11: Joint (Multivariate) Distributions

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April 24th, 2024

Lecture Discussion on Ed
Normal Approximation
Normal Random Variables

\[ X \sim \mathcal{N}(\mu, \sigma^2) \]

- Used to model many real-life situations because it maximizes entropy (i.e., randomness) for a given mean and variance.

- Also useful for approximating the Binomial random variable!
Website testing

- 100 people are presented with a new website design.
- \(X = \#\text{ people whose time on site increases}\)
- PM assumes design has no effect, so assume \(P(\text{stickier}) = 0.5\) independently.
- CEO will endorse the new design if \(X \geq 65\).

What is \(P(\text{CEO endorses change})?\) Give a numerical approximation.

Approach 1: Binomial

Define

\[X \sim \text{Bin}(n = 100, p = 0.5)\]

Want: \(P(X \geq 65)\)

Solve

\[
P(X \geq 65) = \sum_{k=65}^{100} \binom{100}{k} 0.5^k (1 - 0.5)^{100-k}
\]
Don’t worry, Normal approximates Binomial

Galton Board

(We’ll explain why in 2 weeks)
Website testing

100 people are given a new website design.
• $X = \#$ people whose time on site increases
• PM assumes design has no effect, so $P(\text{stickier}) = 0.5$ independently.
• CEO will endorse the new design if $X \geq 65$.

What is $P(\text{CEO endorses change})$? Give a numerical approximation.

Approach 1: Binomial
Define
\[ X \sim \text{Bin}(n = 100, p = 0.5) \]
Want: $P(X \geq 65)$
Solve
\[ P(X \geq 65) \approx 0.0018 \]

Approach 2: approximate with Normal
Define
\[ Y \sim \mathcal{N}(\mu, \sigma^2) \]
\[ \mu = np = 50 \]
\[ \sigma^2 = np(1-p) = 25 \]
\[ \sigma = \sqrt{25} = 5 \]
Solve
\[ P(X \geq 65) \approx P(Y \geq 65) = 1 - F_Y(65) \]
\[ = 1 - \Phi \left( \frac{65 - 50}{5} \right) = 1 - \Phi(3) \approx 0.0013 \]

⚠️⚠️ (this approach is missing something important)

```
jerry$ python
>>> from scipy.stats import binom, norm
>>> binom.pmf(range(65, 101), n, p).sum()
0.001758820861485058
>>> 1 - norm(50, 5).cdf(65)
0.0013498980316301035
```
Website testing (with continuity correction)

In our website testing, $Y \sim \mathcal{N}(50, 25)$ approximates $X \sim \text{Bin}(100, 0.5)$.

$P(X \geq 65)$ Binomial

$\approx P(Y \geq 64.5)$ Normal

$\approx 0.0018$ ✔️ the better

Approach 2

You must perform a continuity correction when approximating a Binomial RV with a Normal RV.
### Continuity correction

If $Y \sim \mathcal{N}(np, np(1 - p))$ approximates $X \sim \text{Bin}(n, p)$, how do we approximate the following probabilities?

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<thead>
<tr>
<th>Discrete (e.g., Binomial) probability question</th>
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Continuity correction

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Who gets to approximate?

\[ X \sim \text{Bin}(n, p) \]
\[ E[X] = np \]
\[ \text{Var}(X) = np(1 - p) \]

\[ Y \sim \text{Poi}(\lambda) \]
\[ \lambda = np \]

\[ Y \sim \mathcal{N}(\mu, \sigma^2) \]
\[ \mu = np \]
\[ \sigma^2 = np(1 - p) \]
Who gets to approximate?

1. If there is a choice, use Gaussian to approximate.
2. When using Normal to approximate a discrete RV, use a continuity correction.
Stanford Admissions (a while back)

Stanford accepts 2480 students.
• Each admitted student matriculates with $p = 0.68$ (independently)
• Let $X = \#$ of students who will attend

What is $P(X > 1745)$? Give a numerical approximation.

Strategy:  
A. Just Binomial  
B. Poisson  
C. Normal  
D. None/other
Stanford Admissions (a while back)

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What is $P(X > 1745)$? *Give a numerical approximation.*

**Strategy:**

A. Just Binomial  
   computationally expensive (also not an approximation)

B. Poisson  
   $p = 0.68$, not small enough

C. Normal  
   ✔️ Variance $np(1 - p) = 540 > 10$

D. None/other

**Define an approximation**

Let $Y \sim \mathcal{N}(E[X], \text{Var}(X))$

$E[X] = np = 1686$

$\text{Var}(X) = np(1 - p) \approx 540 \rightarrow \sigma = 23.3$

$P(X > 1745) \approx P(Y \geq 1745.5)$

**Solve**

$P(Y \geq 1745.5) = 1 - F(1745.5)$

$= 1 - \Phi\left(\frac{1745.5 - 1686}{23.3}\right)$

$\approx 0.0055$
Discrete Joint RVs
From last slide deck

What is the probability that the Warriors win?
How do you model zero-sum games?

$p(A_W > A_B)$

This is a probability of an event involving two random variables!
Joint probability mass functions

Roll two 6-sided dice, yielding values $X$ and $Y$.

<table>
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<tr>
<th>$X$</th>
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</tr>
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<td>random variable</td>
<td>probability of an event</td>
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$P(X = k)$
probability mass function
Joint probability mass functions

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|$X, Y$| $P(X = 1 \cap Y = 6)$| $P(X = a, Y = b)$|
|random variables| $P(X = 1, Y = 6)$| joint probability mass function|
|new notation: the comma| probability of the intersection of two events|
Discrete joint distributions

For two discrete joint random variables $X$ and $Y$, the joint probability mass function is defined as:

$$p_{X,Y}(a,b) = P(X = a, Y = b)$$

The marginal distributions of the joint PMF are defined as:

$$p_X(a) = P(X = a) = \sum_y p_{X,Y}(a,y)$$

$$p_Y(b) = P(Y = b) = \sum_x p_{X,Y}(x,b)$$

Use marginal distributions to extract a 1D RV from a joint PMF.
Two dice

Roll two 6-sided dice, yielding values $X$ and $Y$.

1. What is the joint PMF of $X$ and $Y$?

$p_{X,Y}(a,b) = 1/36$  \hspace{1cm} (a, b) \in \{(1,1), \ldots, (6,6)\}$

<table>
<thead>
<tr>
<th>$Y$</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1/36</td>
<td>...</td>
<td>...</td>
<td>...</td>
<td>...</td>
<td>1/36</td>
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<tr>
<td>6</td>
<td>1/36</td>
<td>...</td>
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Probability table

- All possible outcomes for several discrete RVs
- Not parametric (e.g., parameter $p$ in $\text{Ber}(p)$)
Two dice

Roll two 6-sided dice, yielding values $X$ and $Y$.

1. What is the joint PMF of $X$ and $Y$?

   $$ p_{X,Y}(a, b) = 1/36 \quad (a, b) \in \{(1,1), \ldots, (6,6)\} $$

2. What is the marginal PMF of $X$?

   $$ p_X(a) = P(X = a) = \sum_y p_{X,Y}(a, y) = \sum_{y=1}^{6} \frac{1}{36} = \frac{1}{6} \quad a \in \{1, \ldots, 6\} $$
A computer (or three) in every house.

Consider households in Silicon Valley.
• A household has $X$ Macs and $Y$ PCs.
• Each house has a maximum of 3 computers total (Macs + PCs).

1. What is $P(X = 1, Y = 0)$, the missing entry in the probability table?
Consider households in Silicon Valley.

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1. What is $P(X = 1, Y = 0)$, the missing entry in the probability table?

A joint PMF must sum to 1:

$$\sum_x \sum_y p_{X,Y}(x, y) = 1$$

<table>
<thead>
<tr>
<th>X (# Macs)</th>
<th>0</th>
<th>1</th>
<th>2</th>
<th>3</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>.16</td>
<td>.12</td>
<td>.07</td>
<td>.04</td>
</tr>
<tr>
<td>1</td>
<td>.12</td>
<td>.14</td>
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A computer (or three) in every house.

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- A household has $X$ Macs and $Y$ PCs.
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2. How do you compute the marginal PMF of $X$?

\[
\begin{array}{c|cccc|c}
    & X (\# \text{ Macs}) & & & & \\
    & 0 & 1 & 2 & 3 & \\
    \hline
    0 & A & .16 & .12 & .07 & .04 & .39 \\
    1 & .12 & .14 & .12 & 0 & 38 \\
    2 & .07 & .12 & 0 & 0 & .19 \\
    3 & .04 & 0 & 0 & 0 & .04 \\
    \hline
    B & .39 & .38 & .19 & .04 & \\
\end{array}
\]
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- A household has $X$ Macs and $Y$ PCs.
- Each house has a maximum of 3 computers total (Macs + PCs).

2. How do you compute the marginal PMF of $X$?

\[
p_{X,Y}(x, 0) = P(X = x, Y = 0)
\]

A. Marginal PMF of $X$

\[
p_x(x) = \sum_y p_{x,y}(x, y)
\]

B. Marginal PMF of $Y$

\[
p_y(y) = \sum_x p_{x,y}(x, y)
\]

To find a marginal distribution over one variable, sum over all other variables in the joint PMF.
A computer (or three) in every house.

Consider households in Silicon Valley.

- A household has $X$ Macs and $Y$ PCs.
- Each house has a maximum of 3 computers total (Macs + PCs).

3. Let $C = X + Y$. What is $P(C = 3)$?

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$P(C = 3) = P(X + Y = 3)$

Law of Total Probability

$= \sum_x \sum_y P(X + Y = 3|X = x, Y = y)P(X = x, Y = y)$

$= P(X = 0, Y = 3) + P(X = 1, Y = 2) + P(X = 2, Y = 1) + P(X = 3, Y = 0)$

We’ll come back to sums of RVs next lecture!
Multinomial RV
Recall the good times

Permutations $n!$

How many ways are there to order $n$ objects?
Counting unordered objects

**Binomial coefficient**

How many ways are there to group $n$ objects into two groups of size $k$ and $n - k$, respectively?

$$\binom{n}{k} = \frac{n!}{k! (n-k)!}$$

Called the binomial coefficient because of something from aLgEbRa

**Multinomial coefficient**

How many ways are there to group $n$ objects into $r$ groups of sizes $n_1, n_2, ..., n_r$, respectively?

$$\binom{n}{n_1, n_2, ..., n_r} = \frac{n!}{n_1! n_2! \cdots n_r!}$$

Multinomials generalize Binomials for counting.
Probability

Binomial RV

What is the probability of getting \( k \) successes and \( n - k \) failures in \( n \) trials?

\[
P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}
\]

Binomial # of ways of ordering the successes

Probability of each ordering of \( k \) successes is equal + mutually exclusive

Multinomial RV

What is the probability of getting \( c_1 \) of outcome 1, \( c_2 \) of outcome 2, \ldots, and \( c_m \) of outcome \( m \) in \( n \) trials?

Multinomial RVs also generalize Binomial RVs for probability!
Multinomial Random Variable

Consider an experiment of \( n \) independent trials:
- Each trial results in one of \( m \) outcomes. \( P(\text{outcome } i) = p_i, \sum_{i=1}^{m} p_i = 1 \)
- Let \( X_i = \# \) trials with outcome \( i \)

Joint PMF

\[
P(X_1 = c_1, X_2 = c_2, \ldots, X_m = c_m) = \binom{n}{c_1, c_2, \ldots, c_m} p_1^{c_1} p_2^{c_2} \cdots p_m^{c_m}
\]

where \( \sum_{i=1}^{m} c_i = n \) and \( \sum_{i=1}^{m} p_i = 1 \)

Multinomial \# of ways of ordering the outcomes
Probability of each ordering is equal + mutually exclusive
Hello dice rolls, my old friends

A fair, six-sided die is rolled 7 times. What is the probability of getting:

- 1 one
- 1 two
- 0 threes
- 2 fours
- 0 fives
- 3 sixes
Hello dice rolls, my old friends

A fair, six-sided die is rolled 7 times. What is the probability of getting:

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\[
P(X_1 = 1, X_2 = 1, X_3 = 0, X_4 = 2, X_5 = 0, X_6 = 3) = \binom{7}{1,1,0,2,0,3} \left(\frac{1}{6}\right)^1 \left(\frac{1}{6}\right)^1 \left(\frac{1}{6}\right)^0 \left(\frac{1}{6}\right)^2 \left(\frac{1}{6}\right)^0 \left(\frac{1}{6}\right)^3 = 420 \left(\frac{1}{6}\right)^7
\]
Hello dice rolls, my old friends

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\[ P(X_1 = 1, X_2 = 1, X_3 = 0, X_4 = 2, X_5 = 0, X_6 = 3) \]

\[ = \binom{7}{1, 1, 0, 2, 0, 3} \left( \frac{1}{6} \right)^1 \left( \frac{1}{6} \right)^1 \left( \frac{1}{6} \right)^0 \left( \frac{1}{6} \right)^2 \left( \frac{1}{6} \right)^0 \left( \frac{1}{6} \right)^3 = 420 \left( \frac{1}{6} \right)^7 \]
Probabilistic text analysis

Ignoring the order of words...

What is the probability of any given word that you write in English?

• \( P(\text{word } = \text{"the"}) > P(\text{word } = \text{"susurration"}) \)
• \( P(\text{word } = \text{"Stanford"}) > P(\text{word } = \text{"Cal"}) \)

Probabilities of counts of words = Multinomial distribution

A document is a large multinomial.

(according to the Global Language Monitor, there are 988,968 words in the English language used on the internet.)
Probabilistic text analysis

Probabilities of counts of words = multinomial distribution

Example document:

"When my late husband was alive he deposited some amount of Money with overseas Bank in which the amount will be declared to you once you respond to this message indicating your interest in helping to receive the fund and use it for Heavens work as my wish."

\[ P \left( \begin{array}{c} \text{bank} = 1 \\ \text{fund} = 1 \\ \text{money} = 1 \\ \text{wish} = 1 \\ \text{to} = 3 \\ \ldots \end{array} \bigg| 	ext{spam} \right) = \frac{48!}{1! 1! 1! 1! \cdots 3!} p_1^1 p_1^1 p_1^1 \cdots p_3^3 \]

Note: \( P(\text{bank}|\text{spam}) \gg P(\text{bank}|\text{legit}) \)
Old and New Analysis

Authorship of the Federalist Papers
• 85 essays advocating ratification of the US constitution
• Written under the pseudonym "Publius" (really, Alexander Hamilton, James Madison, John Jay)

Who wrote which essays?
• Analyze probability of words in each essay and compare against word distributions from known writings of three authors
• Curious what the analysis is? Read this!
Statistics of Two RVs
Expectation and Covariance

In real life, we often have many RVs interacting at once.

- We’ve seen some simpler cases (e.g., sum of independent Bernoullis).
- Come Friday, we’ll discuss sums of Binomials, Poissons, etc.
- In general, manipulating joint PMFs is difficult.
- Fortunately, you don’t need to model joint RVs completely all the time.

Instead, we’ll focus next on reporting statistics of multiple RVs:

- **Expectation of sums** (you’ve seen some of this, more on Friday)
- **Covariance**: measure of how two random variable vary with each other (more next Monday and Wednesday)
Properties of Expectation, extended to two RVs

1. Linearity:
\[ E[aX + bY + c] = aE[X] + bE[Y] + c \]

2. Expectation of a sum = sum of expectation:
\[ E[X + Y] = E[X] + E[Y] \]

3. Unconscious statistician:
\[ E[g(X,Y)] = \sum_x \sum_y g(x,y)p_{X,Y}(x,y) \]

True for both independent and dependent random variables!

we’ve seen this!
we’ll prove momentarily.
Proof of expectation of a sum of RVs

\[
E[X + Y] = \sum_x \sum_y (x + y)p_{X,Y}(x, y)
\]

\[
= \sum_x \sum_y xp_{X,Y}(x, y) + \sum_x \sum_y yp_{X,Y}(x, y)
\]

\[
= \sum_x x \sum_y p_{X,Y}(x, y) + \sum_y y \sum_x p_{X,Y}(x, y)
\]

\[
= \sum_x xp_X(x) + \sum_y yp_Y(y)
\]

\[
= E[X] + E[Y]
\]

LOTUS,
\( g(X, Y) = X + Y \)

Linearity of summations
(cont. case: linearity of integrals)

Marginal PMFs for \( X \) and \( Y \)