Announcements

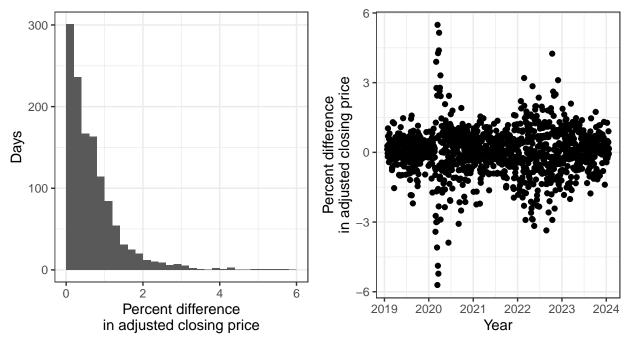
Make sure to sign in on the google form.

Pset 3 due...



Stonks

The following questions deal with the past 5 years of S&P 500 adjusted closing prices available here.



In this section, we will be modeling the day-to-day percent differences in the adjusted closing price of the S&P 500 as $Y_1, ..., Y_n \sim \text{Expo}(\lambda)$.

1. Find the score of λ (recall that the score is $\frac{\partial}{\partial \lambda} \ell(\lambda; \vec{y})$) in terms of the sample mean and verify that $E(s(\lambda^*; \vec{Y})) = 0$.

2. Verify the information equality by showing $-E(s'(\lambda; \overrightarrow{Y})) = \text{Var}(s(\lambda; \overrightarrow{Y}))$.

3. Find the Fisher information $\mathcal{I}_{\overrightarrow{Y}}(\lambda^*)$. Then, find a function g such that $\mathcal{I}_{\overrightarrow{Y}}(g(\lambda^*))$ is constant (this is the variance stabilizing transformation of the Exponential distribution). Hint: Recall that the Fisher information for a transformation is $\mathcal{I}_{\overrightarrow{Y}}(g(\lambda^*)) = \frac{\mathcal{I}_{\overrightarrow{Y}}(\lambda^*)}{g'(\lambda^*)^2}$.

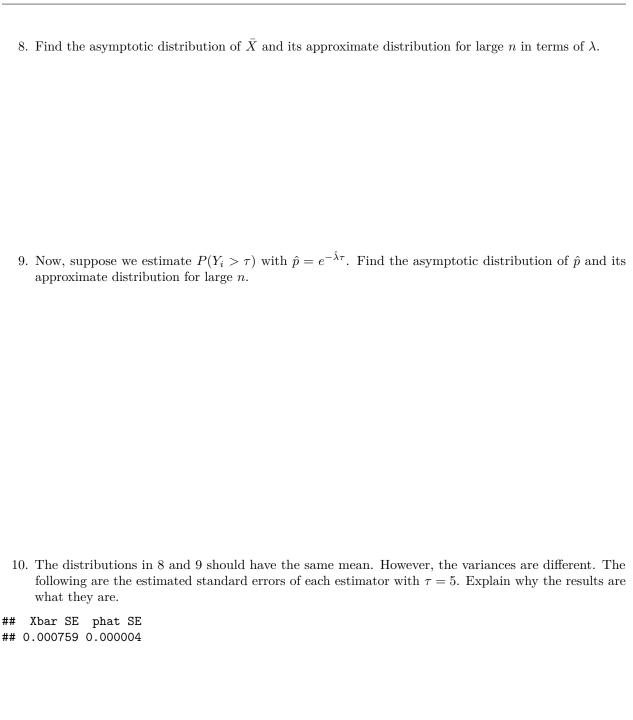
4. Verify this is indeed the variance stabilizing transformation through simulation.

```
set.seed(111)
sapply(c(0.001, 0.01, 0.1, 1, 10, 100, 1000), function(lambda) var(log(rexp(100000, lambda))))
## [1] 1.630103 1.651781 1.654339 1.629304 1.655330 1.641993 1.645989
```

5. Show that the MLE of $\hat{\lambda}$ is consistent for λ . That is, show that $\hat{\lambda} \to \lambda$ as $n \to \infty$ by showing the MSE goes to 0, a LLN holds, making a claim using the CMT, or showing convergence directly.

6. Find the asymptotic distribution of the MLE and its approximate distribution for large n.

7. In his book The Black Swan, Nassim Taleb argues that part of the reason for the 2008 financial crisis was a failure to model market fluctuations and assign sufficient probability to extreme events. Let us consider daily absolute differences above τ to be extreme events. Let $X_i = I(Y_i > \tau)$. Show that \bar{X} is consistent for $p = P(Y_i > \tau)$ first by using MSE and then by using the law of large numbers.



11. Though \hat{p} is more efficient, it might be less robust. The following shows the MSEs of the two estimators for estimating $P(Y > \tau)$ when $Y \sim \text{Expo}(0.5)$ (the correct model) and $Y \sim \text{Log-Normal}(0.1, 1)$ (an incorrect model). Again, we have used $\tau = 5$ and n = 100 with 10^5 simulations.

```
## Log Normal 0.00061 0.00053
## Expo 0.00075 0.00041
```

Ty Mup

Ty Mup is taking an exam with n equally hard questions. He has a probability p_2 of getting each question right independently. However, there is also a $0 < p_1 < 1$ probability he sleeps through his alarm and misses the exam entirely. Let Y be the number of questions he gets right on his exam. (This distribution is called the zero-inflated binomial.)

1. Find E(Y|Y > 0).

2. Unfortunately, the day is February 2nd in Punxsutawney and Ty is destined to repeat this day d times, scoring i.i.d Y_i on the exams. Find the likelihood function, the log-likelihood function, the score for p_1 and p_2 . (Hint: Let m be the number of 0s.)

3. Find a two-dimensional sufficient statistic (a two dimensional statistic that contains all the information about the likelihood).

4. Find the Fisher information for p_1 . (Hint: Write M as $M_1 + M_2$ where M_1 is the number of days Ty slept through the alarm and M_2 is the number of times he took the test and got a 0.)

5. Check that this Fisher information gives the correct result in the cases $p_2 = 1$ and $p_2 = 0$.

6. Let B be the event that Ty sleeps through the alarm at least once. Show that as $d \to \infty$,

$$I_B \frac{d^{1/2}(\bar{Y} - (1 - p_1)np_2)}{\sqrt{np_2(1 - p_2)(1 - p_1) + (np_2)^2p_1(1 - p_1)}} \xrightarrow{d} \mathcal{N}(0, 1)$$