

M378K: January 17th, 2025.

M378K Introduction to Mathematical Statistics

Problem Set #2

Discrete random variables.

2.1. Probability mass function. Recall the following definition from the last class:

Definition 2.1. Given a set B , we say that a random variable Y is B -valued if

$$\mathbb{P}[Y \in B] = 1.$$

We reserve special terminology for random variables Y depending on the cardinality of the set B from the above definition. In particular, we have the following definition:

Definition 2.2. A random variable Y is said to be discrete if there exists a set S such that :

- Y is S -valued, and
- S is either finite or countable.

Problem 2.1. Provide an example of a discrete random variable.

- geometric.
- coin toss. $S = \{H, T\}$
- roll of a die : $S = \{1, 2, 3, 4, 5, 6\}$.

Our next task is to try to keep track of the probabilities that Y takes specific values from S . In order to be more "economical", we introduce the following concept:

Definition 2.3. The support S_Y of a random variable Y is the smallest set S such that Y is S -valued.

Problem 2.2. What is the support of the random variable you provided as an example in the above problem?

roll of a die: $\{1, \dots, 6\}$

Problem 2.3. Let $y \in S_Y$. Is it possible to have $\mathbb{P}[Y = y] = 0$?

No Since Y is discrete, we could have
 $\tilde{S}_Y = S_Y - \{y\}$
and Y would be \tilde{S}_Y -valued
and "smaller".

Usually, we are interested in calculating and modeling probabilities that look like this

$$\mathbb{P}[Y \in A] \text{ for some } A \subset S_Y.$$

Note that, if we know the probabilities of the form

$$\mathbb{P}[Y = y] \text{ for all } y \in S_Y,$$

then we can calculate any probability of the above form. *How?*

$$\mathbb{P}[Y \in A] = \sum_{y \in A} \mathbb{P}[Y = y]$$

So, if we "tabulate" the probabilities of the form $\mathbb{P}[Y = y]$ for all $y \in S_Y$, we have sufficient information to calculate any probability of interest to do with the random variable Y . This observation motivates the following definition:

Definition 2.4. The probability mass function (pmf) of a **discrete** random variable Y is the function $p_Y : S_Y \rightarrow \mathbb{R}$ defined as

$$p_Y(y) = \mathbb{P}[Y = y] \text{ for all } y \in S_Y.$$

Can you think of different ways in which to display the pmf?

- A distribution table

y	y_1	y_2	\dots	y_k	\dots
$p_Y(y)$	p_1	p_2	\dots	p_k	\dots

- roll of a die: $p_Y(k) = \frac{1}{6}, \quad k=1, \dots, 6$

What are the immediate properties of every pmf? Does the "reverse" hold, i.e., if a function p_Y satisfies you stated, is it always a pmf of **some** random variable?

$$\left\{ \begin{array}{l} \bullet p_Y(y) > 0 \text{ for all } y \in S_Y \\ \bullet \sum_{y \in S_Y} p_Y(y) = 1 \end{array} \right.$$

Yes!

What is the pmf of the random variable which you provided as an example above?

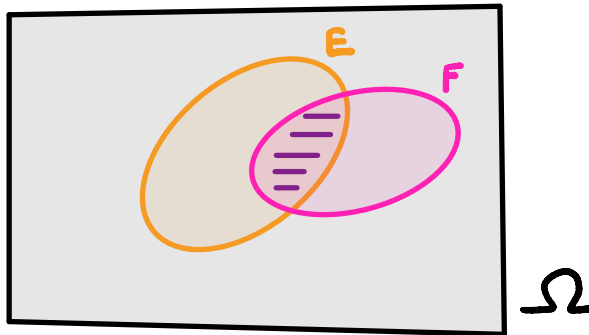
- roll of a die: $p_Y(k) = \frac{1}{6}, \quad k=1, \dots, 6$

2.2. Conditional probability. In order to "build" more complicated (and useful!) random variables, it helps to review a bit more probability.

Definition 2.5. Let E and F be two events on the same Ω such that $\mathbb{P}[E] > 0$. The conditional probability of F given E is defined as

$$\mathbb{P}[F|E] = \frac{\mathbb{P}[E \cap F]}{\mathbb{P}[E]}.$$

Let's spend a moment with the geometric/informational perspective on this definition.



By far, the most popular problems relying on the notion of **conditional probability** are those to do with **specificity** and **sensitivity**¹ of medical tests.

Problem 2.4. At any given time, 2% of the population actually has a particular disease✓

A test indicates the presence of a particular disease 96% of the time in people who actually have the disease. The same test is positive 1% of the time when actually healthy people are tested.

Calculate the probability that a particular person actually has the disease **given** that they tested positive.

→:

E ... the test was positive
 F ... the person has the disease

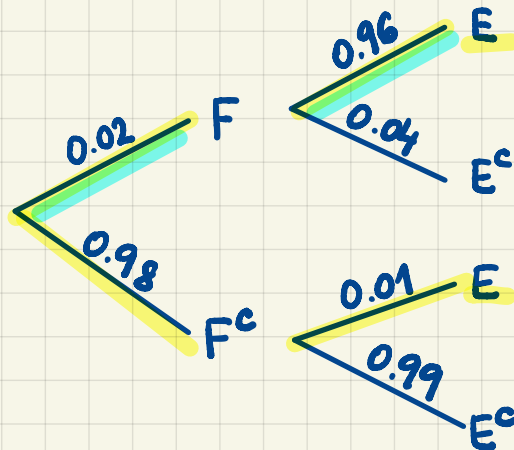
$\mathbb{P}[F] = 0.02$
 $\mathbb{P}[E|F] = 0.96$
 $\mathbb{P}[E|F^c] = 0.01$

$\mathbb{P}[F|E] = ?$

¹https://en.wikipedia.org/wiki/Sensitivity_and_specificity

$$\begin{aligned}
 \mathbb{P}[F|E] &= \frac{\mathbb{P}[F \cap E]}{\mathbb{P}[E]} = \frac{\mathbb{P}[E|F] \cdot \mathbb{P}[F]}{\mathbb{P}[E|F] \cdot \mathbb{P}[F] + \mathbb{P}[E|F^c] \cdot \mathbb{P}[F^c]} \\
 &= \frac{\mathbb{P}[E|F] \cdot \mathbb{P}[F]}{\mathbb{P}[E|F] \cdot \mathbb{P}[F] + \mathbb{P}[E|F^c] \cdot \mathbb{P}[F^c]}
 \end{aligned}$$

Bayes Theorem



$$\begin{aligned}
 \mathbb{P}[F|E] &= \frac{0.02 \cdot 0.96}{0.02 \cdot 0.96 + 0.98 \cdot 0.01} = \frac{2.96}{2.96 + 98} \\
 &= \frac{96}{96 + 49} = \frac{96}{145}
 \end{aligned}$$

