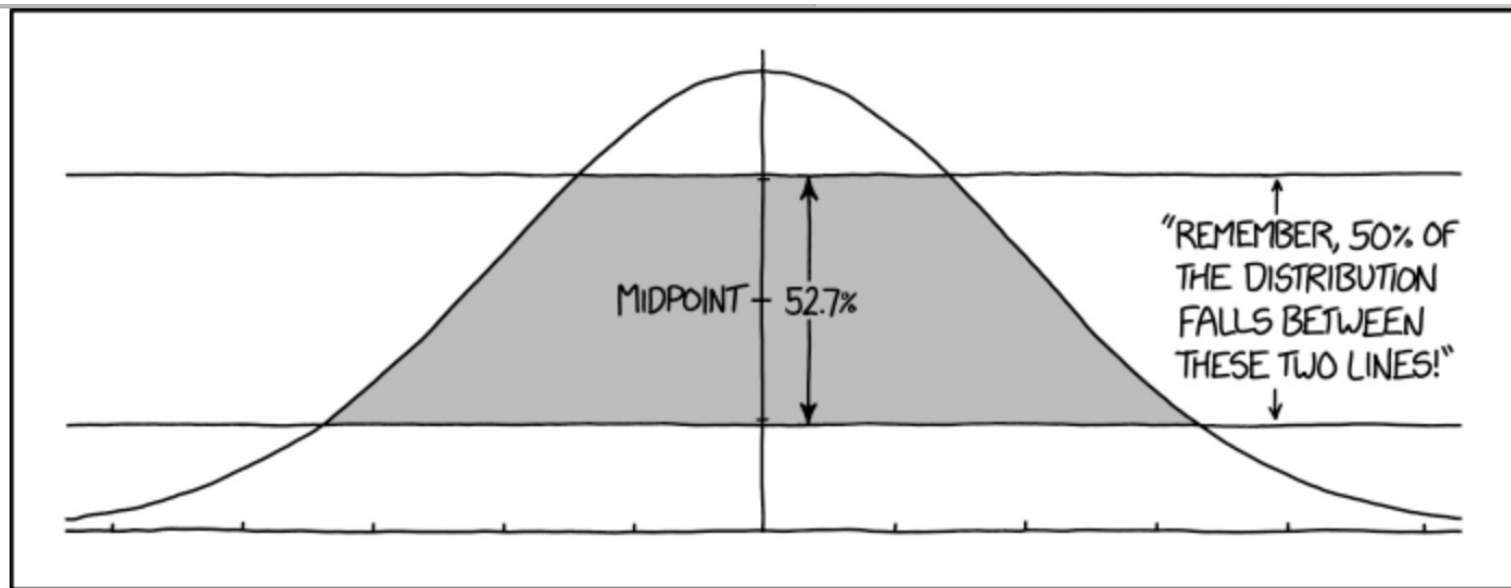


Stat 88: Prob. & Math. Statistics in Data Science



HOW TO ANNOY A STATISTICIAN

xkcd.com/2118

Lecture 28: 4/1/2024

The law of averages, distribution of a sample sum

7.3, 8.1, 8.2

$$S_n = S_n, A_n = \bar{X}$$

Story so far... X_1, X_2, \dots, X_n are iid r.v.

$$S = X_1 + X_2 + \dots + X_n$$

$$E(X_k) = \mu, \text{Var}(X_k) = \sigma^2$$

$$E(S) = n\mu$$

$$S = S_n, A_n = \bar{X}$$

$$A_n = \frac{S}{n}$$

- Variance and SD of sums of iid random variables:

$$\text{Var}(S_n) = n\sigma^2 \quad \boxed{\text{SD}(S_n) = \sqrt{n}\sigma}$$

- Variance of a Binomial rv

$$\boxed{\text{Var}(X) = npq = np(1-p)}$$

$$X_k \sim \text{Bernoulli}(p)$$

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

$$E(X) = np$$

- SD of a sample sum increases with n , whereas the SD of a sample mean decreases with n .

$$\text{SD}(S_n) = \sqrt{n} \cdot \sigma$$

$$\text{SD}(A_n) = \text{SD}\left(\frac{S_n}{n}\right) = \frac{1}{n} \text{SD}(S_n) = \frac{\sqrt{n}\sigma}{\sqrt{n}}$$

$$\boxed{\text{SD}(A_n) = \frac{\sigma}{\sqrt{n}}}$$

$$\boxed{\begin{aligned} \text{Var}(X) &= E(X^2) - E(X)^2 \\ X &\sim \text{Ber}(p) \\ X^2 &\sim \text{Ber}(p) \\ \text{Var}(X) &= p - p^2 \end{aligned}}$$

- When we have a simple random sample (SRS), the draws are without replacement (like drawing cards from a deck).

- Variance of hypergeometric rv:

$$X \sim \text{HG}(N, G, n)$$

$$n \cdot \left(\frac{G}{N}\right) \left(\frac{N-G}{N}\right) \left[\frac{N-n}{N-1}\right] = \text{Var}(X)$$

- Finite population correction:

$$\boxed{\text{f.p.c.} = \sqrt{\frac{N-n}{N-1}}}$$

SD of sum of iid r.v. * f.p.c

= SD of sum of SRS

Suppose X is sum of draws of tickets from a box.

$$\text{SD of sum WITHOUT REPL} = \text{SD of SUM w/Repl} \times \text{fpc}$$

Accuracy of samples (depend on the SD of the sample mean/sum)

- Simple random samples of the same size of 625 people are taken in Berkeley (population: 121,485) and Los Angeles (population: 4 million). True or false, and explain your choice: The results from the Los Angeles poll will be substantially more accurate than those for Berkeley.

Fpc in case of Berkeley: 0.9974285

Fpc in case of LA: 0.999922

Example adapted from Statistics, by FPP

- A survey organization wants to take an SRS in order to estimate the percentage of people who watched the 2022 Oscars. To keep costs down, they want to take as small a sample as possible, but their client will only tolerate a random error of 1 percentage point or so in the estimate. Should they use a sample size of 100, 2500, or 10000? The population is very large and the fpc is about 1.

What n to use? Note that the number of people who have watched the Oscars in the sample is a rv with the $HG(N, G, n)$ distribution.

Note that N is very large so can pretend that we are sampling w/ replacement.

Pretend that we are sampling w/replacement

X = # of people in sample that watched the Oscars

$$X \sim \text{Bin}(n, p) \quad (\text{fpc} \approx 1)$$

↑
approximately

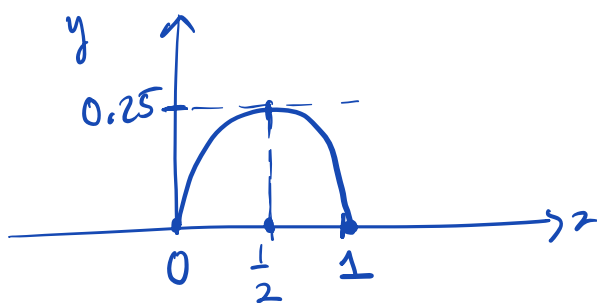
Percent of people in sample that watched Oscars = $\frac{X}{n}$

$$E(X) = np \quad \text{Var}(X) = np(1-p)$$

$$SD(X) = \sqrt{np(1-p)}$$

$$SD\left(\frac{X}{n}\right) = \frac{\sqrt{p(1-p)}}{\sqrt{n}} \leq 0.01$$

$$f(x) = x(1-x), \quad 0 \leq x \leq 1$$



$$x(1-x) \leq 0.25$$

$$\sqrt{x(1-x)} \leq 0.5$$

$$SD\left(\frac{X}{n}\right) \leq \frac{0.5}{\sqrt{n}} \leq 0.01$$

Therefore, what can we say about n ?

$$\frac{0.5}{0.01} \leq \sqrt{n}$$

$$50 \leq \sqrt{n}$$

$$2500 \leq n$$

Example (adapted from *Statistics*, by Freedman, Pisani, and Purves)

- Note that the number of people who have watched the Oscars in the sample is a rv with the $HG(N, G, n)$ distribution, but we are told that N is very large & $fpc \approx 1$, so we can approximate the prob. using the $Bin(n, p)$ distribution, where p is the percentage of people who watched the Oscars (which is what we are trying to estimate).
- $SD\left(\frac{S_n}{n}\right) = \frac{\sigma}{\sqrt{n}} = \frac{\sqrt{pq}}{\sqrt{n}} \leq \frac{0.5}{\sqrt{n}} \leq 0.01 \Rightarrow n \geq 2500$

Exercise 7.4.11


Exercise

Each Data 8 student is asked to draw a random sample and estimate a parameter using a method that has chance 95% of resulting in a good estimate.

Suppose there are 1300 students in Data 8. Let X be the number of students who get a good estimate. Assume that all the students' samples are independent of each other.

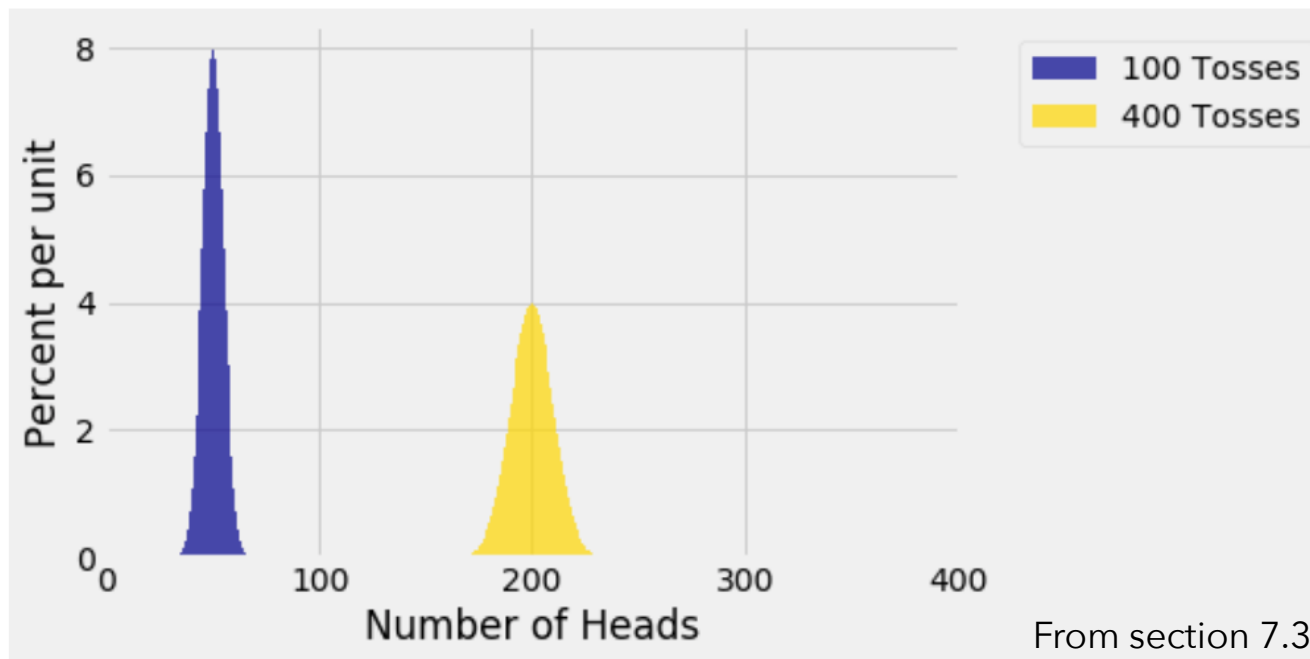
- a) Find the distribution of X
- b) Find $E(X)$ and $SD(X)$.
- c) Find the chance that more than 1250 students get a good estimate.

Law of Averages

- Essentially a statement that you are already familiar with: If you toss a fair coin many times, roughly half the tosses will land heads.
- We are going to consider sample sums and sample means of iid random variables X_1, X_2, \dots, X_n where the mean of each X_k is μ and the variance of each X_k is σ^2 .
- Recall the **sample sum** $S_n = X_1 + X_2 + \dots + X_n$, with $E(S_n) = n\mu$, $Var(S_n) = n\sigma^2$, $SD(S_n) = \sqrt{n}\sigma$

- We see here, as we take more and more draws, the variability of the sum keeps increasing, which means the values get more and more dispersed around the mean ($n\mu$).

Coin tosses

- Consider a fair coin, toss it 100 times & 400 times, count the number of H. Expect in first case, roughly 50 H, and in second, roughly 200 H.
- So do you think chance of 50 H in 100 tosses and 200 H in 400 tosses should be the same?



Example: Coin toss

- $SD(S_{100}) =$
- $SD(S_{400}) =$

- $P(200 \text{ H in } 400 \text{ tosses})$

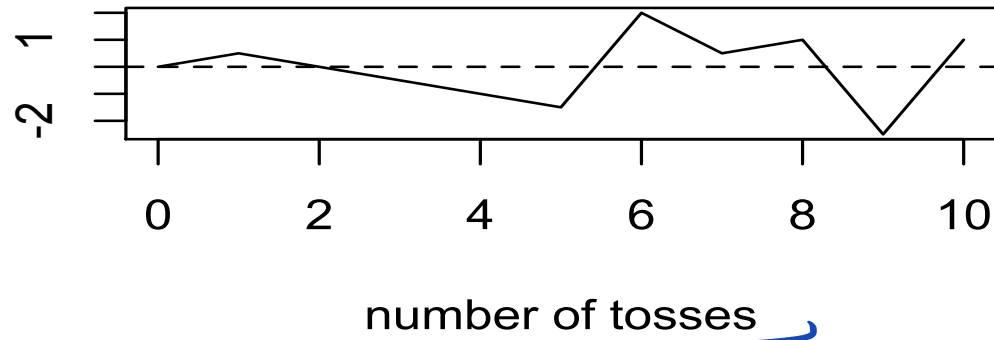
- $P(50 \text{ H in } 100 \text{ tosses})$

Simulating coin tosses: 10 tosses (adapted from FPP)

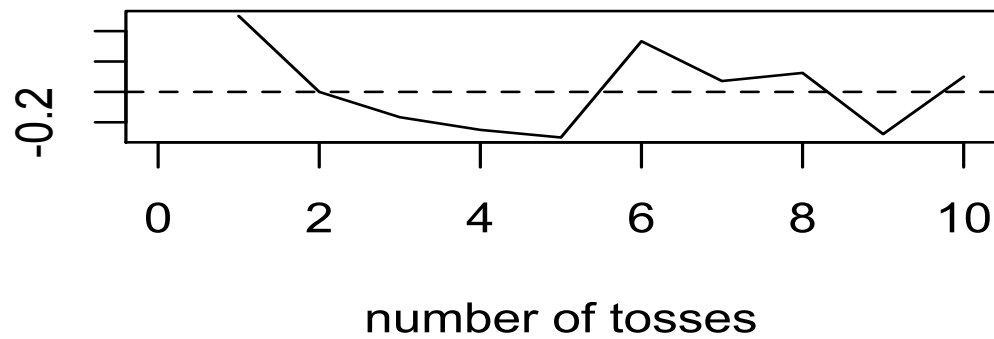
$$\frac{\text{obs. \# of H}}{\text{\# of H}}$$

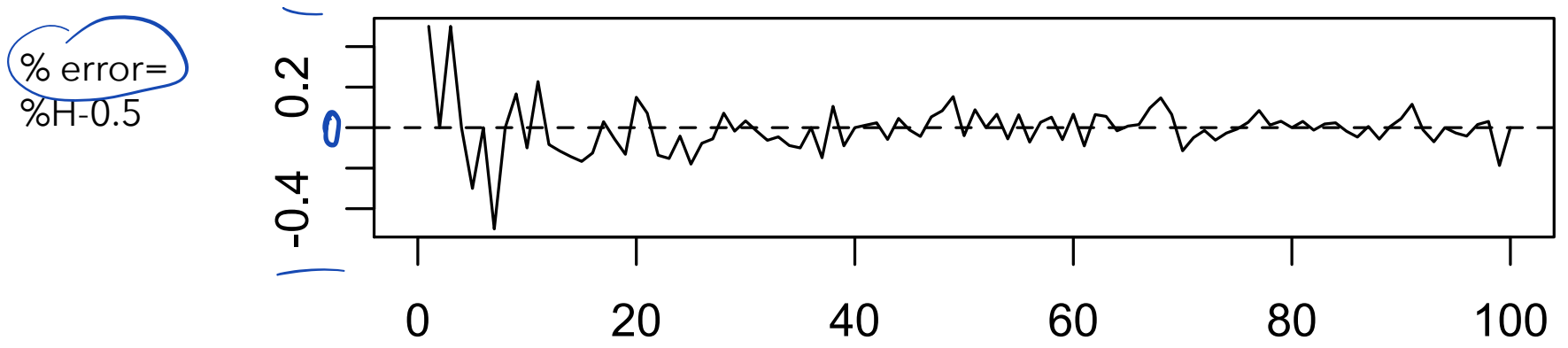
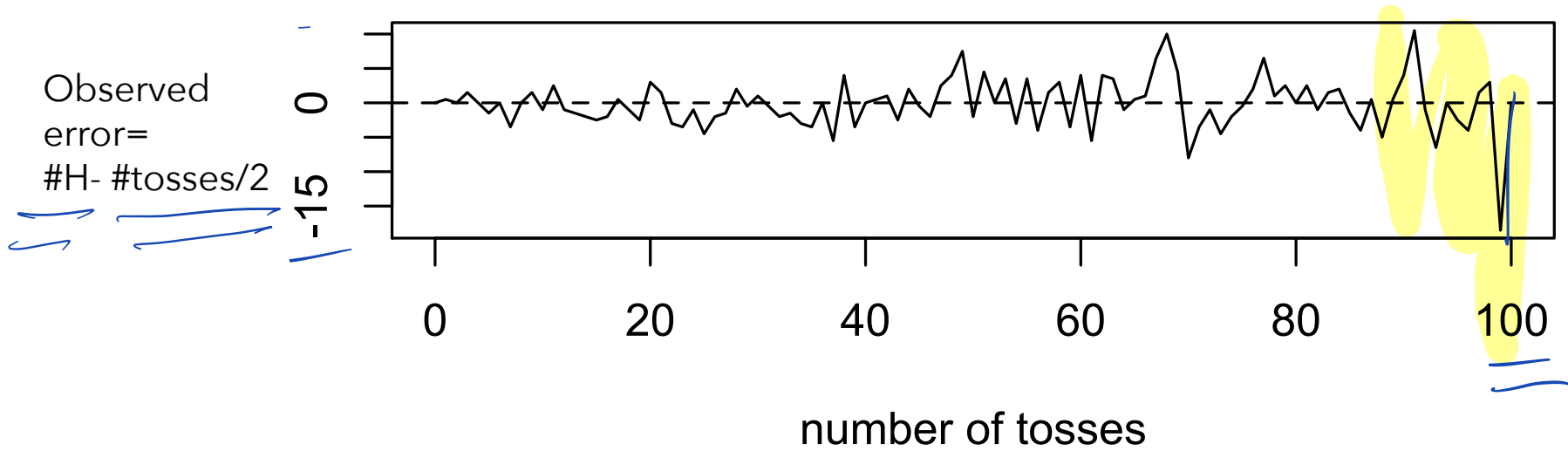
$$\frac{\text{expected \# of H}}{\text{\# of H}}$$

$$\frac{\text{\#(heads) - \#(tosses) * p}}{\text{\# of H}}$$

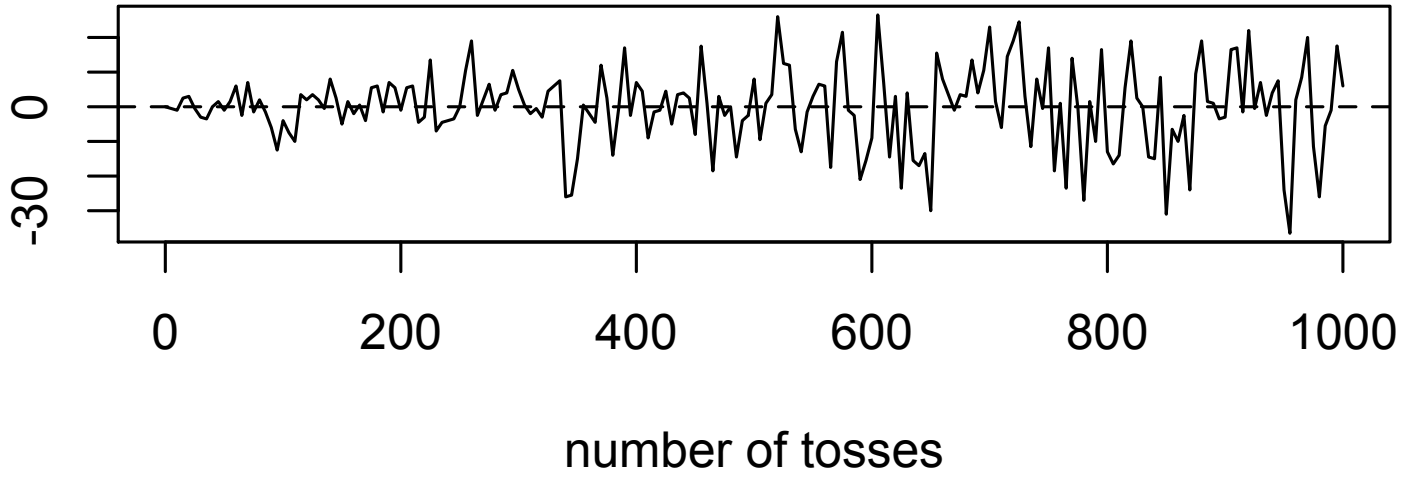


$$\frac{\text{\% heads} - p}{\sqrt{p(1-p)}}$$

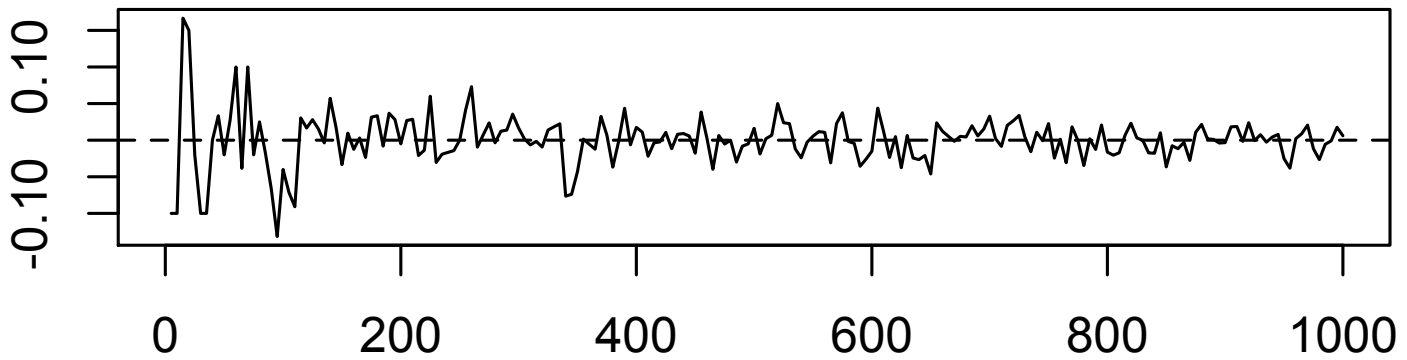


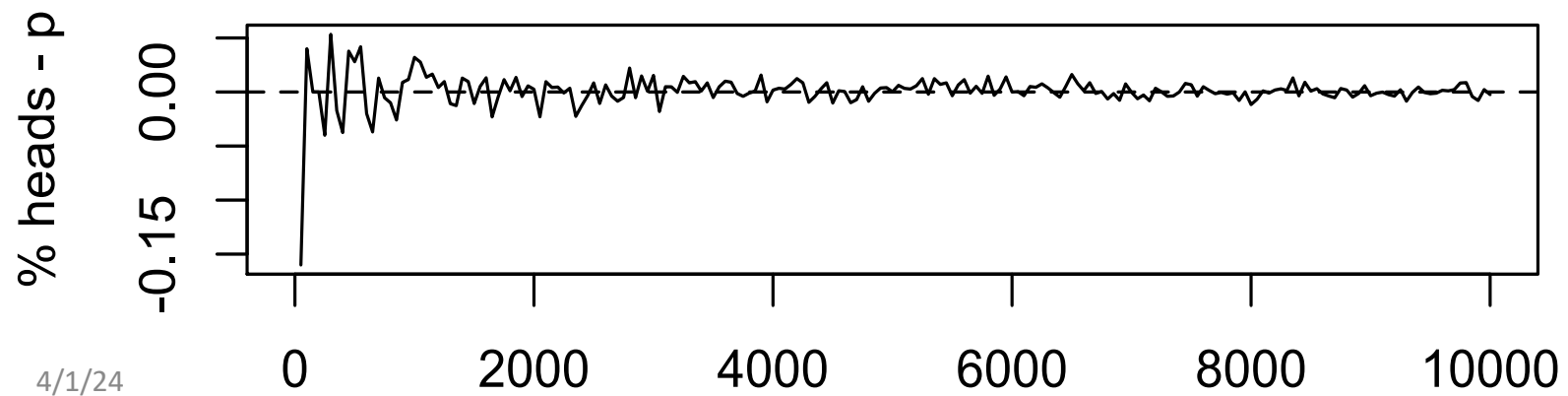
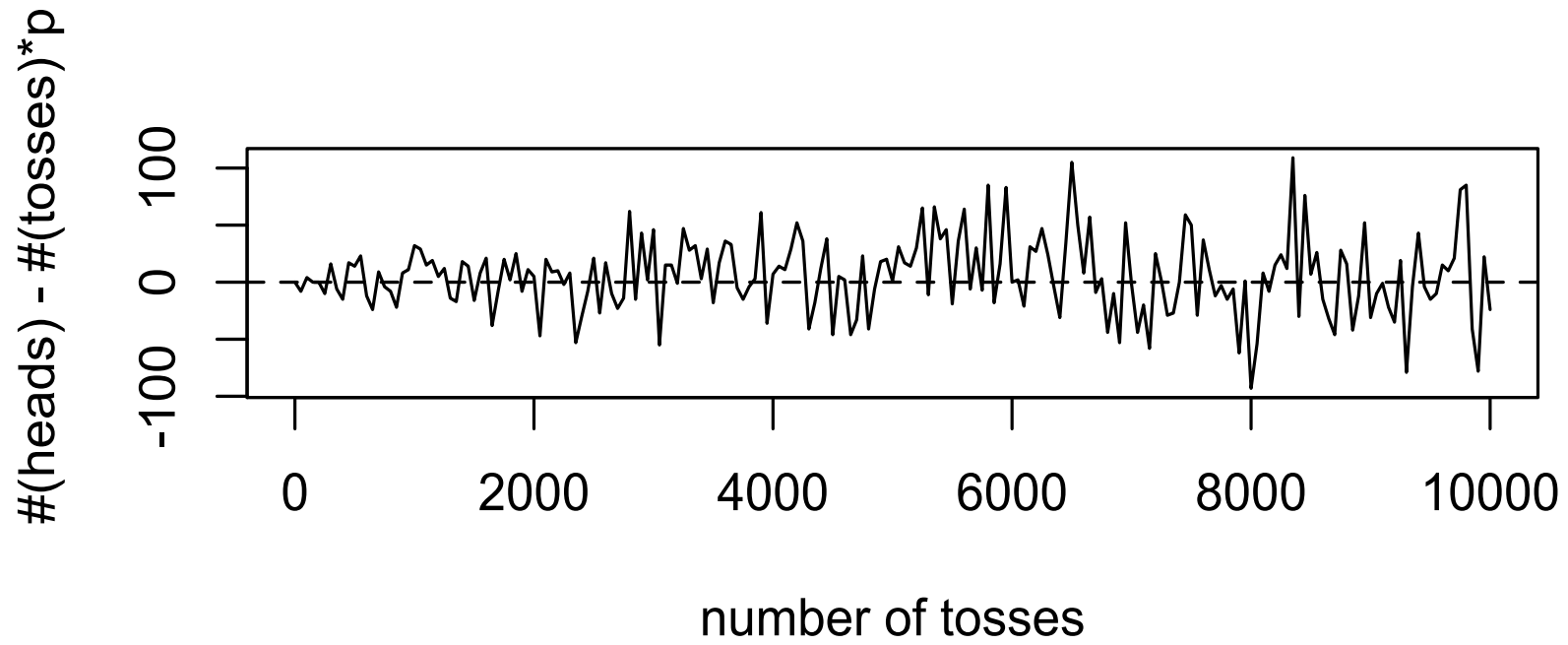


$\#(\text{heads}) - \#(\text{tosses}) * p$



% heads - p





Law of Averages for a fair coin

- Notice that as the number of tosses of a fair coin increases, the *observed error* (number of heads - half the number of tosses) increases. This is governed by the standard error.
- The *percentage* of heads observed comes very close to 50%
- *Law of averages*: The long run *proportion* of heads is very close to 50%.