### Title Pending

#### Using Artificial Intelligence to predict test result

#### By Lukas Schießer

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Thesis Supervisor: Johannes Thesis Supervisor:

### Abstract

# Acknowledgements

I want to thank...

### Contents

1	Intr	roduction	7			
2	Methods					
	2.1	Data	8			
	2.2	Multivariate Imputation by Chained Equations	9			
	2.3	K-fold nested cross validation	9			
	2.4	Classifiers used	9			
		2.4.1 Random Forest	9			
		2.4.2 Logistic regression	9			
3	Res	m ults	10			
	3.1	Results of own implementation	10			
	3.2	Comparison with original paper	10			
4	Discussion					
	4.1	Discussion of Results	11			
	4.2	Scenarios for real-world validation	11			
${f A}$	Apı	pendix	13			

# List of Figures

### List of Tables

2.1	Descriptive	statistics for	numerical	features in	data set								8
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### Introduction

Introduction for this thesis by explaining motivation for this topic and giving a short overview.

Motivation: overwhelmed health care systems, shortages of testing supplies, AI as supplementary technique to help.

#### Methods

#### 2.1 Data

The data used to train the classifiers was provided by Brinati et al. [1]. It was collected between the end of February 2020 and mid of March 2020 from patients admitted to the *IRCSS Ospedale San Raffaele* and consists of 279 individuals who were selected randomly. For each individual, the data set provides their age, gender, results of a routine blood screening, and the result of a PCR test for Sars-CoV-2. Any other datapoint that could possibly identify the patient, i.e., date of admittance or date of PCR test were not recorded. Table 2.1 provides common satisfies for the numerical features of the data set.

Feature	$\operatorname{Unit}$	Mean	$\operatorname{Std}$	Median
Age	Years	61.78	17.81	64
White Blood Cell Count (WBC)	$10^{9}/{\rm L}$	8.55	4.86	7.10
Platelets	$10^{9}/{\rm L}$	226.5	101.2	205.00
Neutrophils	$10^{9}/{\rm L}$	6.20	4.17	5.10
Lymphocytes	$10^{9}/{\rm L}$	1.19	0.80	1.00
Monocytes	$10^{9}/{\rm L}$	0.61	0.41	0.50
Eosinophils	$10^{9}/{\rm L}$	0.06	0.13	0.00
Basophils	$10^{9}/{\rm L}$	0.01	0.04	0.00
C-reactive protein (CRP)	mg/L	90.89	94.42	54.20
Aspartate Aminotransferase (AST)	U/L	54.20	57.61	36.00
Alanine Aminotransferase (ALT)	U/L	44.92	45.50	31.00
Alkaline Phosphatase (ALP)	U/L	89.89	89.09	71.00
Gamma Glutamil Transferasi (GGT)	U/L	82.48	132.70	41.00
Lactate dehydrogenase (LDH)	U/L	380.45	193.98	328.00

Table 2.1: Descriptive statistics for numerical features in data set

# 2.2 Multivariate Imputation by Chained Equations

Describe MICE algorithm (keine Bewertung, findet in Discussion statt)

Maybe include PMM?

#### 2.3 K-fold nested cross validation

#### 2.4 Classifiers used

#### 2.4.1 Random Forest

#### 2.4.2 Logistic regression

### Results

- 3.1 Results of own implementation
- 3.2 Comparison with original paper

### Discussion

- 4.1 Discussion of Results
- 4.2 Scenarios for real-world validation

### Declaration

I declare that..

# Appendix A

# Appendix

### Bibliography

[1] D. Brinati et al. "Detection of COVID-19 Infection from Routine Blood Exams with Machine Learning: A Feasibility Study". In: J Med Syst 44.8 (2020), p. 135. ISSN: 1573-689X (Electronic) 0148-5598 (Linking). DOI: 10.1007/s10916-020-01597-4. URL: https://www.ncbi.nlm.nih.gov/pubmed/32607737,https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7326624/pdf/10916\_2020\_Article\_1597.pdf,https://zenodo.org/record/3886927#.X3xhPO1CRPY.