

*“I am the very model of a modern Major-General;
I’ve information vegetable, animal and mineral;
I know the Kings of England, and I quote the fights historical,
From Marathon to Waterloo, in order categorical.”*

William S. Gilbert, English Lyricist, 1836-1911

COHEN CHAP 19 & 20. CATEGORICAL

For EDUC/PSY 6600

MOTIVATING EXAMPLES

Dr. Fisel wishes to know whether a random sample of adolescents will prefer a new of formulation of 'JUMP' softdrink over the old formulation. The proportion choosing the new formulation is tested against a hypothesized value of 50%.

Dr. Sheary hypothesizes that $1/3$ of women experience increased depressive symptoms following childbirth, $1/3$ experience increases in elevated mood after childbirth, and $1/3$ experience no change. To evaluate this hypothesis Dr. Sheary randomly samples 100 women visiting a prenatal clinic and asks them to complete the Beck Depression Inventory. She then re-administers the BDI to each mother one week following the birth of her child. Each mother is classified into one of the 3 previously mentioned categories and observed proportions are compared to the hypothesized proportions.

Dr. Evanson asks a random sample of individuals whether they see both a physician and a dentist regularly (at least once per year). He compares the distributions of these binary variables to determine whether there is a relationship.

CATEGORICAL METHODS

Instead of means, comparing **counts** and **proportions** within and across groups

- E.g., # ill across different treatment groups

Associations / dependencies among categorical variables

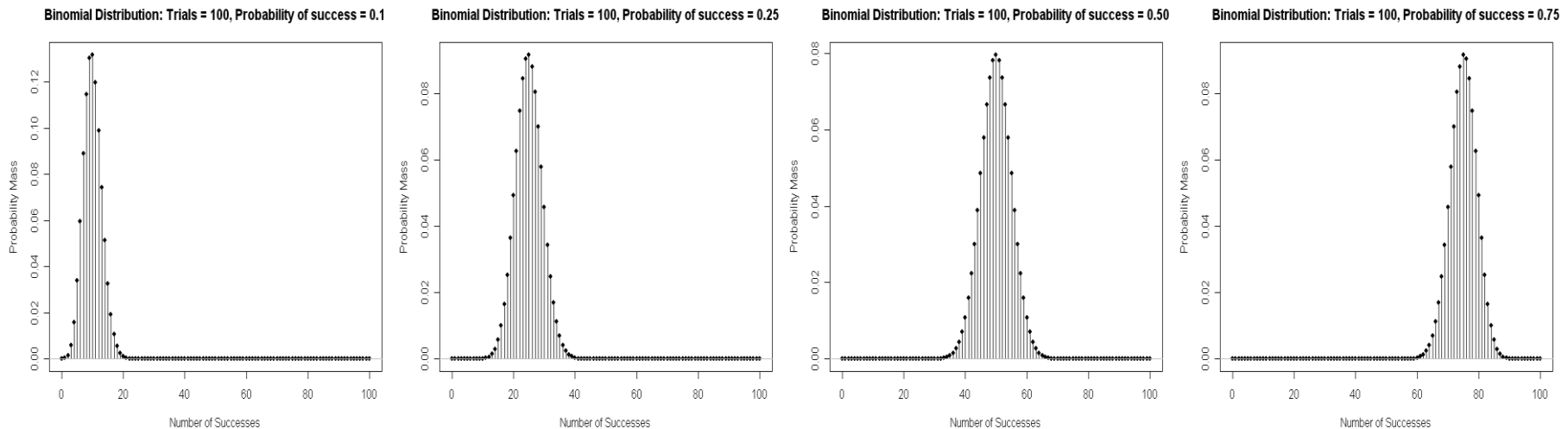
Data are **nominal** or **ordinal**

Discrete probability distribution

- Number of finite values as opposed to infinite

Each subject/event assumes 1 of 2 **mutually exclusive** values (binary or dichotomous)

- Yes/No
- Male/Female
- Well/Ill



THE BINOMIAL DISTRIBUTION: EQ & COIN EXAMPLE

$$p(X) = \frac{N!}{X!(N-X)!} P^X Q^{(N-X)}$$

- $N = \# \text{ events}$
- $X = \# \text{ “successes”}$
- $P = p(\text{“success”})$
 - Hypothesized proportion / probability of success
- $Q = p(\text{“failure”})$
 - Hypothesized proportion / probability of failure
- $P + Q = 1$
- Remember: $0! = 1$; $x^0 = 1$

- (Arbitrarily) assign 1 outcome as ‘success’ and other as ‘failure’
- **Example: Probability of correctly guessing side of coin 4 out of 5 flips?**
 - 5 events, 4 successes, 1 failure
 - $P = p(\text{correct guess on each flip}) = .50$
 - $Q = p(\text{incorrect guess on each flip}) = .50$

Use equation to obtain:

5 out of 5 successes = .03

4 out of 5 successes = .16

3 out of 5 successes = .31

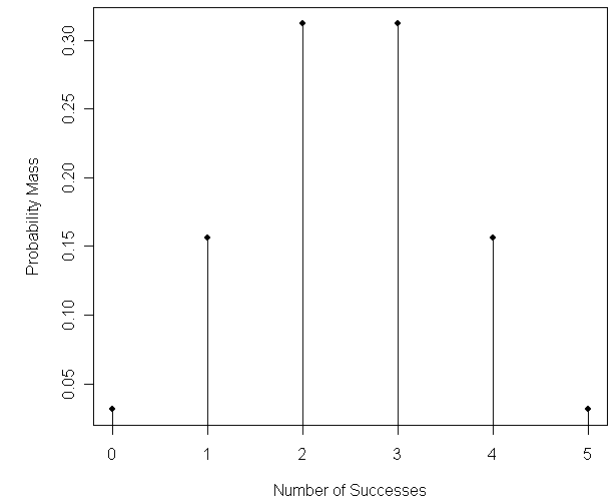
2 out of 5 successes = .31

1 out of 5 successes = .16

0 out of 5 successes = .03

Sum of probabilities = 1.0

Binomial Distribution: Trials = 5, Probability of success = 0.5



SAMPLING DISTRIBUTION FOR THE BINOMIAL DIST

- Binomial probability distribution for $N = 5$ events, and $P = .5$
- Binomial Distribution Table (exact values)
- Sampling distribution as it was derived mathematically
 - We can only reject H_0 with 0 or 5 out of 5 successes (1-tailed)

Sampling Distribution

$$\text{mean} = NP$$

$$\text{variance} = NPQ$$

$$SD = \sqrt{NPQ}$$

$$SE_{MEAN} = \sqrt{\frac{PQ}{N}}$$

Example

$$M = 5 * .5 = 2.5 \text{ (See Histogram)}$$

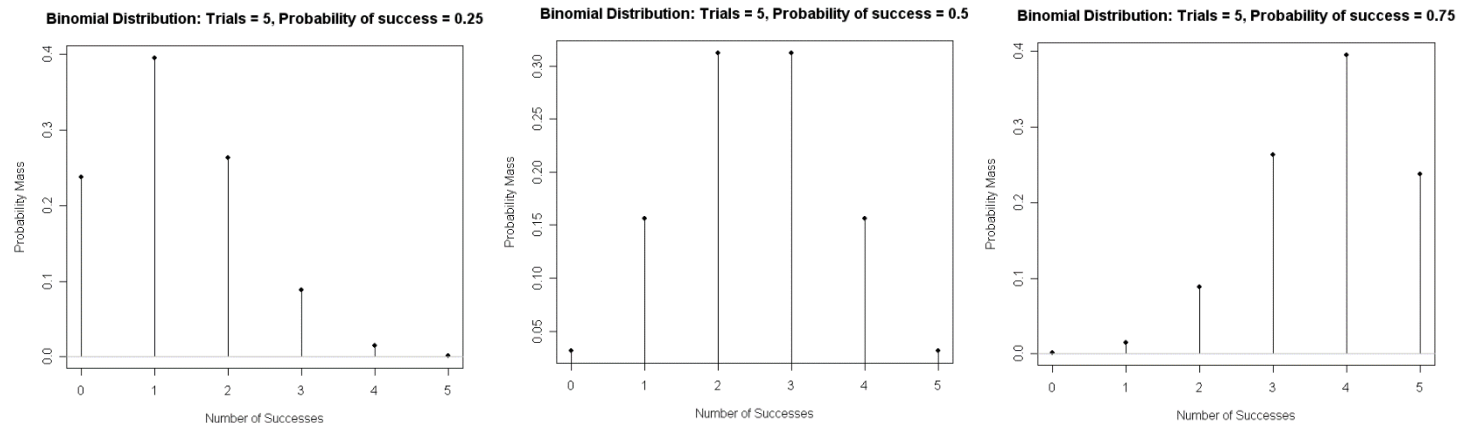
$$VAR = 5 * .5 * .5 = 1.25$$

$$SD = \text{sqrt}(1.25) = 1.12$$

Different binomial distribution for each N

Normal when $P = .50$, skewed when $P \neq .50$

Critical value depends on: N events, X successes, P



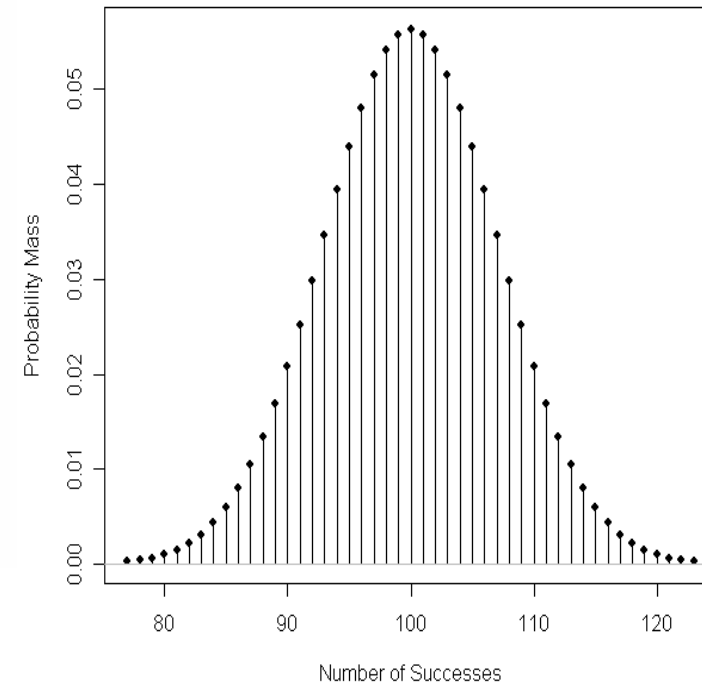
AS N INCREASES, BINOMIAL DISTRIBUTION \rightarrow NORMAL

n	X	p	n	X	p	n	X	p
1	0	.5000		1	.0176	13	0	.0001
	1	.5000		2	.0703		1	.0016
2	0	.2500		3	.1641		2	.0095
	1	.5000		4	.2461		3	.0349
	2	.2500		5	.2461		4	.0873
3	0	.1250		6	.1641		5	.1571
	1	.3750		7	.0703		6	.2095
	2	.3750		8	.0176		7	.2095
	3	.1250		9	.0020		8	.1571
4	0	.0625	10	0	.0010		9	.0873
	1	.2500		1	.0098		10	.0349
	2	.3750		2	.0439		11	.0095
	3	.2500		3	.1172		12	.0016
	4	.0625		4	.2051		13	.0001
5	0	.0312		5	.2461	14	0	.0001
	1	.1562		6	.2051		1	.0009
	2	.3125		7	.1172		2	.0056
	3	.3125		8	.0439		3	.0222
	4	.1562		9	.0098		4	.0611
	5	.0312		10	.0010		5	.1222

Table A.13
Probabilities of the
Binomial Distribution for
 $P = .5$

"Equally Likely"
Means $p = 0.5$

Binomial Distribution: Trials = 200, Probability of success = 0.5



BINOMIAL SIGN TEST

Single sample test with binary/dichotomous data

Proportion or % of 'successes' differ from chance?

- H_0 : % of observations in one of two categories equals a **specified** % in population
 - H_0 : Proportion of 'yes' votes = 50% in population

Assumptions

- Random selection of events or participants
- Mutually exclusive categories
- Probability of each outcome is same for all trials/observations of experiment

- Experiment: Coin flipped 10x, heads 8x
 - Is coin biased (Heads > .50)?
- Experiment: 10 women surveyed, 8 select perfume A
 - Is one perfume preferred over another?
- For both:
 - H_0 : Proportion (X) = .50 in population
 - H_1 : Proportion (X) \neq .50 in population (2-tailed)

BINOMIAL SIGN TEST: EXAMPLE

- Is occurrence of 8 or more observations in either of the 2 categories unusual?
 - Probability of occurrence given H_0 true in pop.?

n	X	p
	1	.0176
	2	.0703
	3	.1641
	4	.2461
	5	.2461
	6	.1641
	7	.0703
	8	.0176
	9	.0020
	10	.0010
	1	.0098
10	2	.0439
	3	.1172
	4	.2051
	5	.2461
	6	.2051
	7	.1172
	8	.0439
	9	.0098
	10	.0010

NORMAL APPROXIMATION TO THE BINOMIAL I.E. “Z-TEST” FOR A SINGLE PROPORTION

What if N were larger, say 15?

- Same proportions: 80% (12/15) Heads & Perfume A
- Sum $p(12, 13, 14, 15/15) = .0178$ (1-tailed p -value)

Reject H_0 under both 1- and 2-tailed tests

- 2-tailed $p = .0178 \times 2 = .0356$

- Earlier: Binomial distribution \rightarrow normal distribution, as $N \rightarrow$ infinity
- Recommendation: Use z -test for single proportion when N is large (>25 -30)
 - When NP and NQ are both > 10 , close to normal
- H_0 and H_1 are same as Binomial Test
- Test statistic:

$$z = \frac{X - PN}{\sqrt{NPQ}} = \frac{p_1 - P}{\sqrt{\frac{PQ}{N}}}$$

Experiment:

Senator supports bill favoring stem cell research. However, she realizes her vote could influence whether or not her constituents endorse her bid for re-election. She decides to vote for the bill only if 50% of her constituents support this type of research. In a random survey of 200 constituents, 96 are in favor of stem cell research.

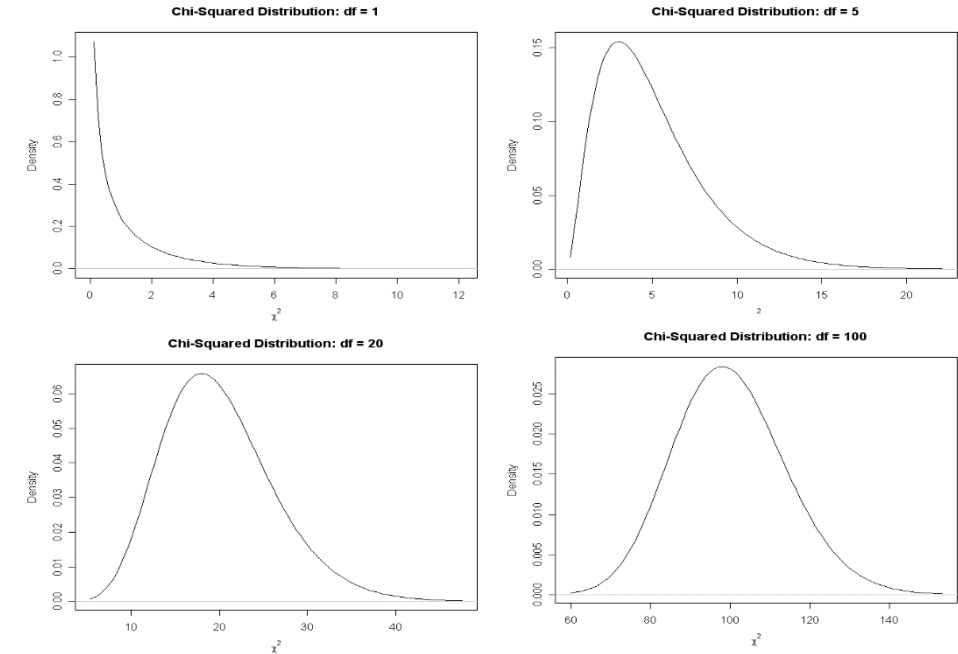
Will the senator support the bill?

Standardize:

$$z = \frac{X - NP}{\sqrt{NPQ}}$$

CHI-SQUARE (χ^2) DISTRIBUTION

- Family of distributions
 - As df (or k categories) \uparrow
 - Distribution becomes more normal, bell-shaped
 - Mean & variance \uparrow
 - Mean = df
 - Variance = $2 * df$
- $\chi^2 = \chi^2$
 - Always positive, 0 to infinity
 - 1-tailed distribution
- χ^2 distribution used in many statistical tests



“GOODNESS OF FIT” Testing:

Are observed frequencies **similar** to frequencies expected by chance?

Expected frequencies

Frequencies you'd expect if H_0 were true
Usually equal across categories of variable (N / k)
Can be unequal if theory dictates

CHI-SQUARED: GOODNESS OF FIT TESTS “GOF”

Hypotheses

- H_0 : Observed = Expected frequencies in population
- H_1 : Observed \neq Expected frequencies in population

General form:

- O = observed frequency
- E = expected frequency

If H_0 were true, numerator would be small

Denominator standardizes difference in terms of expected frequencies

Aka: Pearson or ‘1-way’ χ^2 test

- 1 nominal variable
- 2 or more categories

$$\chi^2 = \sum \frac{(O_i - E_i)^2}{E_i}$$

If nominal variable ONLY has 2 categories, χ^2 GoF test:

- Is another large sample approximation to Binomial Sign Test
- Gives same results as z -test for single proportion as $z^2 = \chi^2$
- Has same H_0 and H_1 as binomial or z -tests

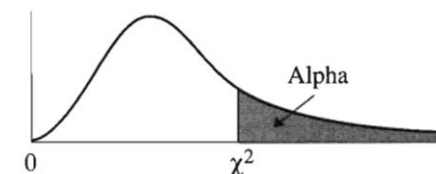
Compare obtained χ^2 statistic to critical value based on $df = \underline{k - 1}$, $k = \#$ categories

Assumptions

Independent random sample

Mutually exclusive categories

Expected frequencies: ≥ 5 per each cell



df	ALPHA (AREA IN THE UPPER TAIL)				
	.10	.05	.025	.01	.005
1	2.71	3.84	5.02	6.63	7.88
2	4.61	5.99	7.38	9.21	10.60
3	6.25	7.81	9.35	11.35	12.84
4	7.78	9.49	11.14	13.28	14.86
5	9.24	11.07	12.83	15.09	16.75
6	10.64	12.59	14.45	16.81	18.55
7	12.02	14.07	16.01	18.48	20.28
8	13.36	15.51	17.54	20.09	21.96
9	14.68	16.92	19.02	21.67	23.59
10	15.99	18.31	20.48	23.21	25.19
11	17.28	19.68	21.92	24.72	26.75
12	18.55	21.03	23.34	26.22	28.30
13	19.81	22.36	24.74	27.69	29.82
14	21.06	23.69	26.12	29.14	31.32
15	22.31	25.00	27.49	30.58	32.80
16	23.54	26.30	28.85	32.00	34.27
17	24.77	27.59	30.19	33.41	35.72

GOODNESS OF FIT TESTS — EXAMPLE: $K = 2$

Hypotheses:

- $H_0: P = 0.50$
- Observed frequencies: 96 and 104
- Expected frequencies: $N / k = 200 / 2 = 100$ $df = 2 - 1 = 1$

Test Statistic:

$$\chi^2_{OBSERVED} =$$

$$\chi^2 = \sum \frac{(O_i - E_i)^2}{E_i}$$

Critical Value:

$$\chi^2_{CRIT} (__) =$$

Conclusion:

Note:

ALWAYS USE COUNTS!!!	1 = “success”	0 = “failure”
OBSERVED (the data)	96	
EXPECTED (based on N, P, Q)		

GOODNESS OF FIT TESTS — EXAMPLE: $K > 2$

(ANY NUMBER OF CATEGORIES WITHIN 1 VARIABLE)

ALWAYS USE COUNTS!!!

Hypotheses:

- H_0 : “equally likely” ($k = 6$ & $N = 120$)
- Expected frequencies: $N / k = 120 / 6 = 20$
- Observed frequencies: 20, 14, 18, 17, 22, 29 {Mon – Sat}
- $df = 6 - 1 = 5$

	M	T	W	Th	F	S
OBS	20	14	18	17	22	29
EXP						

Test Statistic:

$$\chi^2_{OBSERVED} =$$

Critical Value:

$$\chi^2_{CRIT}(\text{---}) =$$

Conclusion:

We do NOT have evidence the # of books checked out is NOT the same EVERY day

QUESTION:

Is there a difference
in # books checked
out for different days
of the week?

GOODNESS OF FIT TESTS: CONFIDENCE INTERVALS

- **CIs for proportions**
 - If $k > 2$, original table converted into table with 2 cells
 - Proportion for category of interest vs proportion in **all other** categories
 - Use same formula for z -test for single proportion:
- **Say we wanted a CI for proportion of books from Saturday ($29/120=0.242$)**

$$P_{obs} \pm z_{crit} \times \sqrt{\frac{P_{obs} \times Q_{obs}}{N}}$$

GOODNESS OF FIT TESTS: EFFECT SIZE

$$\chi^2_{\text{Effect Size}} = \frac{\chi^2}{N(k-1)}$$

Ranges from 0 to 1

- 0: Expected = Observed frequencies exactly
- 1: Expected \neq Observed frequencies as much as possible

GOODNESS OF FIT TESTS: POST HOC PAIRWISE TESTS

Like ANOVA, omnibus test, but where do differences lie?

- ‘Pinpointing the action’ in contingency tables
- Post-hoc Binomial, z-tests, or smaller 1-way χ^2 tests
 - Collapsing, ignoring levels
 - Bonferonni correction, more conservative α per comparison
- Examining
 - Observed vs. expected frequencies per cell
 - Contributions to χ^2 per cell
- Visual analysis of differences in proportions

2-WAY PEARSON χ^2 TEST OF “INDEPENDENCE” OR “ASSOCIATION”

Aka: Contingency table, cross-tabulation, or *row x column* ($r \times c$) analysis

- > 1 nominal variable

Is distribution of 1 variable *contingent* on distribution of another?

- Is there an association or dependence between 2 categorical variables

Extension of χ^2 Goodness of Fit Test

Hypotheses:

- H_0 : Variables are independent in population
- H_1 : Variables are dependent in population

Again, χ^2_{obt} is compared with χ^2_{crit} $\rightarrow df = (r-1)(c-1)$

2-WAY PEARSON χ^2 TEST OF “INDEPENDENCE” OR “ASSOCIATION”

Same equation: Standardized squared deviations summed for all cells

$$\chi^2 = \sum \frac{(O_{ij} - E_{ij})^2}{E_{ij}}$$

Different method for computing E

- For each cell: Multiply corresponding row and column totals (marginals), divide by N

$$E_{Cell_A} = \frac{(a+b)(a+c)}{N}$$

	Var1		
Var2	a	b	a + b
	c	d	c + d
	a + c	b + d	a + b + c + d = N

$$EXP_{cell} = \frac{Total_{row} \times Total_{column}}{Total_{grand}}$$

χ^2 TEST OF “INDEPENDENCE” — EXAMPLE:

Experiment:

Random sample of 200 inmates are surveyed about abuse and violent criminal histories

- Relationship between history of abuse and violent crime?

H_0 : **No association** between abuse history and violent criminal history in population of prison inmates

- $O_{ij} = E_{ij}$ for all cells in population

H_1 : **Association** between abuse history and violent criminal history in population of prison inmates

- $O_{ij} \neq E_{ij}$ for at least one cell in population

Observed frequencies

Abuse	Violent Crime		Row Sum
	Yes	No	
Yes	70	30	100
No	40	60	100
Column Sum	110	90	200

Expected frequencies:

Test Statistic:

APA format: