11: Joint (Multivariate) Distributions

Lisa Yan April 29, 2020

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11a_normal_approx

Normal Approximation

Normal RVs

 $X\!\sim\!\mathcal{N}\,(\mu,\sigma)$ 2) mean variance

- 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 given a new website design.
- $X = #$ people whose time on site increases
- The design actually has no effect, so P (time on site increases) = 0.5 independently.
- CEO will endorse the new design if $X \geq 65$.

What is P (CEO endorses change)? Give a numerical approximation.

Approach 1: Binomial

Define

$$
X \sim Bin(n = 100, p = 0.5)
$$

Want: $P(X \ge 65)$

Solve

$$
P(X \ge 65) = \sum_{i=65}^{100} {100 \choose i} 0.5^{i} (1-0.5)^{100-i}
$$

Don't worry, Normal approximates Binomial

Galton Board

(We'll explain *why* in 2 weeks' time)

Website testing

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Approach 1: Binomial

Define

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

Want: $P(X \ge 65)$

Solve

 $P(X \ge 65) \approx 0.0018$

 \mathbf{A}

Approach 2: approximate with Normal

Define $Y \sim \mathcal{N}(\mu, \sigma^2)$ $\sigma^2 = np(1-p) = 25$ $\mu = np = 50$ $\sigma = \sqrt{25} = 5$ Solve

$$
P(X \ge 65) \approx P(Y \ge 65) = 1 - F_Y(65)
$$

= 1 - $\Phi\left(\frac{65 - 50}{5}\right) = 1 - \Phi(3) \approx 0.0013$?

Lisa Yan, CS109, 2020 ⚠️ (this approach is actually missing something)

Website testing (with continuity correction)

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

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Continuity correction

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

Continuity correction

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

Who gets to approximate?

Who gets to approximate?

1. If there is a choice, use Normal to approximate. 2. When using Normal to approximate a discrete RV, use a continuity correction.

11b_discrete_joint

Discrete Joint RVs

From last time

 $P(A_W > A_B)$

This is a probability of an event involving *two* random variables!

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

Review

Joint probability mass functions

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

$$
P(X=1)
$$

probability of an event

 $P(X=k)$

probability mass function

Joint probability mass functions

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

$$
P(X = k)
$$

probability mass function

X

random variable

random variables

 $P(X = 1 \cap Y = 6)$

 $P(X = 1)$

probability of

an event

$$
P(X=1, Y=6)
$$

 $P(X = a, Y = b)$

new notation: the comma

probability of the intersection of two events

joint probability mass function

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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 get a 1-D 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 ?

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Probability table

- All possible outcomes for several discrete RVs
- Not parametric (e.g., parameter p in Ber(p))

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Two dice

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

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

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

2. What is the marginal PMF of X ?

$$
p_X(a) = P(X = a) = \sum_{y} p_{X,Y}(a, y) = \sum_{y=1}^{6} \frac{1}{36} = \frac{1}{6}
$$
 $a \in \{1, ..., 6\}$

Consider households in Silicon Valley.

- A household has X Macs and Y PCs.
- Each house has a maximum of 3 computers (Macs $+$ PCs) in the house.
- 1. What is $P(X = 1, Y = 0)$, the missing entry in the probability table?

 $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$

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:

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

Consider households in Silicon Valley.

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

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

\n- A.
$$
p_{X,Y}(x, 0) = P(X = x, Y = 0)
$$
\n- B. Marginal PMF of X $p_X(x) = \sum_y p_{X,Y}(x, y)$
\n- C. Marginal PMF of Y $p_Y(y) = \sum_x p_{X,Y}(x, y)$
\n

To find a marginal distribution over one variable, sum over all other variables in the joint PMF.

Consider households in Silicon Valley.

- A household has X Macs and Y PCs.
- Each house has a maximum of 3 computers (Macs $+$ PCs) in the house.
- 3. Let $C = X + Y$. What is $P(C = 3)$?

Consider households in Silicon Valley.

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

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

0 1 2 3 0 .16 .12 .07 .04 1 .12 .14 .12 0 2 .07 .12 0 0 3 .04 0 0 0 (# Macs) (# PCs) = 3 = + = 3 = + = 3| = , = = , = = = 0, = 3 + = 1, = 2 + = 2, = 1 + = 3, = 0 Law of Total Probability

We'll come back to sums of RVs next lecture!

11c_multinomial

Multinomial RV

Recall the good times

Permutations $n!$ How many ways are there to order n 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?

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

Multinomials generalize Binomials for counting.

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Probability

Binomial RV

What is the probability of getting k successes and $n - k$ failures in n trials?

Multinomial RV

What is the probability of getting c_1 of outcome 1, c_2 of outcome 2, ..., and c_m of outcome m in n trials?

$$
P(X = k) = {n \choose k} p^k (1-p)^{n-k}
$$

Binomial # of ways of ordering the successes

Probability of each ordering of k successes is equal + mutually exclusive

Multinomial 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. $P(\text{outcome } i) = p_i, \sum_{i=1}^{n} p_i$ $p_i = 1$
- Let $X_i = #$ trials with outcome i

 $i=1$

 \overline{m}

Hello dice rolls, my old friends

A 6-sided die is rolled 7 times. What is the probability of getting:

- 1 one • 0 threes • 0 fives
- 1 two • 2 fours • 3 sixes

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$$
P(X_1 = 1, X_2 = 1, X_3 = 0, X_4 = 2, X_5 = 0, X_6 = 3)
$$

$$
= {7 \choose 1,1,0,2,0,3} \left(\frac{1}{6}\right)^1 \left(\frac{1}{6}\right)^1 \left(\frac{1}{6}\right)^0 \left(\frac{1}{6}\right)^2 \left(\frac{1}{6}\right)^0 \left(\frac{1}{6}\right)^3 = 420 \left(\frac{1}{6}\right)^7
$$

Hello dice rolls, my old friends

A 6-sided die is rolled 7 times. What is the probability of getting:

- \cdot 1 one • 0 threes • 0 fives
- 1 two • 2 fours • 3 sixes

of times a six appears

 $P(X_1 = 1, X_2 = 1, X_3 = 0, X_4 = 2, X_5 = 0, X_6 = 3)$

$$
= {7 \choose 1,1,0,2,0,3} \left(\frac{1}{6} \right)^1 \left(\frac{1}{6} \right)^1 \left(\frac{1}{6} \right)^0 \left(\frac{1}{6} \right)^2 \left(\frac{1}{6} \right)^0 \left(\frac{1}{6} \right)^3 = 420 \left(\frac{1}{6} \right)^7
$$

\nchoose *probability*
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\nthe sixes appear
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(live) 11: Joint (Multivariate) Distributions

Slides by Lisa Yan April 29, 2020

Normal RVs

 $X\!\sim\!\mathcal{N}\,(\mu,\sigma)$ 2) mean variance

- 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!

Who gets to approximate?

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

- Computing probabilities on Binomial RVs is often computationally expensive.
- Two reasonable approximations, but when to use which?

 $Y \sim Poi(\lambda)$ $\lambda = np$

n large (> 20) p small (< 0.05) slight dependence okay

$$
Y \sim \mathcal{N}(\mu, \sigma^2)
$$

$$
\mu = np
$$

$$
\sigma^2 = np(1-p)
$$

n large (> 20), *p* mid-ranged $(np(1-p) > 10)$ independence need continuity correction

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Review

Think

Check out the question on the next slide. Post any clarifications here!

<https://us.edstem.org/courses/667/discussion/90049>

Stanford Admissions (a while back)

Stanford accepts 2480 students.

- Each accepted student has 68% chance of attending (independent trials)
- 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

Poisson Normal None/other

$p = 0.68$, not small enough

$$
\checkmark
$$
 Variance $np(1-p) = 540 > 10$

not an approximation (also computationally expensive)

Define an approximation and Solve

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

\n
$$
E[X] = np = 1686
$$
\n
$$
\text{Var}(X) = np(1-p) \approx 540 \to \sigma = 23.3
$$
\n
$$
P(Y \ge 1745.5) = 1 - \Phi\left(\frac{1745.5 - 1686}{23.3}\right)
$$
\n
$$
P(X > 1745) \approx P(Y \ge 1745.5) \text{ A } \frac{\text{Continuity}}{\text{correction}} = 1 - \Phi(2.54) \approx 0.0055
$$

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23.3

Changes in Stanford Admissions

Stanford accepts 2480 students.

- Each accepted student has 68% chance of attending (independent trials)
- Let $X = #$ of students who will attend

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

The Stanford Baily

OPINIONS \cdot **| ARTS & LIFE** \cdot **| THE GRIND | MULTIMEDIA** \cdot FEATURES | ARCHIVES

Class of 2018 admit rates lowest in University history

March 28, 2014 16 Comments

 $\frac{1}{2}$ Like 901

Alex Zivkovic Desk Editor

Stanford admitted 2,138 students to the

Class of 2018 in this year's admissions cycle, producing $-$ at 5.07 percent $-$ the lowest admit rate in University history.

The University received a total of 42,167 applications this year, a record total and a 8.6 percent increase over last year's figure of 38,828. Stanford accepted 748 students

Overview for the Class of 2022

- Total Applicants: 47,451 Admit rate: 4.3%
- Total Admits: 2,071 ٠

Yield rate: 81.9%

Total Enrolled: 1,706 ٠

People love coming to Stanford!

Yield rate 20

years ago

Multinomial Random Variable

Review

 $i=1$

Consider an experiment of n independent trials:

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

Joint PMF
\n
$$
P(X_1 = c_1, X_2 = c_2, ..., X_m = c_m) = {n \choose c_1, c_2, ..., c_m} p_1^{c_1} p_2^{c_2} ... p_m^{c_m}
$$
\nwhere
$$
\sum_{i=1}^m c_i = n \text{ and } \sum_{i=1}^m p_i = 1
$$

Example:

- Rolling 2 twos, 3 threes, and 5 fives on 10 rolls of a fair-sided die
- Generating a random 5-word phrase with 1 "the", 2 "bacon", 1 "put", 1 "on"

Review

A 6-sided die is rolled 7 times. What is the probability of getting:

• 1 one • 0 threes • 0 fives

• 1 two • 2 fours • 3 sixes

of times a six appears

$$
P(X_1 = 1, X_2 = 1, X_3 = 0, X_4 = 2, X_5 = 0, X_6 = 3)
$$

$$
= {7 \choose 1,1,0,2,0,3} \left(\frac{1}{6} \right)^1 \left(\frac{1}{6} \right)^1 \left(\frac{1}{6} \right)^0 \left(\frac{1}{6} \right)^2 \left(\frac{1}{6} \right)^0 \left(\frac{1}{6} \right)^3 = 420 \left(\frac{1}{6} \right)^7
$$

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Parameters of a Multinomial RV?

 $X \sim Bin(n, p)$ has parameters $n, p...$

$$
P(X = k) = {n \choose k} p^k (1-p)^{n-k}
$$

p: probability of success outcome on a single trial

A Multinomial RV has parameters $n, p_1, p_2, ..., p_m$ (Note $p_m = 1 - \sum_{i=1}^{m-1} p_i)$

$$
P(X_1 = c_1, X_2 = c_2, ..., X_m = c_m) = {n \choose c_1, c_2, ..., c_m} p_1^{c_1} p_2^{c_2} \cdots p_m^{c_m}
$$

 $p_{\bm{i}}$: probability of outcome \bm{i} on a single trial

Where do we get p_i from?

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Interlude for fun/announcements

Announcements

More OH!

Estimating Coronavirus Prevalence by Cross-Checking Countries

We'll make the modeling assumption that N_{ij} is a Poisson distribution with rate **parameter A**_{ij} $*$ λ _i $*$ α _j. What this means is that the expected number of cases should be equal to the total amount of travel, times some source-dependent multiplier α _i ..., times some countrydependent multiplier λ_i (the infection prevalence in country i).

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POISSON!!

!!!!!!!!

https://medium.com/@jsteinhardt/estimating-coronavirus[prevalence-by-cross-checking-countries-c7e4211f0e18](https://medium.com/@jsteinhardt/estimating-coronavirus-prevalence-by-cross-checking-countries-c7e4211f0e18)

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Ethics in Probability: Biased Data + Bayes

Amazon scraps secret AI recruiting tool that showed bias against women

"In effect, Amazon's system *taught itself that male candidates were preferable*. It penalized resumes that included the word 'women's,' as in 'women's chess club captain."'

Basic Bayes Algorithm**:** Pick highest P(H|resume)

Let resumeF be a resume associated with a female applicant.

P(H|resumeF) = P(resumeF|H)*P(H)/P(resumeF)

Because of biased historical data, P(resumeF|H) is small $\odot \rightarrow$

Therefore P(H|resumeF) may be higher than P(H|resumeM) *simply because of biased historical data, rather than comparative candidate skillsets.*

GLOBAL HEADCOUNT

 \blacksquare Male \blacksquare Female

EMPLOYEES IN TECHNICAL ROLES

https://www.reuters.com/article/us-amazon-com-jobs-automation[insight/amazon-scraps-secret-ai-recruiting-tool-that-showed-bias](https://www.reuters.com/article/us-amazon-com-jobs-automation-insight/amazon-scraps-secret-ai-recruiting-tool-that-showed-bias-against-women-idUSKCN1MK08G)against-women-idUSKCN1MK08G

Ethics in Probability: Biased Data + Bayes

What if we ignore gender traits?

Amazon edited the programs to make them neutral to these particular terms. But that was no guarantee that the machines would not devise other ways of sorting candidates that could prove discriminatory

[After re-training…] the technology favored candidates who described themselves using verbs more commonly found on male engineers' resumes, such as "executed" and "captured," one person said.

This is an open question in a field called Algorithmic Fairness.

EMPLOYEES IN TECHNICAL ROLES

https://www.reuters.com/article/us-amazon-com-jobs-automation[insight/amazon-scraps-secret-ai-recruiting-tool-that-showed-bias](https://www.reuters.com/article/us-amazon-com-jobs-automation-insight/amazon-scraps-secret-ai-recruiting-tool-that-showed-bias-against-women-idUSKCN1MK08G)against-women-idUSKCN1MK08G

LIVE

The Federalist Papers

Ignoring the order of words…

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

- $P(word = "the") > P(word = "pokemon")$
- $P(word = "Stanford") > P(word = "Cal")$

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.)

Probabilities of *counts* of words = Multinomial distribution

Example document:

#words: $n = 48$

"When my late husband was alive he deposited some amount of Money with china 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 Gods work as my wish."

bank = 1
\n
$$
P\left(\begin{array}{c}\n\tanh = 1 \\
\tanh = 1 \\
\textrm{wish} = 1\n\end{array}\right) = \frac{n!}{1! \ 1! \ 1! \ 1! \ 1! \ \cdots 3!} p_{bank}^{1} p_{fund}^{1} \cdots p_{to}^{3}
$$
\nto = 3
\nNote: $P\left(\text{bank} \mid \text{spam}\right) \gg P\left(\text{bank} \mid \text{writer}\mid \text{you}\right)$

Probabilities of *counts* of words = Multinomial distribution

What about probability of those same words in someone else's writing? • P (word = "probability" writer = $\left(\begin{matrix} 1 & 1 \\ 1 & 1 \end{matrix}\right) > P\left(\text{word} = \text{``probability''}\right)$ writer = non−CS109 student

To determine authorship:

- 1. Estimate P (word writer) from known writings
- 2. Use Bayes' Theorem to determine $P(\text{writer}|\text{document})$ for a new writing!

Who wrote the Federalist Papers?

 \overline{f}

LIVE

See recordings 10e_all….

LIVE

Up next: Independent RVs and Sums!