sequence of die rolls.

```
## Look at the first 10 dice throws
print(df.head(n=10))
```

	die1	die2	dots
0	6	2	8
1	6	4	10
2	1	6	7
3	5	3	8
4	3	5	8
5	5	2	7
6	1	6	7
7	6	5	11
8	2	4	6
9	6	6	12

```
# view the first 50 throws
tplot = df['dots'].head(n=50).plot()
```



Data generating mechanism

This is the actual process we are trying to model. In the case of the dice, it was the throwing and counting of dots where we are lookign to learn (and use) a statistical model of outcome probabilities. In a count toss experiment, it is the binary outcome of heads or tails.

GRADED EVALUATION (15 mins)

1. Two events A and B are independent, their joint probability is given as:
- a. $P(A, B) = P(A) * P(B)$
- b. $P(A, B) = 0$
1. Given both a fair coin and 6-side die, we know $P(\text{coin}) = 1/2$ and $P(\text{die}) = 1/6$ for any landing of the coin and die. What is the joint probability for any of the possible combinations of outcomes?
- a. $P(A, B) = P(A) * P(B) = \frac{1}{12}$
- b. $P(A, B) = P(A) + P(B) - P(A \text{ or } B) = \frac{1}{3}$
1. In a certain game, you are given $P(A, B) = 1/12$, $P(A) = 1/2$ and $P(B) = \frac{1}{4}$. Are the events independent?
- a. no
- b. yes
1. For the last question, what is $P(A|B)$?

a. $P(A|B) = P(A, B)/P(B)$

b. 0

1. Two events are said to be exchangeable if

a. The order of occurrence doesn't matter.

b. They are independent.

1. A random variable is:

a. A real valued function mapping outcomes to the real number line.

b. A real valued function mapping occurrence of events to the real interval $[0,1]$.

1. The marginal distribution of X for $f(x, y) = \frac{1}{4}xy$, $x, y \in [0, 2]$ is

a. $\frac{1}{2}x$

b. $\frac{1}{2}y$

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