

1) If today's stock costs Y_0 dollars,
the price in 10 days will be $Y_0 + X_1 + X_2 + \dots + X_{10}$
Let $S_{10} = X_1 + \dots + X_{10}$. change for next
10 days

As the X_i are independent normal r.v.'s
w/ mean 0, var 1, we see S_{10} is normal w/
 $E(S_{10}) = 0, \quad V(S_{10}) = 10 \cdot 1 = 10$

Then for $Y_0 = 100$,

$$\begin{aligned} P(Y_{10} \geq 105) &= P(Y_0 + S_{10} \geq 105) \\ &= P(100 + S_{10} \geq 105) \\ &= P(S_{10} \geq 5) \end{aligned}$$

standardize $\rightarrow = P\left(\frac{S_{10}}{\sqrt{10}} \geq \frac{5}{\sqrt{10}}\right)$

$\left(\frac{S_{10}}{\sqrt{10}} \text{ is std normal}\right) \approx \frac{1}{2} - \text{NA}(1.58) \approx \frac{1}{2} - .4429$
 $= .0571$

2) We wish to construct a 99% confidence interval.
Consulting the normal table, we observe $NA(2.58) = .4951$,
so $2NA(2.58) = .9902$.

We then seek n such that $P(|\bar{p}_n - p| \leq .005) \geq .99$

where: p = proportion of females

$\bar{p}_n = S_n/n$ is our estimate of p .

As $V(\bar{p}_n) = pq/n$ and our 99% confidence interval
ranges from -2.58σ to 2.58σ ,

we see that we need $p \in (\bar{p} - 2.58\sqrt{pq/n}, \bar{p} + 2.58\sqrt{pq/n})$
so we must have

$$2.58\sqrt{\frac{pq}{n}} \leq .005$$

Note $\sqrt{pq} \leq 1/2$, so this becomes

$$\frac{2.58}{\sqrt{n}} \leq .01,$$

$$\text{i.e. } 258 \leq \sqrt{n} \Rightarrow n \geq 258^2 = 66,564$$

3) a) Let S_{1000} = # heads. There are two ways we can fail:

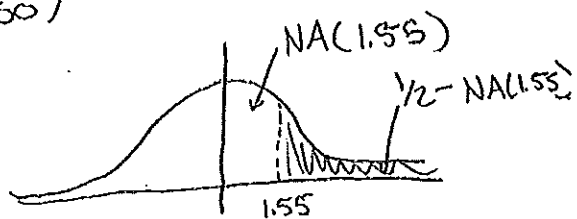
- ① The coin is fair & $S_{1000} \geq 525$
- ② The coin is biased & $S_{1000} < 525$

Case 1: Coin is fair: $E(S_{1000}) = .5(1000) = 500$
 $V(S_{1000}) = (.5)(.5)(1000) = 250$

$P(S_{1000} \geq 525) = P(S_{1000} \geq 524.5)$ (move to continuous)

$= P\left(\frac{S_{1000} - 500}{\sqrt{250}} \geq \frac{24.5}{\sqrt{250}}\right)$ (standardize)

(normal approx. using CLT) $\rightarrow \approx \frac{1}{2} - NA(1.55)$



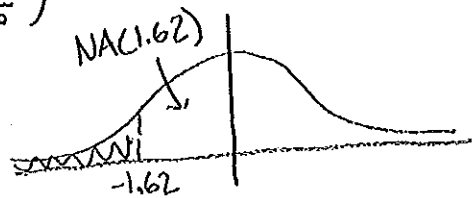
$\approx \frac{1}{2} - .4394 = .0616$

Case 2: Coin is biased: $E(S_{1000}) = (.55)(1000) = 550$
 $V(S_{1000}) = (.55)(.45)(1000) = 247.5$
 $= 247.5$

$P(S_{1000} < 525) = P(S_{1000} < 524.5)$ (move to cts)

$= P\left(\frac{S_{1000} - 550}{\sqrt{247.5}} < \frac{-25.5}{\sqrt{247.5}}\right)$ (standardize)

normal approx. $\rightarrow \approx \frac{1}{2} - NA(1.62)$



$\approx .0526$

So $P(\text{guess wrong}) = \frac{1}{2}(.0616) + \frac{1}{2}(.0526) = .0571$
 fair coin \rightarrow \leftarrow biased coin

b) Using ~~guess~~ guess & check (in R), see 525 is optimal,

4) Let X_i represent the rolls of dice.

We compute

$$\begin{aligned} P(X_1 \cdot X_2 \cdot \dots \cdot X_{100} \leq a^{100}) &= P(\log(X_1 \cdot X_2 \cdot \dots \cdot X_{100}) \leq \log(a^{100})) \\ &= P(\log(X_1) + \dots + \log(X_{100}) \leq 100 \log(a)) \end{aligned}$$

We can now use the central limit theorem to approximate this probability w/

$$E(\log X_i) = \sum_{i=1}^6 \log(i) / 6 = 1.0965 = \mu \quad \left(\begin{array}{l} \text{here } \log \text{ is} \\ \text{natural } \log, \text{ or } \ln \end{array} \right)$$

$$V(\log X_i) = E((\log X_i - \mu)^2) = .4391 = \sigma^2 \quad \uparrow$$

so for $S_n = \log X_1 + \dots + \log X_n$,

$$E(S_n) = n\mu, \quad V(S_n) = n\sigma^2$$

$$\left(\begin{array}{l} \log_{10} \text{ gives } \\ E(\log_{10} X_i) = .4762, \\ V(\log_{10} X_i) = .0828 \end{array} \right)$$

Then

$$P(S_{100} \leq 100 \log(a)) = P\left(\frac{S_{100} - 100\mu}{\sqrt{100\sigma^2}} \leq \frac{100 \log(a) - 100\mu}{\sqrt{100\sigma^2}} \right)$$

standardize \nearrow

$$\approx P(Z \leq \frac{100 \log(a) - 109.6542}{6.6264})$$

$$= F_Z\left(\frac{100 \log(a) - 109.6542}{6.6264} \right)$$

w/ Z the standard normal r.v. & F_Z its cdf

$$\left(\text{for } \log_{10}, \text{ have } F_Z\left(\frac{100 \log(a) - 47.6222}{2.8778} \right) \right)$$

5) Assume each customer is using the phone w/ prob.
 $\approx 1/30$ during peak hours.

(as some have pointed out, this is a kind of fishy assumption, but here a necessary one)

Then we have a Bernoulli trial process

w/ $p = 1/30$, $n = 2000$. We seek k s.t.

for $S_{2000} = \#$ successes, (phone users)

$$P(S_{2000} \leq k) \leq .01$$

We have $P(S_{2000} \leq k) = P\left(\frac{S_{2000} - \overset{E(S_{2000})}{2000/30}}{\underset{SD(S_{2000})}{\sqrt{2000(\frac{1}{30})(\frac{29}{30})}}} \leq \frac{k - 2000/30}{\sqrt{2000(\frac{1}{30})(\frac{29}{30})}}\right)$

standardize \nearrow

$$= P(S_{2000}^* \leq \frac{k - 66\frac{2}{3}}{8.0277})$$

Consulting the normal table, we see we need 2.34 standard deviations to reach the desired likelihood,

$$\text{so } \frac{k - 66\frac{2}{3}}{8.0277} \geq 2.34 \Leftrightarrow k \geq 2.34(8.0277) + 66\frac{2}{3} \geq 85.4516$$

so $k = 86$ (it must be an integer)

Using ~~poiss cdf~~ $(2000, 86)$ in ~~MATLAB~~, we see the same value suffices. ~~ppois~~ $(86, \frac{2000}{30})$ we see same value suffices

b) We will prove for X_1, X_2, \dots ind. identically dist. r.v.'s, and $S_n = X_1 + \dots + X_n$ that for any $\varepsilon > 0$, w/ finite mean, variance

$$P\left(\left|\frac{S_n - \mu}{n}\right| \leq \varepsilon\right) \rightarrow 1 \text{ as } n \rightarrow \infty$$

(this is one statement of the law of large numbers)

Pf. Let $\varepsilon > 0$. Then for $\sigma = \text{SD}(X_i)$, we have

$$\begin{aligned} P\left(\left|\frac{S_n - \mu}{n}\right| \leq \varepsilon\right) &= P(-\varepsilon \leq \frac{S_n - \mu}{n} \leq \varepsilon) \\ &= P\left(-\frac{\varepsilon\sqrt{n}}{\sigma} \leq \frac{S_n - \mu}{\sqrt{n}\sigma} \leq \frac{\varepsilon\sqrt{n}}{\sigma}\right) \end{aligned}$$

$$= P\left(-\frac{\varepsilon\sqrt{n}}{\sigma} \leq S_n^* \leq \frac{\varepsilon\sqrt{n}}{\sigma}\right)$$

as $n \rightarrow \infty$, this goes to $\int_{-\frac{\varepsilon\sqrt{n}}{\sigma}}^{\frac{\varepsilon\sqrt{n}}{\sigma}} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$

by the central Limit Theorem.

Since $\frac{\varepsilon\sqrt{n}}{\sigma} \rightarrow \infty$ & $-\frac{\varepsilon\sqrt{n}}{\sigma} \rightarrow -\infty$ as $n \rightarrow \infty$,

this integral approaches $\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = 1$.

□