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


### 6.2 COUNTING THE WAYS

- Suppose 1000 people each toss a fair coin three times

| $\begin{gathered} \text { Heads } \\ 500 \\ (50 \%) \end{gathered}$ | Heads | Heads 125 |
| :---: | :---: | :---: |
|  | 25 | Tails 125 |
|  | T | Heads 125 |
|  | 250 | Tails 125 |
| $\begin{gathered} \text { Tails } \\ 500 \\ (50 \%) \end{gathered}$ | Heads | Heads 125 |
|  | 250 | Tails 125 |
|  | Tai | Heads 125 |
|  | 25 | Tails 125 |

- 125 get 3 heads, 125 get 0 heads,

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

- There is only 1 way to get 10 heads: HННННННННH

There are 10 ways to get 9 heads and 1 tail
In all about 10 people will get 9 heads.

- 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.

Rest of this page for your notes

- Back to the fair coin, tossed many times.

On average, we will get $50 \%$ heads.

| number of | within 10 of $50 \%$ |  | $40 \%$ to $60 \%$ of tosses |  |
| :---: | :---: | :---: | :---: | :---: |
| tosses | number | chance | number | chance |
| 1 | 0 or 1 | $100 \%$ | - | - |
| 10 | 0 to 10 | $100 \%$ | 4 to 6 | $66 \%$ |
| 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 | $16 \%$ | 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
- 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 proportions 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?

Proportion 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 $) \times($ 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 standard error (SE)
- For a sum of draws from a box:
$\mathrm{SE}=\sqrt{\text { number of draws }} \times(\mathrm{SD}$ of box).
- 100 times as many draws: SE multiplied by only 10
- If SD of box is large, SE of sum will be large.
- 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 $=0.5$

$$
\mathrm{SE}=(\mathrm{SD} \text { of } \mathrm{BOX})=\sqrt{(1 / 2) \times(1 / 2)}=1 / 2=0.5
$$

- For the $20 \%$ heads coin: one " 1 " and four " 0 "

On single draw: expectation $=0.2$

$$
\mathrm{SE}=(\mathrm{SD} \text { of } \mathrm{BOX})=\sqrt{(0.2 \times 0.8)}=\mathbf{0 . 4}
$$

|  | Fair coin |  |  | $20 \%$ heads coin |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| number <br> nof tosses | exp- <br> exted | SE | expected <br> $\pm 2$ SE | exp- <br> ected | SE | expected <br> $\pm 2 \mathrm{SE}$ |
| 1 | - | 0.5 | - | - | 0.4 | - |
| 10 | 5 | 1.58 | 2 to 8 | 2 | 1.26 | 0 to 5 |
| 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

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

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

- Fair coin: box mean $=0.5$, box $\mathrm{SD}=0.5$
sum of 5 draws, expected $=2.5, \mathrm{SE}=1.12$
sum of 50 draws, expected $=25, \mathrm{SE}=3.54$
- $20 \%$ heads coin: box mean $=0.2$, box $\mathrm{SD}=0.4$
sum of 5 draws, expected $=1.0, \mathrm{SE}=0.89$
sum of 50 draws, expected $=10, \mathrm{SE}=2.83$
- 4-ticket box: values $0,2,5,10$ :
box mean $4.25, \mathrm{SD}=3.77$
sum of 5 draws, expected $=21.25, \mathrm{SE}=8.42$
sum of 50 draws, expected $=212.5, \mathrm{SE}=26.63$
- 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 $=\mathrm{SE}$
- So we can standardize 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 standard normal curve.
- So we can use the table on FPP, P.A105.
- 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 $\$ 10 \mathrm{~K}$ interval.
- Now make a box with 100 tickets, and label the right proportion with each $\$ 10 \mathrm{~K}$ interval. For example, if $8 \%$ of household have incomes $\$ 50 \mathrm{~K}$ to $\$ 59 \mathrm{~K}$, label 8 of the 100 tickets $" \$ 50 \mathrm{~K}$ to $\$ 59 \mathrm{~K} "$.
- Repeated draws for the box with replacement is just like sampling from the population. The probability histogram for the box is like the population or sample histogram (in intervals of $\$ 10 \mathrm{~K}$ ).
- We can use our box models to find out what samples from the population will look like - means, SD, etc.
- 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 $\mathrm{n}=3000$ from a large population (without replacement) is almost like taking $\mathrm{n}=3000$ 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 law of averages showed us that while chance error 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 $=$ $\sqrt{n} \times(\mathrm{SD}$ of box $) / \mathrm{n}=(\mathrm{SD}$ of box $) / \sqrt{n}$.
- For our example: SD of box $=\sqrt{0.55 \times 0.45}=\mathbf{0 . 4 9 7}$ 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 $S E$ is just about $1 \%$ if we sample 3000 voters.
- The normal distribution works for us as before: $95 \%$ of the time our chance error is less than 2 SE , or 2 percentage points.
- For draws from a box: sum the ticket values expected sum $=$ (number of draws) $\times$ (box average) SE of sum $=\sqrt{\text { number of draws }} \times($ box SD)
- Now take average of the ticket values drawn from box: expected average $=($ box average $)$
SE of average $=\mathrm{SE}$ for sum / (number of draws)

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

- 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 chance error

SE of average $=(S D$ 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

