

Lecture 6
(start Chap 2)

10/3/16

1.

Counting # of elements in (large!) sets

1. Multiplication rule: Let n experiments be performed in order, and let m_1, m_2, \dots, m_n be the # of possible outcomes of exp. 1, 2, ..., n resp. Then there are $m_1 m_2 \dots m_n$ possible outcomes for the ordered seq of n exp.

Ex 1: A has 8 pairs of earrings, 7 dresses and 12 pairs of shoes.

How many possible outfits: $8 \cdot 7 \cdot 12$

Ex 2: 30 students in class \rightarrow how many seating configurations

$30!$

n Factorial: $n! = n(n-1)(n-2)\dots 1$ ($0! = 1$)
 $n! = n \cdot (n-1)!$

2. Permutations:

Ex: Soccer team of 16 \rightarrow $\left. \begin{array}{l} 1 \text{ Pres} \\ 1 \text{ VP} \\ 1 \text{ Treas} \end{array} \right\}$.

$$16 \cdot 15 \cdot 14 = \frac{16!}{(16-3)!} = P_{16,3}$$

For $k \leq n$: $P_{n,k} = \frac{n!}{(n-k)!} \sim$ # of permutations of k objects from a set of n (ordered!)

3. Combinations

Ex: Soccer team of 16 \sim choose admin committee of 3 members

$\frac{P_{16,3}}{(3!)}$ \sim # of ways 3 people can be ordered

For $k \leq n$: $C_{n,k} = \frac{P_{n,k}}{k!} = \frac{n!}{(n-k)!k!} \sim$ # of combinations of k objects from set of n (unordered!)

2.7 #5

#10: how many 4-letter words: ^{1. no letter repeated twice} $\rightarrow P_{26,4} = \# \text{ of 4 letter words with no repeated letters}$
^{2. must contain at least 1 vowel}

Important properties

Now, subtract those without vowels: $P_{21,4}$

• For $k \leq n$: $C_{n,k} = \frac{n!}{(n-k)!k!} = C_{n,n-k}$ (Symmetry)

• Pascal's triangle

$$\begin{array}{ccccccc}
 & & & & 1 & & & & \\
 & & & & & 1 & & 1 & \\
 & & & 1 & & 2 & & 1 & \\
 & & 1 & & 3 & & 3 & & 1 \\
 1 & & 4 & & 6 & & 4 & & 1 \\
 & & & & & C_{n,k} & & &
 \end{array}$$

$$C_{n,k} = C_{n-1,k-1} + C_{n-1,k}$$

• Binomial Thm: $(x+y)^n = \sum_{i=0}^n C_{n,i} x^i y^{n-i}$

2.7 # 15 : 10 G (L+R) # of lin cups?

12 C, F(L+R)

$P_{10,2} \cdot P_{12,3}$ (order matters!)

2.7 # 13 : 12 F, 7 invites

(a) A, B feud and can't go both to party : $C_{12,7} - C_{10,5} = 540$

(b) A, B either together, or both don't go : $C_{10,5} + C_{10,7} = 372$

Multinomial

Recall : $C_{m,k}^1 =$ # of ways to pick k objects from n (order does not matter)

Thus : 2 groups, n_1 - those selected (k from n)
 those not selected ($n-k$ from n)

Q: What if we have multiple groups?

Ex: 30 students : $\begin{cases} 5 A's \\ 10 B's \\ 15 C's \end{cases}$ How many ways

$C_{30,5} =$ # of ways to select the 5 A students from the class

\Rightarrow 25 left $\Rightarrow C_{25,10}$ ways to pick

\Rightarrow Multiplication Rule: $C_{30,5} \cdot C_{25,10} = \frac{30!}{5!25!} \cdot \frac{25!}{10!15!} = \frac{30!}{5!10!15!}$

Then: n objects, k groups with $m_1, m_2, \dots, m_k \in \{0, \dots, n\}$ but $m_1 + \dots + m_k = n$

$\Rightarrow \frac{n!}{m_1! m_2! \dots m_k!}$ ways to distribute n objects over these k groups.

2.7 # 27 (a) $4! = 24$

(b) From (a) there are 24 different ways to form 4 couples.
 Each such group can marry in order in $4!$ ways $\left\{ = \binom{4}{1} \cdot 4! = 576 \right\}$

(c) $\frac{8!}{2!2!2!2!} / 4! = 105$

(d) $\frac{8!}{2!2!2!2!} = 2520$

(e) $8!$

(f) Note: There are 4 fixed, married couples!

There are 4 MF couples. These can be lined up in $4!$ ways.

Each couple can line up in 2 ways: (M, F) or (F, M)

$$2^4 \cdot 4! = 384$$

Binomial and Multinomial distributions

Do n experiments independently

Let p be the prob. of success in 1 experiment

$X =$ RV for the # of successes in n experiments (so $X \in \{0, 1, 2, \dots, n\}$)

Then X is a Binomial RV $(n, p) = \text{Bin}(n, p)$

Distribution fcn: $P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$, $k=0, 1, \dots, n$

Ex: Roll a die 12 times.

What is prob. to get 4 twos? At most 4 twos? More than 4 twos?

A: Roll a die once \rightarrow prob. to get a two = $\frac{1}{6}$

Let $X =$ RV for the # of twos in 12 rolls

$$P(X > 4) = 1 - P(X \leq 4)$$

$$P(X=4) = \binom{12}{4} \left(\frac{1}{6}\right)^4 \left(\frac{5}{6}\right)^8$$

$$P(X \leq 4) = \binom{12}{0} \left(\frac{1}{6}\right)^0 \left(\frac{5}{6}\right)^{12} + \binom{12}{1} \left(\frac{1}{6}\right)^1 \left(\frac{5}{6}\right)^{11} + \binom{12}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^{10} + \binom{12}{3} \left(\frac{1}{6}\right)^3 \left(\frac{5}{6}\right)^9 + \binom{12}{4} \left(\frac{1}{6}\right)^4 \left(\frac{5}{6}\right)^8$$

Thm: If $X \sim \text{Bin}(n, p)$, then $E(X) = np$
 $\text{Var}(X) = np(1-p)$

(typos: \oplus p 43, l 2 no m in Σ
 \oplus p 45, l 4: " $38/51$ " \rightarrow " $38/49 = 0.776$ ")

2.7# 29

33

Thm: $X \sim \text{Bin}(n, p)$ has $E(X) = np$ and $\text{Var}(X) = np(1-p)$

Pf: Proof 1 (short): Let x_i be the RV for success in 1 try

$$\text{Thus } E(x_i) = 0 \cdot (1-p) + 1 \cdot (p) = p$$

$$\text{Var}(x_i) = E(x_i^2) - (E(x_i))^2 = 0^2 \cdot (1-p) + 1^2 \cdot (p) - p^2 = p - p^2 = p(1-p)$$

Now: $X = x_1 + x_2 + \dots + x_n$, where the x_i are i.i.d.

$$\rightarrow E(X) = \overset{\substack{\text{proved later} \\ \uparrow}}{E(x_1)} + \dots + E(x_n) = p + \dots + p = np$$

$$\text{Var}(X) = \text{Var}(x_1) + \dots + \text{Var}(x_n) = p(1-p) + \dots + p(1-p) = np(1-p)$$

because the x_i are independent (proved later)

Proof 2 (longer):

$$E(X) = \sum_{k=0}^n k \binom{n}{k} p^k (1-p)^{n-k} = \sum_{k=1}^n k \frac{n!}{k!(n-k)!} p^k (1-p)^{n-k} = \sum_{k=1}^n \frac{n!}{(k-1)!(n-k)!} p^k (1-p)^{n-k}$$

$$\text{(KEY)} = np \sum_{k=1}^n \frac{(n-1)!}{(k-1)!(n-1-(k-1))!} p^{k-1} (1-p)^{n-1-(k-1)} = np \sum_{l=0}^{n-1} \frac{(n-1)!}{l!(n-1-l)!} p^l (1-p)^{n-1-l} = np \sum_{l=0}^n \binom{n-1}{l} p^l (1-p)^{n-1-l}$$

$$= np \cdot 1 = np$$

$\text{Var}(X) = E(X^2) - (E(X))^2 \stackrel{\text{trick}}{=} E(X(X-1)) + E(X) - (E(X))^2$. Since $E(X)$ known, let's compute $E(X(X-1))$

$$E(X(X-1)) = \sum_{k=0}^n k(k-1) \binom{n}{k} p^k (1-p)^{n-k} = \sum_{k=2}^n k(k-1) \frac{n!}{k!(n-k)!} p^k (1-p)^{n-k}$$

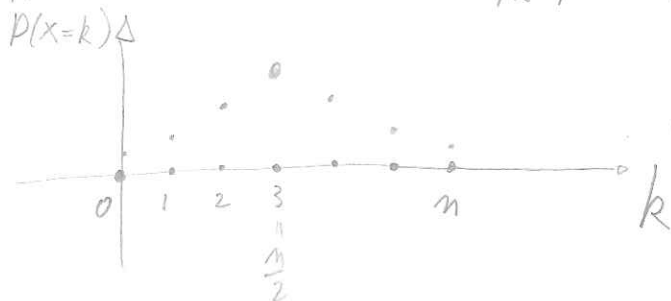
$$\text{(KEY)} = n(n-1) p^2 \sum_{k=2}^n \frac{(n-2)!}{(k-2)!(n-2-(k-2))!} p^{k-2} (1-p)^{n-2-(k-2)}$$

$$= n(n-1) p^2 \sum_{l=0}^{n-2} \frac{(n-2)!}{l!(n-2-l)!} p^l (1-p)^{n-2-l} = n(n-1) p^2 \sum_{l=0}^n \binom{n-2}{l} p^l (1-p)^{n-2-l}$$

$$= n(n-1) p^2$$

$$\rightarrow \text{Var}(X) = n(n-1) p^2 + np - (np)^2 = \cancel{n^2 p^2} - np^2 + np - \cancel{(np)^2} = np(1-p)$$

Plot $\text{Bin}(n, p) \rightarrow P(x=k) = C_{n,k} p^k (1-p)^{n-k}$



Ex: $p = \frac{1}{2} \rightarrow np = \frac{n}{2}$. Then $P(x=k) = \left(\frac{1}{2}\right)^n C_{n,k} = \left(\frac{1}{2}\right)^n \frac{n!}{(n-k)!k!}$ which is

maximized (as a f^{ion} of k, for fixed n) when $k = \frac{n}{2}$, i.e. when $k = E(x)$.

Furthermore, since $C_{n,k} = C_{n,n-k}$ and $p = \frac{1}{2}$, the plot of the distribution function will be symmetrical around the expected value $\frac{n}{2}$.

Ex: What if $p < \frac{1}{2}$ or $> \frac{1}{2}$ (experiment at home, or online)

EX 2.16, p44 in text: An example of when not to use Bin!

Pick 13 cards from deck of 52.

Q: $P(1 \text{ ace}), P(2 \text{ aces}), P(3 \text{ aces}), P(4 \text{ aces})$

Tempting, but false A: Use $\text{Bin}(4, \frac{1}{4})$.

This is false because the 4 experiments (of success or unsuccessfully drawing aces) are not independent \rightarrow Reason: There is no replacement

k Exact

Bin(4, 1/4)

0 $\frac{C_{4,0} C_{48,13}}{C_{52,13}}$

$C_{4,0} \left(\frac{1}{4}\right)^0 \left(\frac{3}{4}\right)^4$

1 $\frac{C_{4,1} C_{48,12}}{C_{52,13}}$

$C_{4,1} \left(\frac{1}{4}\right)^1 \left(\frac{3}{4}\right)^3$

2 $\frac{C_{4,2} C_{48,11}}{C_{52,13}} \approx 0.2134$

$C_{4,2} \left(\frac{1}{4}\right)^2 \left(\frac{3}{4}\right)^2 \approx 0.2109$

3 $\frac{C_{4,3} C_{48,10}}{C_{52,13}}$

$C_{4,3} \left(\frac{1}{4}\right)^3 \left(\frac{3}{4}\right)^1$

4 $\frac{C_{4,4} C_{48,9}}{C_{52,13}}$

$C_{4,4} \left(\frac{1}{4}\right)^4 \left(\frac{3}{4}\right)^0$

2.7 (#29)

33: 10 samples of beer
guess at lead std
(a) if guessing
(b) if $p=0.9$

Ex 2.16 on p 44

Draw 13 cards from deck of 52.

$$P(\text{ace of H is drawn}) = \frac{C_{51,12}}{C_{52,13}} = \frac{\frac{51 \cdot 50 \cdots 40}{12!}}{\frac{52 \cdot 51 \cdots 40}{13 \cdot 12!}} = \frac{13}{52} = \frac{1}{4} = P(\text{ace of D}) = P(\text{ace of C}) = P(\text{ace of S})$$

But: These probabilities are not independent:

$$\text{ex: } P(\text{ace of H} \cap \text{ace of D}) = \frac{C_{50,11}}{C_{52,13}} = \frac{\frac{50 \cdots 40}{11!}}{\frac{52 \cdot 51 \cdot 50 \cdots 40}{13 \cdot 12 \cdot 11!}} = \frac{18 \cdot 12^3}{52 \cdot 51} = \frac{1}{17} \neq P(\text{ace of H})P(\text{ace of D}) = \frac{1}{16}$$

k # of aces drawn	Exact	Bin(4, $\frac{1}{4}$)
0	$\frac{C_{48,13}}{C_{52,13}} = 0.304$	$C_{4,0} \left(\frac{3}{4}\right)^4 = 0.316$
1	$\frac{C_{4,1} C_{48,12}}{C_{52,13}} = 0.439$	$C_{4,1} \left(\frac{1}{4}\right) \left(\frac{3}{4}\right)^3 = 0.422$
2	$\frac{C_{4,2} C_{48,11}}{C_{52,13}} = 0.213$	$C_{4,2} \left(\frac{1}{4}\right)^2 \left(\frac{3}{4}\right)^2 = 0.211$
3	$\frac{C_{4,3} C_{48,10}}{C_{52,13}} = 0.041$	$C_{4,3} \left(\frac{1}{4}\right)^3 \left(\frac{3}{4}\right) = 0.047$
4	$\frac{C_{4,4} C_{48,9}}{C_{52,13}} = 0.003$	$C_{4,4} \left(\frac{1}{4}\right)^4 = 0.004$

Compare pretty well!

(Ex 2.7# 33)Remark: Multinomial distribution

↳ more than 2 categories, compared to the F, S in binomial

Ex: Experiment: height of person:

$$P(H \leq 5'5") = 30\%$$

$$P(5'5" < H \leq 6') = 40\%$$

$$P(H > 6') = 30\%$$

Measure 20 people: Prob that 5 people are $\leq 5'5"$, 10 between $5'5"$ and $6'$, and that 5 people are $> 6'$

$$= \frac{20!}{5!10!5!} (0.3)^5 (0.4)^{10} (0.3)^5$$

Poisson distribution

$X \sim \text{Pois}(\lambda)$ if $P(X=k) = e^{-\lambda} \frac{\lambda^k}{k!}$, $k=0, 1, 2, \dots$
($\lambda > 0$ is parameter)

Thm: If $X \sim \text{Pois}(\lambda)$, then $E(X) = \lambda = \text{Var}(X)$

$$\begin{aligned}
 \text{Pf: } E(X) &= \sum_{k=0}^{\infty} k e^{-\lambda} \frac{\lambda^k}{k!} = e^{-\lambda} \sum_{k=1}^{\infty} \frac{\lambda^k}{(k-1)!} = e^{-\lambda} \cdot \lambda \sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!} \\
 &= e^{-\lambda} \lambda \sum_{l=0}^{\infty} \frac{\lambda^l}{l!} = e^{-\lambda} \lambda \cdot e^{\lambda} = \lambda
 \end{aligned}$$

$$\bullet \text{Var}(X) = E(X^2) - (E(X))^2 \quad (1)$$

Will calculate $E(X(X-1))$, which is easier than $E(X^2)$.

Then: $E(X(X-1)) = E(X^2) - E(X)$, hence $E(X^2) = E(X(X-1)) + E(X)$ (2)

$$\begin{aligned}
 E(X(X-1)) &= \sum_{k=0}^{\infty} k(k-1) e^{-\lambda} \frac{\lambda^k}{k!} = e^{-\lambda} \sum_{k=2}^{\infty} \frac{\lambda^k}{(k-2)!} = e^{-\lambda} \lambda^2 \sum_{k=2}^{\infty} \frac{\lambda^{k-2}}{(k-2)!} \\
 &= e^{-\lambda} \lambda^2 \sum_{l=0}^{\infty} \frac{\lambda^l}{l!} = e^{-\lambda} \lambda^2 e^{\lambda} = \lambda^2 \quad (3)
 \end{aligned}$$

(1),(2),(3)

$$\Rightarrow \text{Var}(X) = \lambda^2 + \lambda - \lambda^2 = \lambda$$

An approximation result:

Link between $\text{Bin}(n, p)$ and $\text{Pois}(\lambda)$

Thm 2.4: If $X_n \sim \text{Bin}(n, p_n)$ such that $\left. \begin{array}{l} p_n \rightarrow 0 \\ \text{and} \\ mp_n \rightarrow \lambda \\ (\text{as } n \rightarrow \infty) \end{array} \right\}$ Then $P(X_n = k) \rightarrow e^{-\lambda} \frac{\lambda^k}{k!}$
for all $k = 0, 1, \dots, n$
as $n \rightarrow \infty$

Ex 2.7 #38 $P(\text{at least 1 doubly 6 in 24}) = 1 - P(\text{no doubly 6 in 24})$
 $= 1 - \left(\frac{35}{36}\right)^{24} \approx \underline{\underline{0.4914}}$ (exact)

Poisson approx. = $1 - P(X=0)$, where $X \sim \text{Pois}(\lambda = np = \frac{24}{36} = \frac{2}{3})$
 $= 1 - e^{-\frac{2}{3}} \cdot \frac{\left(\frac{2}{3}\right)^0}{0!} = \underline{\underline{0.4866}}$ pretty close, as expected.

Sec. 2.7 # 52

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61

Capture-recapture experiments (Ex 2.42 on p 61)

Q: A population with unknown number of individuals.

Can we estimate this number based on a capture-recapture experiment:

So (k, r, m) are known.

- day 1: capture k individuals and mark them all
- day 2: capture r individuals of which m turn out to be marked.

A: Main idea is maximum likelihood estimate:If there were N individuals, then the probability of capturing m marked individuals on day 2 is:

$$P_N = \frac{C_{k,m} C_{N-k, r-m}}{C_{N,r}}$$

Think of P_N as a function of N (k, r, m are known), and find that value of N which maximizes P_N .So let's maximize P_N : First note $C_{p,q} = \frac{p!}{(p-q)!q!} = \frac{p \cdot (p-1)!}{(p-q)(p-1-q)!q!}$

$$\text{So } C_{p,q} = \frac{p}{p-q} C_{p-1,q} = C_{p-1,q}$$

$$\text{Thus, } P_N = C_{k,m} \frac{C_{N-k, r-m}}{C_{N,r}} = P_{N-1} \frac{N-k}{N-k-(r-m)} \cdot \frac{N-r}{N}$$

$$\Rightarrow P_N \geq P_{N-1} \text{ iff } (N-k)(N-r) \geq (N-k-(r-m))N \text{ iff } N \leq \frac{kr}{m}$$

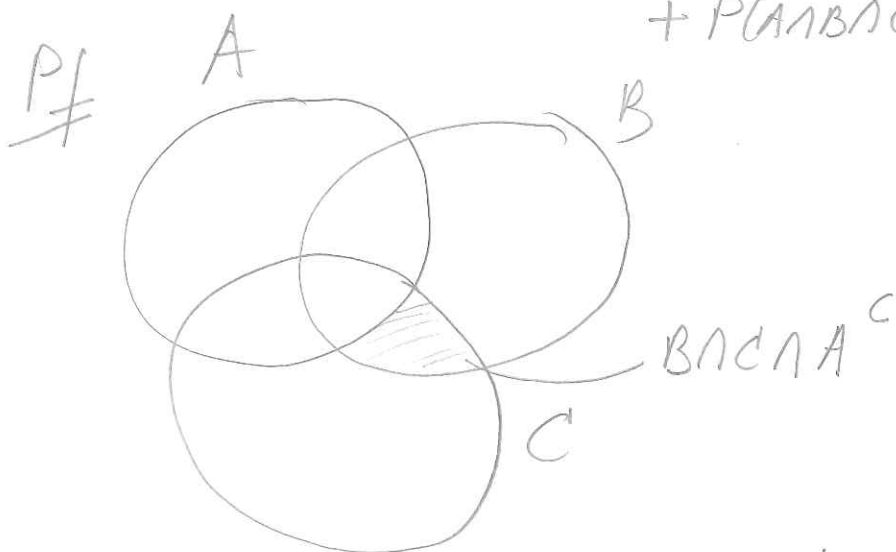
$$\Rightarrow P_N \text{ is maximized for } N = \lfloor \frac{kr}{m} \rfloor \quad \text{Do numerical example.}$$

- Finish Capture - Recapture experiment from last class.

- Inclusion - Exclusion generalized

↳ For 2 events: $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

For 3 events: $P(A \cup B \cup C) = P(A) + P(B) + P(C) - (P(A \cap B) + P(A \cap C) + P(B \cap C)) + P(A \cap B \cap C)$



There are 7 parts to $A \cup B \cup C$. Check how often each of these are counted in expression above.

$A \cup B \cup C$ is disjoint union of:	A	B	C	$A \cap B$	$A \cap C$	$B \cap C$	$A \cap B \cap C$	Net count
$A \cap B \cap C$	+1	+1	+1	-1	-1	-1	+1	(+1)
$A \cap B \cap C^c$								
$A \cap C \cap B^c$								
$B \cap C \cap A^c$								
$A \cap (B \cup C)^c$								
$B \cap (A \cap C)^c$								
$C \cap (A \cap B)^c$								

- This counting principle can be extended to n events

$$P\left(\bigcup_{i=1}^n A_i\right) = \sum_{i=1}^n P(A_i) - \sum_{i < j} P(A_i \cap A_j) + \sum_{i < j < k} P(A_i \cap A_j \cap A_k) - \dots + (-1)^{n+1} P\left(\bigcap_{i=1}^n A_i\right)$$