# C3\_W1\_Assignment

#### April 23, 2023

## 1 Estimating Treatment Effect Using Machine Learning

Welcome to the first assignment of AI for Medical Treatment!

You will be using different methods to evaluate the results of a randomized control trial (RCT). You will learn: - How to analyze data from a randomized control trial using both: - traditional statistical methods - and the more recent machine learning techniques - Interpreting Multivariate Models - Quantifying treatment effect - Calculating baseline risk - Calculating predicted risk reduction - Evaluating Treatment Effect Models - Comparing predicted and empirical risk reductions - Computing C-statistic-for-benefit - Interpreting ML models for Treatment Effect Estimation - Implement T-learner

#### 1.1 Table of Contents

- Section ??
  - Section ??
  - Section ??
    - \* Section ??
    - \* Section ??
  - Section ??
- Section ??
  - Section ??
    - \* Section ??
  - Section ??
    - \* Section ??
  - Section ??
    - \* Section ??
    - \* Section ??
- Section ??
  - Section ??
    - \* Section ??
    - \* Section ??
- Section ??

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  - \* Section ??
  - \* Section ??
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## 1.2 Packages

We'll first import all the packages that we need for this assignment.

- pandas is what we'll use to manipulate our data
- numpy is a library for mathematical and scientific operations
- matplotlib is a plotting library
- sklearn contains a lot of efficient tools for machine learning and statistical modeling
- random allows us to generate random numbers in python
- lifelines is an open-source library that implements c-statistic
- itertools will help us with hyperparameters searching

#### 1.3 Import Packages

Run the next cell to import all the necessary packages, dependencies and custom util functions.

```
In [16]: import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import sklearn
    import random
    import lifelines
    import itertools
    from public_tests import *

plt.rcParams['figure.figsize'] = [10, 7]
```

#### ## 1. Dataset ### 1.1 Why RCT?

In this assignment, we'll be examining data from an RCT, measuring the effect of a particular drug combination on colon cancer. Specifically, we'll be looking the effect of Levamisole and Fluorouracil on patients who have had surgery to remove their colon cancer. After surgery, the curability of the patient depends on the remaining residual cancer. In this study, it was found that this particular drug combination had a clear beneficial effect, when compared with Chemotherapy. ### 1.2 Data Processing In this first section, we will load in the dataset and calculate basic statistics. Run the next cell to load the dataset. We also do some preprocessing to convert categorical features to one-hot representations.

```
In [17]: data = pd.read_csv("data/levamisole_data.csv", index_col=0)
```

Let's look at our data to familiarize ourselves with the various fields.

Data Dimensions: (607, 14)

Out[18]:		sex	age	obstruct	perfor	adher	e nodes	node4	outcome	TRTMT	\
	1	1	43	0	0		0 5.0	1	1	True	
	2	1	63	0	0		0 1.0	0	0	True	
	3	0	71	0	0		1 7.0	1	1	False	
	4	0	66	1	0		0 6.0	1	1	True	
	5	1	69	0	0		0 22.0	1	1	False	
		diff	er_2.	O differ_	3.0 ex	tent_2	extent_	3 exten	t_4		
	1			1	0	0		1	0		
	2			1	0	0		1	0		
	3			1	0	1		0	0		
	4			1	0	0		1	0		
	5			1	0	0		1	0		

Below is a description of all the fields (one-hot means a different field for each level): - sex (binary): 1 if Male, 0 otherwise - age (int): age of patient at start of the study - obstruct (binary): obstruction of colon by tumor - perfor (binary): perforation of colon - adhere (binary): adherence to nearby organs - nodes (int): number of lymphnodes with detectable cancer - node4 (binary): more than 4 positive lymph nodes - outcome (binary): 1 if died within 5 years - TRTMT (binary): treated with levamisole + fluoroucil - differ (one-hot): differentiation of tumor - extent (one-hot): extent of local spread

In particular pay attention to the TRTMT and outcome columns. Our primary endpoint for our analysis will be the 5-year survival rate, which is captured in the outcome variable.

### Exercise 1 - proportion\_treated

Since this is an RCT, the treatment column is randomized. Let's warm up by finding what the treatment probability is.

$$p_{treatment} = \frac{n_{treatment}}{n}$$

- $n_{treatment}$  is the number of patients where TRTMT = True
- *n* is the total number of patients.

```
### START CODE HERE (REPLACE INSTANCES OF 'None' with your code) ###
             proportion = sum(df.TRTMT==1)/len(df.TRTMT)
             ### END CODE HERE ###
             return proportion
In [20]: ### test cell ex1: you cannot edit this cell
         proportion_treated_test(proportion_treated)
Test Case 1:
Example df:
   outcome TRTMT
0
         0
1
         1
                1
                1
         1
                1
Proportion of patient treated: 0.75
Test Case 2:
Example df:
   outcome
           TRTMT
0
         0
                0
                0
1
         1
2
         0
                0
Proportion of patient treated: 0.0
Test Case 3:
Example df:
   outcome
           TRTMT
0
         0
                0
                1
2
         1
                0
Proportion of patient treated: 0.25
Error: Data-type mismatch.
 4 Tests passed 1 Tests failed
```

```
AssertionError
                                              Traceback (most recent call last)
    <ipython-input-20-e01969b005ef> in <module>()
      1 ### test cell ex1: you cannot edit this cell
---> 2 proportion_treated_test(proportion_treated)
    ~/work/W1A1/public_tests.py in proportion_treated_test(target)
     59
            ]
     60
           multiple_test(test_cases, target)
---> 61
     62
     63
    ~/work/W1A1/test_utils.py in multiple_test(test_cases, target)
    121
                print('\033[92m', success," Tests passed")
                print('\033[91m', len(test\_cases) - success, "Tests failed")
    122
                raise AssertionError("Not all tests were passed for {}. Check your equation
--> 123
```

AssertionError: Not all tests were passed for proportion\_treated. Check your equations

### **Expected Output:**

```
Test Case 1:
```

# Example df: outcome TRTMT 0 0 0 1 1 1 2 1 1 3 1 1

Proportion of patient treated: 0.75

Test Case 2:

Example df:

outcome	TRTMT
0	0
1	0
0	0
0	0
	0 1 0

Proportion of patient treated: 0.0

#### Test Case 3:

Example df:

	outcome	TRTMT
0	0	0
1	1	1
2	1	0
3	1	0

Proportion of patient treated: 0.25

All tests passed.

Next let's run it on our trial data.

Proportion Treated: 0.49093904448105435 ~ 49%

### Exercise 2 - event\_rate

Next, we can get a preliminary sense of the results by computing the empirical 5-year death probability for the treated arm versus the control arm.

The probability of dying for patients who received the treatment is:

$$p_{\mathrm{treatment, death}} = \frac{n_{\mathrm{treatment, death}}}{n_{\mathrm{treatment}}}$$

- *n*<sub>treatment,death</sub> is the number of patients who received the treatment and died.
- *n*<sub>treatment</sub> is the number of patients who received treatment.

The probability of dying for patients in the control group (who did not received treatment) is:

$$p_{\text{control, death}} = \frac{n_{\text{control, death}}}{n_{\text{control}}}$$

-  $n_{\text{control,death}}$  is the number of patients in the control group (did not receive the treatment) who died. -  $n_{\text{control}}$  is the number of patients in the control group (did not receive treatment).

Returns:

```
treated_prob (float): empirical probability of death given treatment
                 untreated_prob (float): empirical probability of death given control
             treated_prob = 0.0
             control_prob = 0.0
             ### START CODE HERE (REPLACE INSTANCES OF 'None' with your code) ###
             treated_prob = len(df[(df["TRTMT"] == True) & (df["outcome"] == 1)]) / len(df[df["TRTMT"]
             control_prob = len(df[(df["TRTMT"]==False) & (df["outcome"]==1)]) / len(df[df["TR"]
             ### END CODE HERE ###
            return treated_prob, control_prob
In [23]: ### test cell ex2: you cannot edit this cell
        event_rate_test(event_rate)
Test Case 1:
Example df:
   outcome TRTMT
0
        0
        1
1
2
        1
3
        0
               1
4
       1
               0
5
        1
               0
6
        1
                0
7
        0
Treated 5-year death rate: 0.5
Control 5-year death rate: 0.75
Error: Data-type mismatch. in variable 0, Got <class 'float'> but expected type <class 'numpy
Error: Data-type mismatch. in variable 1, Got <class 'float'> but expected type <class 'numpy
2 Tests passed 1 Tests failed
                                                  Traceback (most recent call last)
        AssertionError
        <ipython-input-23-fb106dbcb840> in <module>()
          1 ### test cell ex2: you cannot edit this cell
   ---> 2 event_rate_test(event_rate)
```

```
~/work/W1A1/public_tests.py in event_rate_test(target)
        102
        103
    --> 104
                multiple_test(test_cases, target)
        105
        106
        ~/work/W1A1/test_utils.py in multiple_test(test_cases, target)
                    print('\033[92m', success," Tests passed")
        121
                    print('\033[91m', len(test_cases) - success, " Tests failed")
        122
                    raise AssertionError("Not all tests were passed for {}. Check your equation
    --> 123
        AssertionError: Not all tests were passed for event_rate. Check your equations and avo
Expected Output:
Test Case 1:
Example df:
    outcome TRTMT
         0
                1
         1
                1
         1
         0
         1
         1
                0
         1
                0
         0
                0
Treated 5-year death rate: 0.5
Control 5-year death rate: 0.75
All tests passed.
  Now let's try the function on the real data.
In [24]: treated_prob, control_prob = event_rate(data)
         print(f"Death rate for treated patients: {treated_prob:.4f} ~ {int(treated_prob*100)}
         print(f"Death rate for untreated patients: {control_prob:.4f} ~ {int(control_prob*100
```

On average, it seemed like treatment had a positive effect.

Death rate for treated patients: 0.3725 ~ 37% Death rate for untreated patients: 0.5000 ~ 50%

0

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