6. CHANCE VARIABILITY (FPP, Ch 16,17,18, etc.) 6.1 SAMPLING VARIABILITY

- When we toss a (fair) coin: we do not get the same result every time.
- When we toss a fair coin 100 times, we probably won't get exactly 50 heads, but we will get about 50% heads.
- \bullet When we toss a fair coin 1000 times, we would be very surprised to get EXACTLY 500 heads. But we will get very close to 50% heads.
- When we take a random sample from a population, we do not get the same sample every time. The results will be a bit different.
- If we (could) repeat a randomized controlled experiment: different subjects would be randomized to treatment/control. The results will be a bit different.
- If have someone measure our own height to 0.01 inches, it will not be the same every time. Part of this is measurement error, part may be true variation in the way we stand, the time of day,

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- NOT ALL VARIATION IS ERROR.
 - That is, NOT A "MISTAKE"

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6.2 COUNTING THE WAYS

• Suppose 1000 people each toss a fair coin three times

Heads	Heads	Heads 125
500	250	Tails 125
(50%)	Tails	Heads 125
	250	Tails 125
Tails	Heads	Heads 125
500	250	Tails 125
(50%)	Tails	Heads 125
	250	Tails 125

- On average, 125 get 3 heads, 125 get 0 heads, On average, 375 get 2 heads, 375 get 1 head
- There are 3 ways to get 2 heads: HHT, HTH, THH
 There is only 1 way to get 3 heads: HHH
- Suppose 1000 people each toss a fair coin 10 times. About 2 (0.2%) will get heads the first 9 times.
- 50% of these 2 (1 person) will get heads on 10 th toss. The other 50% of the 2 (1 person) will get tails. This person has 9 heads, but there are many other ways to get 9 heads
- \bullet Roughly 25% will get 5 heads: there are MANY ways to get 5 heads.

- Having 9 heads already does not change the chance that the 10 th toss is heads.
- It is still a fair coin.

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6.2 THE LAW OF AVERAGES: FPP Ch 16

• Back to the fair coin, tossed many times. On average, we will get 50% heads.

number of	within 10 o	m f~50%	40% to $60%$ of tosses		
tosses	number	chance	number	chance	
1	0 or 1	100%		— <u>-</u>	
10	0 to 10	100%	4 to 6	$\boldsymbol{66\%}$	
50	15 to 35	99.7%	20 to 30	88%	
100	40 to 60	96%	40 to 60	96%	
1,000	490 to 510	49%	400 to 600	$\sim 100\%$	
10,000	4990 to 5010	$\boldsymbol{16\%}$	4000 to 6000	100%	

- As number of tosses goes up: the chance of being within a given number of expected goes down within a given percent of expected goes up
- FPP calls the difference between observed and expected the chance error

The chance error in number of heads goes up The chance error in proportion of heads goes down

- The LAW OF AVERAGES says

 The chance error in proportion of heads goes down
 as the number of tosses goes up.
- The LAW OF AVERAGES does NOT say because we had more heads, the chance of getting a head goes down

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6.3 BOX MODELS: FPP Ch 16

• Many chance models can be most easily thought of as drawing tickets repeatedly from a box.

Put the tickets back each time!!

- Tossing a fair coin: counting number of heads
 2 tickets: we draw with replacement
 one labeled "0" (tails), one labeled "1" (heads)
 Number of tosses = number of draws.
 Number of heads = sum of values on the tickets.
- Tossing an unfair coin; suppose coin gives 20% heads box of 100 tickets: 20 with "1" and 80 with "0" box of 10 tickets: 2 with "1" and 8 with "0" box of 5 tickets: 1 with "1" and 4 with "0".

 Number of heads = sum of values on the tickets.
- Same unfair coin:

suppose a head wins me \$5: tail loses \$1. box of 5 tickets: 1 with "5", 4 with "-1"

Number of tosses = number of draws from box.

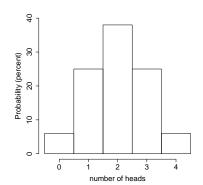
Total winnings = sum of values on the tickets.

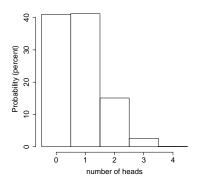
• We don't have to be tossing coins:
Chance baby is a boy: just about 50%
Chance of correct answer in 5 choices: 20%
Colors of M-and-M's: 10% orange, 30% green,
50% yellow, 10% brown: Box of 10 tickets
1 orange, 3 green, 5 yellow, 1 brown.

6.4 PROBABILITY HISTOGRAMS: FPP Ch 18.2

- A histogram of sample values represents the percents by areas: the total area is 100%.
- A probability histogram represents the chances of outcomes by areas: the total area is 100%.
- Toss a fair coin 4 times: we could get 0, 1, 2, 3, 4 heads. The chances of these 5 possibilities are 6.25%, 25%, 37.5%, 25%, and 6.25%.
- What does this mean?

 Percent of times something happens in many, many repetitions.
- If the coin has only 20% chance of showing heads, the chances are 41%, 41%, 15%, 3%, 0.1%.
- The probability histograms are





6.5 EXPECTED VALUES AND STANDARD ERRORS FPP Ch 17

- Suppose the values on the tickets are quantitative.
- Suppose a large number of people each make one draw from the box, with replacement. On average, the value of their ticket is the average of the ticket values in the box: the box average.
- This "on average" value is the expected value
- Suppose we make some number of draws from the box, with replacement, and add them up expected value = (number of draws) × (box-average)
- But probably we will not get exactly the expected value there is chance variation.
- The difference of our value from the expected value is the chance error.

How big do we expect the chance error to be? Answer: the <u>standard error</u> (SE)

- For a sum of draws from a box: $SE = \sqrt{\text{number of draws}} \times (SD \text{ of box}).$
- 100 times as many draws: SE multiplied by only 10
- If SD of box is large, SE of sum will be large.

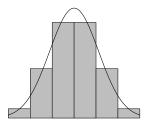
6.6 CHANCE VARIATION IN COUNTS

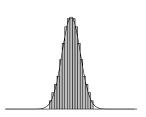
- For counts: we have only two types of tickets "1" and "0".
- For a box with only "1" and "0" tickets SD of box = $\sqrt{\text{fraction of "1"} \times \text{fraction of "0"}}$
- For the fair coin: one "1" and one "0" On single draw: expectation = 0.5 SE = (SD of BOX) = $\sqrt{(1/2) \times (1/2)} = 1/2 = 0.5$
- For the 20% heads coin: one "1" and four "0" On single draw: expectation = 0.2 SE = (SD of BOX) = $\sqrt{(0.2 \times 0.8)}$ = 0.4

	Fair coin			20% heads coin		
\mathbf{number}	exp-	\mathbf{SE}	expected	exp-	\mathbf{SE}	expected
of tosses	ected		\pm 2 SE	ected		\pm 2 SE
1		0.5			0.4	
10	5	1.58	2 to 8	2	1.26	0 to 4
50	25	3.53	18 to 32	10	2.82	5 to 15
100	50	5.00	40 to 60	20	4.00	12 to 28
1,000	500	15.8	469 to 531	200	12.6	175 to 225
10,000	5000	50.0	4900 to	2000	40.0	1920 to
			5100			2080

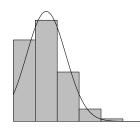
6.7 THE NORMAL APPROXIMATION FOR PROBABILITIES

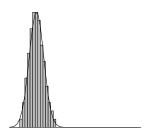
• Toss a fair coin: 5 times, 50 times



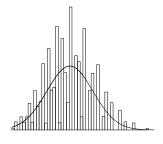


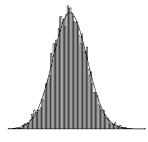
 \bullet Toss a coin with chance 20% heads: 5, 50 times





• Box with 4 tickets, values 0, 2, 5, 10: sum 5 draws, 50 draws





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6.8 THE CENTRAL LIMIT THEOREM: FPP Ch 18

- Fair coin: box mean = 0.5, box SD = 0.5 sum of 5 draws, expected = 2.5, SE = 1.12 sum of 50 draws, expected = 25, SE = 3.54
- 20% heads coin: box mean = 0.2, box SD = 0.4 sum of 5 draws, expected = 1.0, SE = 0.89 sum of 50 draws, expected = 10, SE = 2.83
- 4-ticket box: values 0, 2, 5, 10: box mean 4.25, SD=3.77 sum of 5 draws, expected = 21.25, SE = 8.42 sum of 50 draws, expected = 212.5, SE = 26.63
- As number of draws increases, the probability histogram FOR THE SUM always gets closer to normal shape, with

normal distribution mean = expected normal distribution SD = SE

- So we can <u>standardize</u> the values using z-score: Recall for population histograms: (value-mean)/SD Now for probability histograms: (value - expected)/SE
- The probability histogram for the standardized value of the sum gets close to the <u>standard normal curve</u>.
- So we can use the table on FPP, P.A104.

6.9 SAMPLING POPULATIONS and DRAWS FROM BOXES: FPP Ch 20

• When we sample subjects from a population, we observe a value or characteristic associated with each:

height of an individual income of a household miles per year driven by a car (or driver) whether voter will vote "D" or "R" whether vehicle is SUV (yes/no questions)

- When we do repeated draws from a box, we observe the "value" on the ticket.
- From a population we sample without replacement. From a box we draw with replacement.

 But for a large population it makes no difference
- Population histograms give us the distribution of incomes in the population: for example, the percentage in each \$10K interval.
- \bullet Now make a box with 100 tickets, and label the right proportion with each \$10K interval. For example, if 8% of household have incomes \$50K to \$59K, label 8 of the 100 tickets "\$50K to \$59K".
- Repeated draws for the box <u>with replacement</u> is just like sampling from the population. The probability histogram for the box is like the population or sample histogram (in intervals of \$10K).
- We can use our box models to find out what samples from the population will look like means, SD, etc.

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6.10 CHANCE VARIATION IN PROPORTIONS: FPP Ch 21

• In a population some percentage will have a given characteristic of interest. For example, 55% will vote Democrat.

- We take a sample of 3000 (say) registered voters.
- Taking a sample size n=3000 from a large population (without replacement) is <u>almost</u> like taking n=3000 draws from a box.
- A population with 55% people who will vote Democrat, is like a box, with 100 tickets, 55% marked "D" or "1", 45% marked "R" or "0".
- Or, use just 20 tickets: 11 with "1", 9 with "0".
- In 6.2, the <u>law of averages</u> showed us that while <u>chance error</u> in count got larger, the chance error in proportion got smaller, as the number of draws gets larger.
- The SE for the proportion of 1's in n draws = $\sqrt{n} \times (SD \text{ of box})/n = (SD \text{ of box})/\sqrt{n}$.
- For our example: SD of box = $\sqrt{0.55 \times 0.45}$ =0.497 For 3000 draws; SE of proportion = 0.497/ $\sqrt{3000}$ = 0.009. Or just under 1%
- So, in sampling: we expect 55% of "D" tickets and SE is just about 1% if we sample 3000 voters.
- The normal distribution works for us as before: 95% of the time our <u>chance error</u> is less than 2 SE, or 2 percentage points.

6.11 CHANCE VARIATION IN AVERAGES: FPP Ch 23

- For draws from a box: sum the ticket values
 expected sum = (number of draws) × (box average)
 SE of sum = √number of draws× (box SD)
- Now take average of the ticket values drawn from box: expected average = (box average) SE of average = SE for sum / (number of draws)

 $= (\text{box SD}) / \sqrt{\text{number of draws}}$

• As before: the normal distribution curve can be used to figure the chances for the average.

In 95% of repetitions, average is within 2 SE of box average. In 68% of repetitions, average is within 1 SE of box average

- We take a sample from a population to find out about characteristics of the population for example, average household income.
- Example: sample 1000 households from 100,000 in city sample average should be about population (box) average: it differs by the <u>chance error</u>

SE of average = $(SD \text{ of box})/\sqrt{1000}$ SD of sample should be about the population SD

• As before: the normal distribution curve can be used to figure the chances for the average.

95% of repetitions, sample average is within 2 SE of population average

68% of repetitions, sample average is within 1 SE of population average