## Lecture Summary

3.1 Discrete Random Variables
3.4 Bivariate Distributions
3.5 Marginal Distributions
3.6 Conditional Distributions
4.1 Expectations

Some slides from MIT open courseware

## Binomial Distribution

Now imagine you toss the fair coin 5 times. Some possible outcomes are 00000, 1000, 01100 etc. Let $X$ be the number of successes (heads) after the 5 times. Then $X$ follows a Binomial distribution with parameters $(5,0.5)$.

## Definition

The binomial distribution with parameters $n$ and $p$ is the discrete probability distribution of the number of successes in a sequence of $n$ independent experiments, each with a binary outcome: success (with probability $p$ ) or failure ( with probability $q=1-p$ ). The pmf of the binomial distribution is

$$
P_{X}(x)=\binom{n}{x} p^{x}(1-p)^{1-x} \text { for } x=0, \ldots, n
$$

## Geometric distribution

Now imagine you toss the fair coin until you get heads. times. Some possible outcomes are $1,01,0001,0 \ldots 1$, etc. Let $X$ be the trial of the first success (heads). Then $X$ follows a Geometric distribution parameter 0.5.

Definition

## Joint Probability Mass Function



$$
\begin{aligned}
& P_{X, Y}(x, y)=P(X=x, Y=y)=P(X=x \& Y=y) \\
& \sum_{\text {all }(x, y)} P_{X, Y}(x, y)=1 \text { (still a probability mass function) }
\end{aligned}
$$

Marginal Probability: $P_{X}(x)=\sum_{y} P_{X, Y}(x, y)$ (sum over all possible y)

## Joint Probability mass function

## Definition

Let $X$ and $Y$ be random variables. If there are at most countable possible outcomes $(x, y)$ for the pair $(X, Y)$, we say that $X$ and $Y$ have a discrete joint distribution. The joint probability function (joint pf) is

$$
P(X=x \text { and } Y=y):=P(X=x, Y=y) \forall(x, y) \in R
$$

As for the univariate case we have

- $P(x, y) \geq 0$
- $\sum_{(x, y)} P_{X, Y}(x, y)=1$
- Two variables: Bivariate Distribution
- More than two variables: Multivariate Distribution


## Marginal Probability Mass Functions

Theorem
Let $(X, Y)$ be a discrete random vector with joint probability mass function $P_{X, Y}(x, y)$, then the marginal pmfs of $X$ and $Y$ are given by

$$
\begin{aligned}
& P_{X}(x)=\sum_{y \in R} P_{X, Y}(x, y) \\
& P_{Y}(y)=\sum_{x \in R} P_{X, Y}(x, y)
\end{aligned}
$$

## Conditional Probability Mass Functions



Conditional Probability: $P_{X \mid Y}(x \mid y)=\frac{P_{X, Y}(x, y)}{P_{Y}(y)}$

$$
\text { e.g., } P_{X \mid Y}(x \mid y=2)=\{2 / 9,4 / 9,1 / 9,2 / 9\}
$$

$\sum_{x} P_{X \mid Y}(x \mid y)=1$ (still a probability mass function)

## Independence of Random Variables

Independence for random variables is defined in the same way as for events

## Definition (Independent Random Variables)

Two random variables are independent if for every two sets $A$ and $B$ in $R$ the events $\{s: X(s) \in A\}$ and $\{s: Y(s) \in B\}$ are independent

Theorem
Two random variables $X$ and $Y$ with joint pmf $P_{X, Y}(x, y)$ and marginal pmfs $P_{X}(x)$ and $P_{Y}(y)$ are independent if and only if

$$
P_{X, Y}(x, y)=P_{X}(x) P_{Y}(y) \text { for all }(x, y) \in R^{2}
$$

## Conditional Independence

|  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- |
| 3 |  |  |  |  |
| 3 | $1 / 20$ | $2 / 20$ | $2 / 20$ |  |
| 2 | $2 / 20$ | $3 / 20$ | $1 / 20$ | $2 / 20$ |
| 1 |  | $2 / 20$ | $3 / 20$ |  |
| 0 | $1 / 20$ |  |  |  |

- Assume you know that $X \leq 2$ and $Y \geq 4$
- Are X and Y independent in this new universe?

Here we conditioned on an event, but in general $X$ and $Y$ are independent given $Z$ if

$$
P_{X, Y \mid Z}(x, y \mid z)=P_{X \mid Z}(x \mid z) P_{Y \mid Z}(y \mid z) \quad \forall(x, y, z) \in R^{3}
$$

## Exercise

A fair coin is tossed three times. Let

- $X=$ number of heads on the first toss
- $\mathrm{Y}=$ total number of heads
- Find the joint distribution of $X, Y$.
- Find the marginal distributions of $X$ and $Y$.
- Find the conditional distribution of $Y \mid X$
- Are $X$ and $Y$ independent?

Definition (Expectation)
The expected value or mean of $X$ is defined to be

$$
E(X)=\sum_{x} x P_{X}(x)
$$

assuming that the sum is well-defined.

We can think of the expectation as the average of a very large number of independent draws from the distribution (IID draws).

## Examples

- Let $X \sim \operatorname{Bernoulli}(p) . E(X)=$ ?
- Flip a fair coin twice. Let $X$ be the number of heads. $E(X)=$ ?


## Practice Exercises

| Section | Exercises |
| :--- | :--- |
| 3.1 | 9,10 |
| 3.4 | 2 |
| 3.5 | 2,5 |
| 3.6 | 2 |

