

# Introduction to Statistics Exam (0560.1823)- MOED A

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## **General Instructions**

The test has five questions, each with 20 points.

The test is with open materials, e.g. books, formula pages, or whatever you want. You can use a calculator. You cannot use a laptop.

The test's duration is 3 hours.

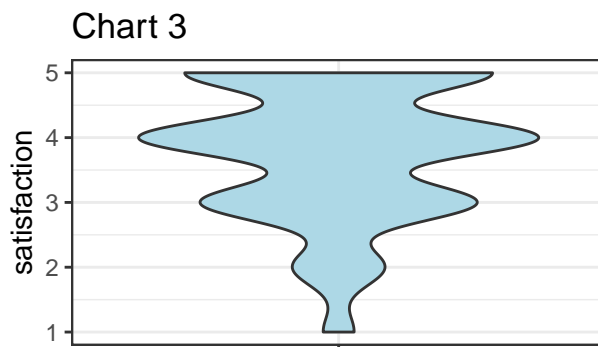
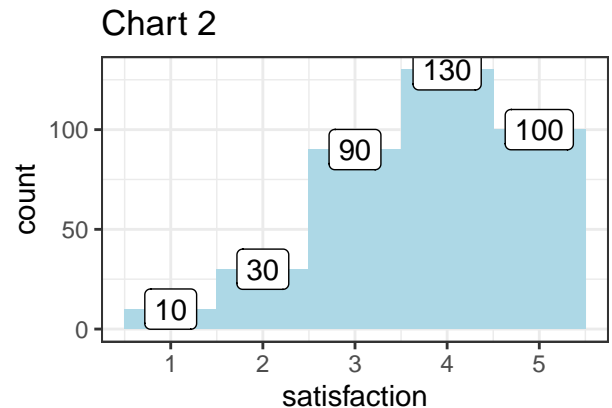
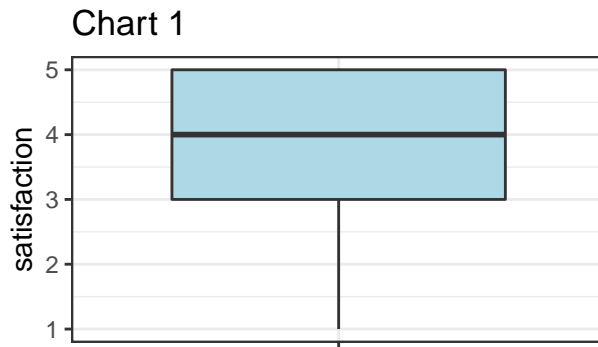
I will be circulating around in case you have any questions during the test.

Good luck!

## Question 1 (20 pts)

A researcher conducted a survey with undergraduate students, about their satisfaction from their studies. The researcher included a question about the age of the respondent and a satisfaction question on a 5-point scale (i.e., 1 = extremely dissatisfied, 2 = dissatisfied, 3 = neither satisfied nor dissatisfied, 4 = satisfied, 5 = extremely satisfied).

The following charts show three different ways in which the distribution of satisfaction can be expressed.



### Questions:

- Which of the charts is better suited to show the distribution of **satisfaction**? explain.
- What is the median satisfaction?
- What chart contains the most information?
- Use the chart (from previous question) to compute: (1) the proportion of satisfied+extremely satisfied students; (2) the average satisfaction level (i.e., a number on the 1-5 scale).
- The researcher wants to visualize the relationship between age and satisfaction. What **geom\_\*** and aesthetic mapping would you use for that? explain and draw an illustration (it doesn't have to be accurate).

## Question 2 (20 pts)

Suppose we have a random sample of size  $2n$  from a population denoted by  $X$ , and  $E[X] = \mu$ ,  $\text{Var}[X] = \sigma^2$ . Let

$$\bar{X}_1 = \frac{1}{2n} \sum_{i=1}^{2n} X_i \quad \text{and} \quad \bar{X}_2 = \frac{1}{n} \sum_{i=1}^n X_i$$

are two estimates for  $\mu$ . Which is a better estimator of  $\mu$ ? Explain your choice (prove it!).

### Question 3 (20 pts)

The ministry for health is planning the purchase of flu vaccines for next year. For that purpose, data of vaccine consumption of the last few years has been concentrated:

- 2014: 1.52 million doses
- 2015: 1.56 million doses
- 2016: 1.63 million doses
- 2017: 1.91 million doses
- 2018: 1.53 million doses
- 2019: 1.83 million doses

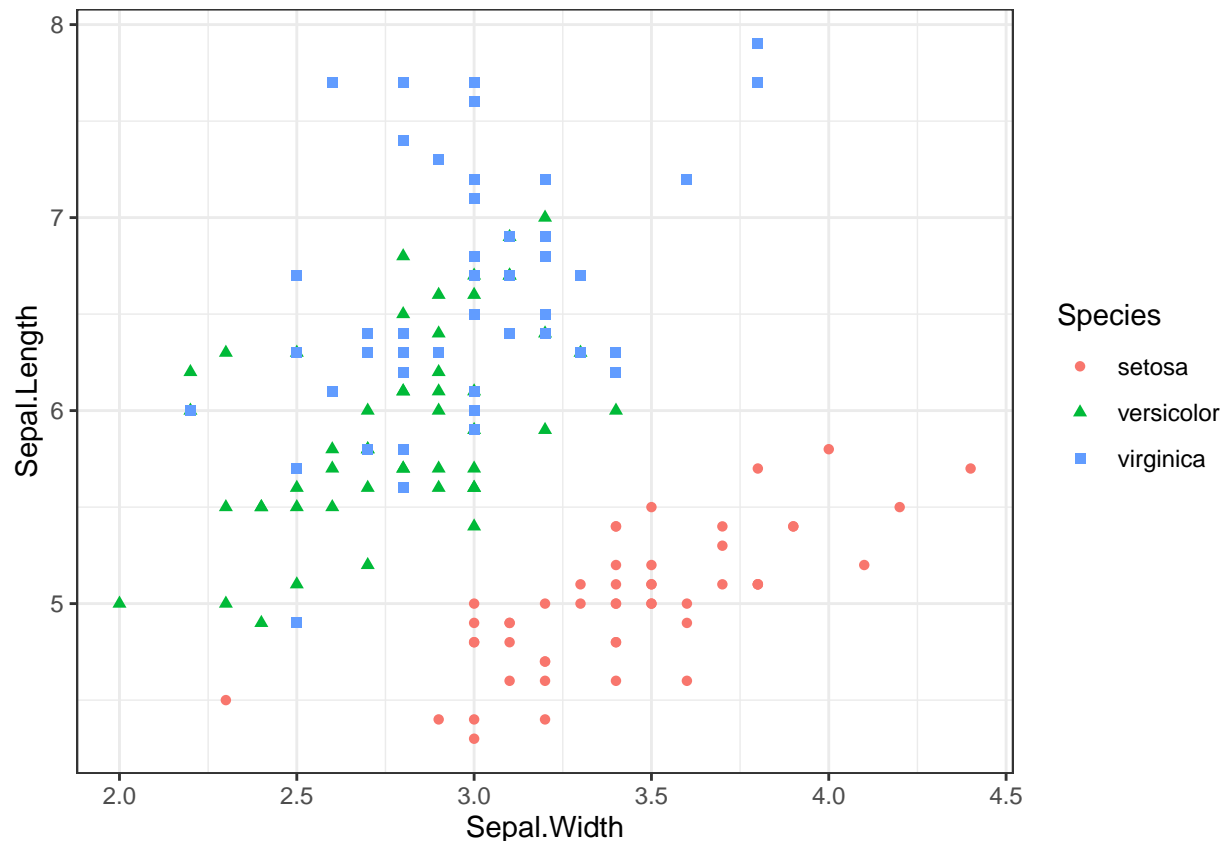
#### Questions:

- (a) Provide a 95% one-sided confidence interval for the number of doses used (an upper bound).
- (b) Provide a 95% one-sided prediction interval for the number of doses used (an upper bound).
- (c) Explain the difference between a confidence interval and a prediction interval.
- (d) The minister wants to cover 95% probability that the vaccines do not run out for next year. What estimate would you use for planning? explain.
- (e) Bonus: What kind of factors might intervene with the number of vaccine doses consumed during a specific year? How would you consider such factors (with what statistical model)?

## Question 4 (20 pts)

The `iris` dataset contains three types of flowers: `setosa`, `virginica`, and `versicolor`. The following chart shows the scatter plot of `Sepal.Width` and `Sepal.Length` variables for the species (the width and length of the parts which hold the base of the flower).

```
ggplot(iris, aes(Sepal.Width, Sepal.Length, color = Species, shape = Species)) +  
  geom_point() +  
  theme_bw()
```



The following linear models (simple linear regression models) show the relationship of the `Sepal.Width` and `Sepal.Length` variables. The first model includes all species and the second model only includes the `setosa` species.

### Questions:

- What kind of measure can you use to compare the two models?
- Explain which model is better and why.
- In the second model (which includes just the `setosa`) what is the coefficient  $\beta$  of `Sepal.Length`? What is its meaning? (i.e., what is the interpretation of the coefficient in the relationship between the `Sepal.Width` and `Sepal.Length`)
- Can you think of a way to extend the first model so that it will still include all the species but have a much better fit? explain!

```
# The first model:  
lm(formula = Sepal.Width ~ Sepal.Length, data = iris) %>% summary()
```

```
##
## Call:
## lm(formula = Sepal.Width ~ Sepal.Length, data = iris)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.1095 -0.2454 -0.0167  0.2763  1.3338
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.41895    0.25356   13.48  <2e-16 ***
## Sepal.Length -0.06188    0.04297   -1.44    0.152
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4343 on 148 degrees of freedom
## Multiple R-squared:  0.01382, Adjusted R-squared:  0.007159
## F-statistic: 2.074 on 1 and 148 DF, p-value: 0.1519

# The second model:
lm(formula = Sepal.Width ~ Sepal.Length, data = iris %>% filter(Species == "setosa")) %>%
  summary()

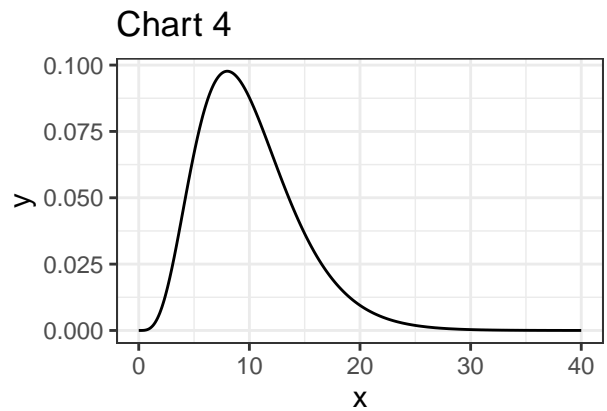
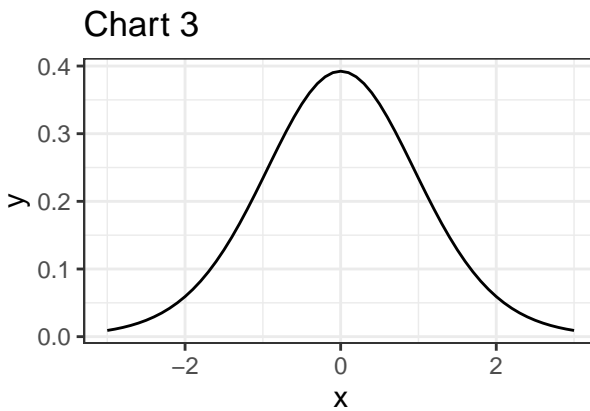
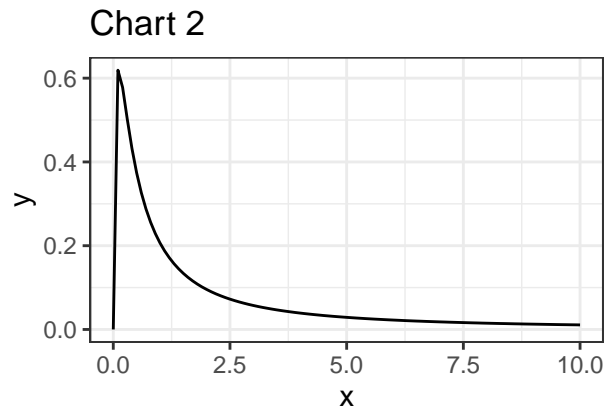
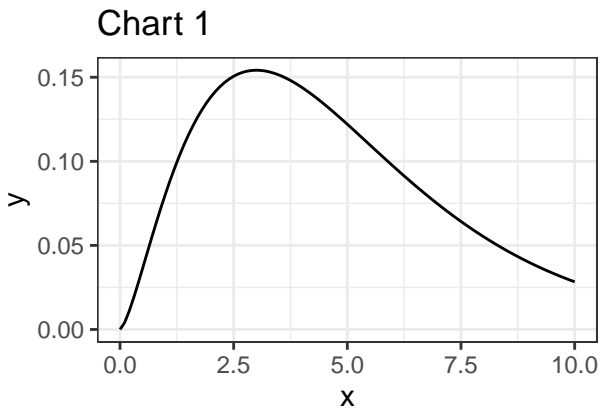
##
## Call:
## lm(formula = Sepal.Width ~ Sepal.Length, data = iris %>% filter(Species ==
##      "setosa"))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.72394 -0.18273 -0.00306  0.15738  0.51709
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.5694    0.5217  -1.091    0.281
## Sepal.Length  0.7985    0.1040   7.681 6.71e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2565 on 48 degrees of freedom
## Multiple R-squared:  0.5514, Adjusted R-squared:  0.542
## F-statistic: 58.99 on 1 and 48 DF, p-value: 6.71e-10
```

## Question 5 (20 pts)

Before you are four charts, for each chart write which distribution it depicts out of the following list:

- Student's t
- Chi square
- F

(Please note that there are additional parts of this question after the charts!)



For each of the following statistical tests, which distribution is used for the test statistic:

- Linear regression: a single  $\beta_i$  test, i.e.,  $H_0 : \beta_i = 0$  and  $H_1 : \beta_i \neq 0$ .
- Linear regression: there exists a non-zero coefficient. I.e.,  $H_0 : \forall i \beta_i = 0$  and  $H_1 : \exists i : \beta_i \neq 0$ .
- ANOVA test.
- Goodness of fit.