

DATA 5690: Midterm

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Introduction

This midterm exam covers basic statistical inference from both the frequentist (objectivist) and Bayesian (subjectivist) perspectives. Emphasis is on fundamental understanding and comparative interpretation.

Note

If you get stuck make appropriate assumptions, document them, and proceed.

Frequentist Analysis

1. In this question your task is to carry out statistical inference for a binomial proportion from the frequentist perspective.

The frequentist agent confronts a tootsie roll candy machine with a fixed but unknown probability of dispensing a cherry tootsie roll denoted by θ .

a. Generate artificial data for this scenario with the following code:

```
import numpy as np

np.random.seed(42)

theta = 0.30
tootsie_rolls = np.random.binomial(n=1, p=theta, size=50)
```

The agent is given these data and told they represent draws from the candy machine where an observation of 1 represents a cherry tootsie roll and an observation of 0 represents a vanilla tootsie roll.

- b. Compute the maximum likelihood estimator $\hat{\theta}_{MLE}$ as though you were the agent. What is the agent's numerical point estimate of this maximum likelihood estimator?
- c. What is the sampling distribution of $\hat{\theta}_{MLE}$ according to the *Central Limit Theorem*? Make a plot of it using `matplotlib.pyplot`.
- d. Compute a 95% confidence interval for $\hat{\theta}_{MLE}$. What are the upper and lower bounds? Give a formal interpretation of this confidence interval.
- e. Conduct a hypothesis test that the tootsie roll machine is biased towards dispensing vanilla tootsie rolls with a level of significance of 5%.
 - State the null hypothesis.
 - State the alternative hypothesis.
 - Compute the test statistic and report its numerical value.
 - Compute the rejection region and report its numerical value.
 - Is this a one-tailed or two-tailed test?
 - What does the agent conclude? State it formally.

f. Please redo parts a-e for $\theta = 0.45$.

2. In this question your task is to carry out statistical inference for count data from the frequentist perspective. Assume that these data represent visitors that arrive per hour to take a turn at the tootsie roll machine. Let λ be the hourly arrival rate of the visitors.

a. Generate artificial data for this problem with the following code:

```
import numpy as np

np.random.seed(42)

lam = 20
visits = np.random.poisson(lam=lam, size=50)
```

b. The agent doesn't see the data-generating process but assumes that they come from a Poisson distribution with a fixed but unknown λ parameter. Compute the maximum likelihood estimator $\hat{\lambda}_{MLE}$.

c. What is the sampling distribution of $\hat{\lambda}_{MLE}$ according to the *Central Limit Theorem*? Make a plot of it using `matplotlib.pyplot`.

d. Compute a 95% confidence interval for $\hat{\lambda}_{MLE}$. What are the upper and lower bounds? Give a formal interpretation of this confidence interval.

e. Conduct a hypothesis test that the true arrival rate is 18 visitors per hour.

- State the null hypothesis.
- State the alternative hypothesis.
- Compute the test statistic and report its numerical value.
- Compute the rejection region and report its numerical value.
- Is this a one-tailed or two-tailed test?
- What does the agent conclude? State it formally.

3. Use the IID bootstrap procedure to generate an approximate sampling distribution for $\hat{\lambda}_{MLE}$ in the previous problem using the same data that were given to the agent.

a. You can produce a single bootstrap sample with the following code:

```
np.random.seed(42)

x_b = np.random.choice(a=x, size=50, replace=True)
```

Given this bootstrap sample you would then compute a bootstrap replication of the MLE: $\hat{\lambda}_{MLE}^b$.

- b. Repeat the above for $b = 1, \dots, B$ with $B = 10,000$.
- c. Reproduce the confidence interval and hypothesis test from question 2 above but using the bootstrap sampling distribution rather than appealing to the CLT.
- d. Compare this computational procedure to the classical approach using the CLT.

Bayesian Analysis

4. Reproduce the statistical inference for the data from problem 1 above but from the subjective Bayesian perspective.

- Assume the agent has a prior of $\theta \sim \text{Beta}(a = 1, b = 1)$.
- Compute the posterior distribution.
- Make plots of the prior, likelihood and posterior using `matplotlib.pyplot`.
- Calculate the posterior probability that $\theta = 0.5$.
- Compute a 95% equal-tailed credibility interval.
- Using Bayes' factors conduct a hypothesis test for $H_1 : \theta = 0.5$ (i.e. a fair coin) against $H_2 : \theta \neq 0.5$ (i.e. a biased coin). See Clyde et al (2022) Chapter 3 for details on implementing Bayes' factors.
- Interpret the results. Compare the results to the frequentist procedure.

5. Reproduce the statistical inference for the data from problem 2 above but from the subjective Bayesian perspective.

- Assume the agent has the prior: $\lambda \sim \text{Gamma}(\alpha, \beta)$, which is the conjugate prior for the Poisson likelihood function.
- Compute the posterior distribution.
- Make plots of the prior, likelihood, and posterior using `matplotlib.pyplot`.
- Compute a 95% equal-tailed credibility interval.
- Using Bayes' factors conduct a hypothesis test for $H_1 : \lambda = 18$ against $H_2 : \lambda \neq 18$. Use a diffuse prior for H_2 .

Additional Topics

More to come...

References

Clyde et al. 2022. *An Introduction to Bayesian Thinking: A Companion to the Statistics with R Course*. <https://statswithr.github.io/book/>.