

# Analyses and Management of Healthcare Data for Cancer Care

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# A Brief Overview

## Front-End



Visualisation  
Reporting Service  
Dashboard

## Back-End



Migration  
API Gateway

Toxicity Predictor  
(1,2,3,4,...)  
ML, NLP, DL

Issues...  
Issues...

PoC

Synthetic Data



# Previous Projects

- Developed Patient timeline Visualisation, the SESO-Gateway
- The Oncology DB migration
- Designed and developed a reporting service for the Oncology DB

# On Predicting the Outcomes of Chemotherapy Treatment in Breast Cancer

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# Outline

- Project Overview
- Related Work
- Data Analysis
- Models Creation
- Result
- Conclusion



# Project Overview

- Cancer is
  - a mutation caused by an abnormal reproduction of cells.
  - happens in in different organs (e.g., **breast**, lungs, bone, etc.)
- The treatments vary from surgery with **chemotherapy** and/or radiotherapy (i.e. usually take a long time and in sequence)
  - However some treatments are toxic and expensive
- We **compared several different techniques** (Markov Model, HMM, RF, RNN) applied to the same data set to **predict the toxicity outcome of different treatment options**

# Related Work

- Many ongoing research looks at prediction of cancer treatment outcomes
  - Bayesian Logistic Regression (Subramani et al.)
  - Random Forest (Hui-Ling Chen et al.)
  - SVM (Nguyen et al.)
- We also consider **HMM** and **RNN** (common in different fields like NLP)



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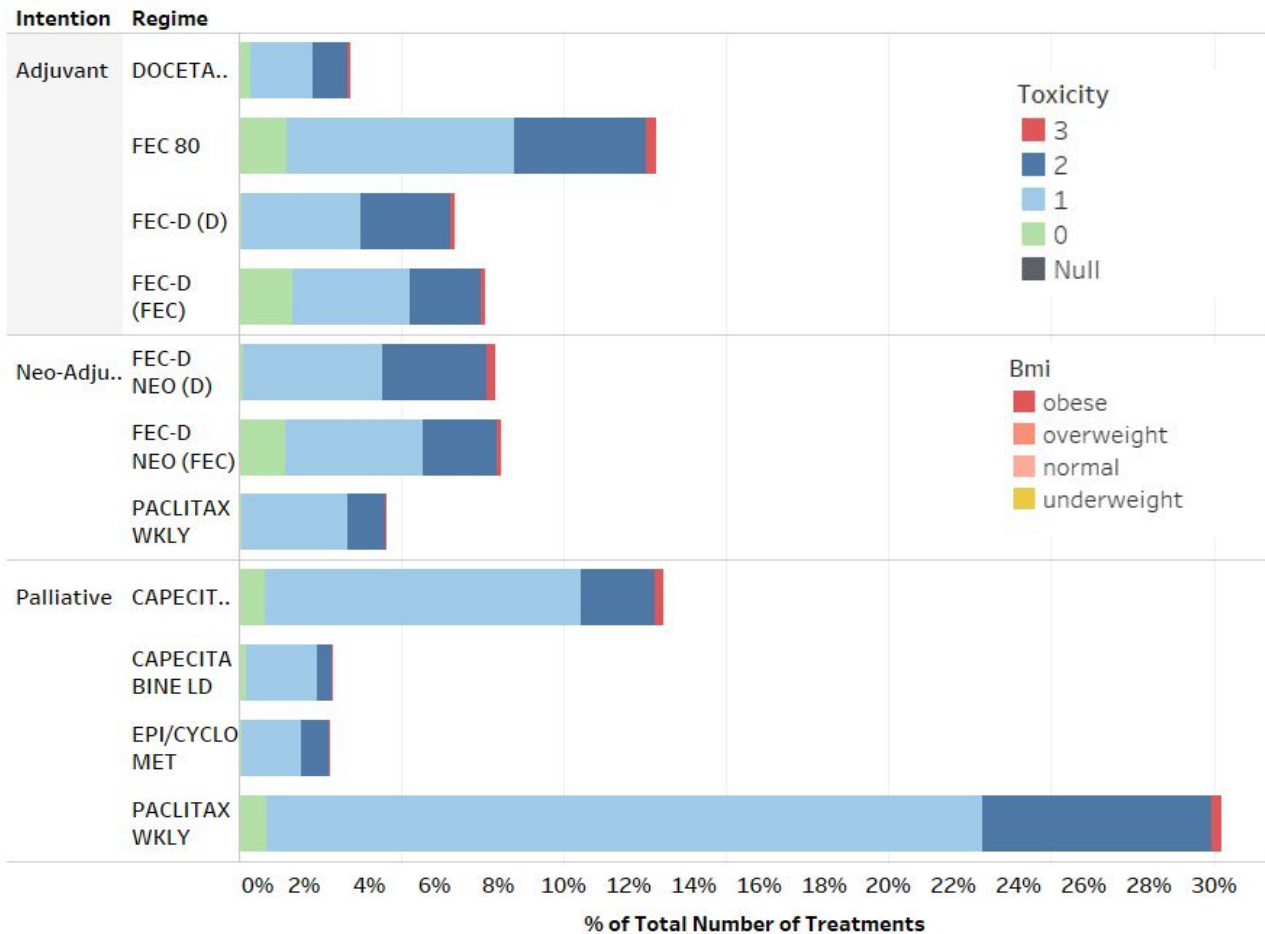


# Data Analysis

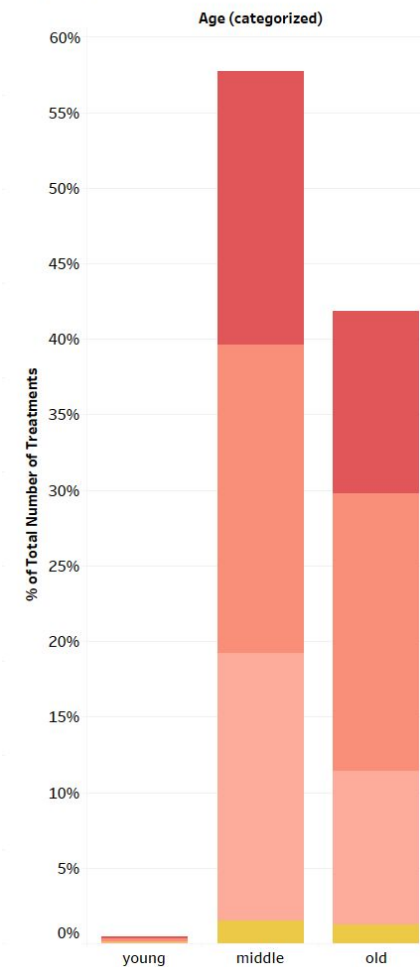
- We use data extraction from an oncology department in Scotland
  - 3 years (2014 - 2016 )
  - Includes various observations concerning breast cancer treatments (e.g. **intention, regime, cycles**), recorded side effects (here, **toxicity level**), and patient characteristics (e.g. **age, BMI, performance status**).

Intention	Number of Treatments	Number of Patients
Adjuvant	1209	205
Neo-adjuvant	1855	382
Palliative	2752	213

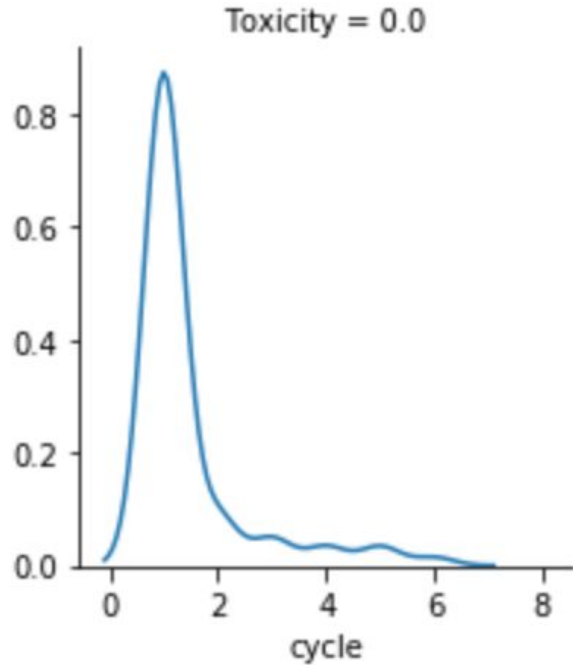
# Toxicity proportion



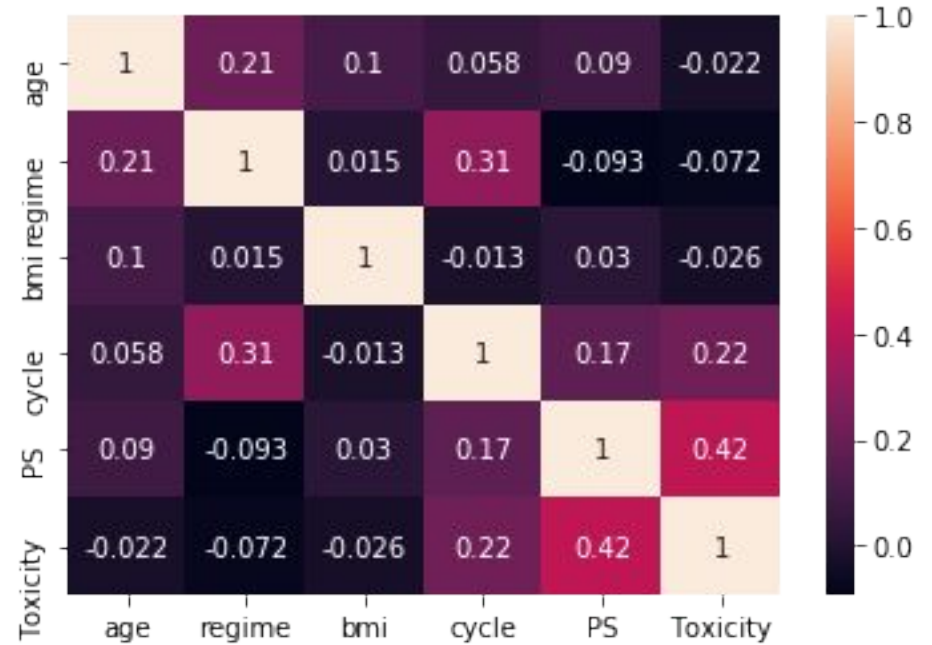
## Age - BMI



# Features correlation



Treatment cycle - toxicity (No toxicity)

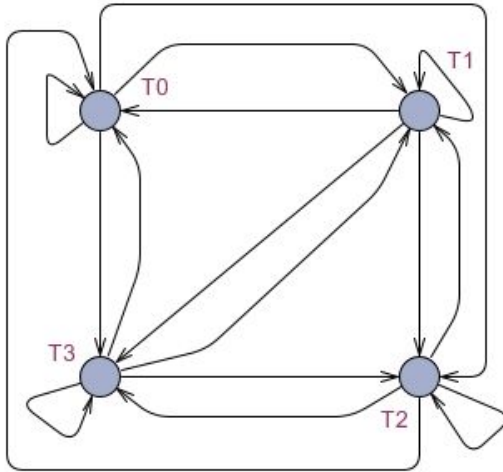


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# Markov Model (MM)



- A stochastic model with the assumption that a future state only depends on the current state.
- T0, T1, T2, T3 denote the toxicity state(i.e. No toxicity, Low, Medium, High)

The transition probability matrix

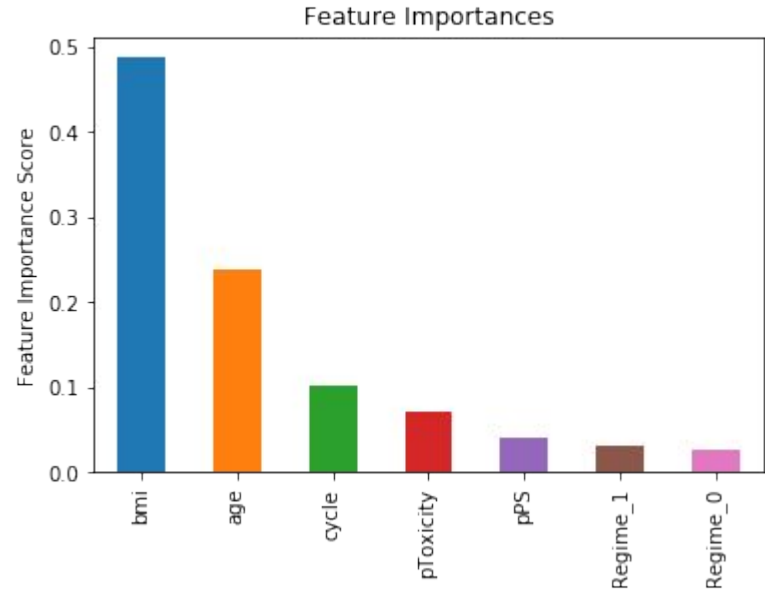
# Hidden Markov Model (HMM)

- Based on augmenting a Markov chain to observe the hidden states of events
- Our HMM components:
  - States: **T0, T1, T2, T3**
  - Transitions: from **T0 to T1**, from **T1 to T3**, etc
  - Observations: **cycle, age, BMI, regime** (categorised and coded). For example, 1-2-3-1 denotes the observation for an overweight patient who gets the FEC-D (D) in their first cycle and is aged less than 50 years
  - Emissions: the probability of the observations generated from the toxicity state



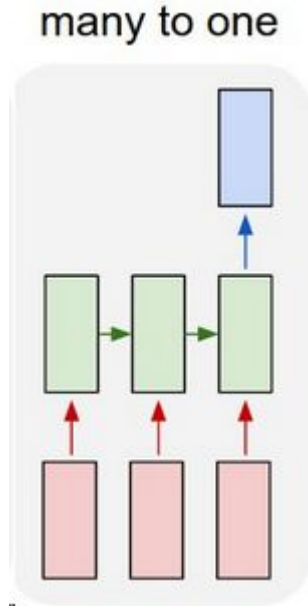
# Random Forest

- An ensemble of decision trees for solving classification problems.
- We created three RF models for each treatment intention (i.e., adjuvant, neoadjuvant, palliative)
  - Predictors: **age, BMI, Regime, cycle, previous toxicity**, and previous performance status.
  - Outcome: **Patients' toxicity**



# Recurrent Neural Network (RNN)

- A class of NN where connections between nodes form a directed graph along a temporal sequence.
- Implemented using **tensorflow LSTM** module.
- Used similar features as for our RF model.  
However, **we do not use** the previous **performance status** and **previous toxicity fields**.



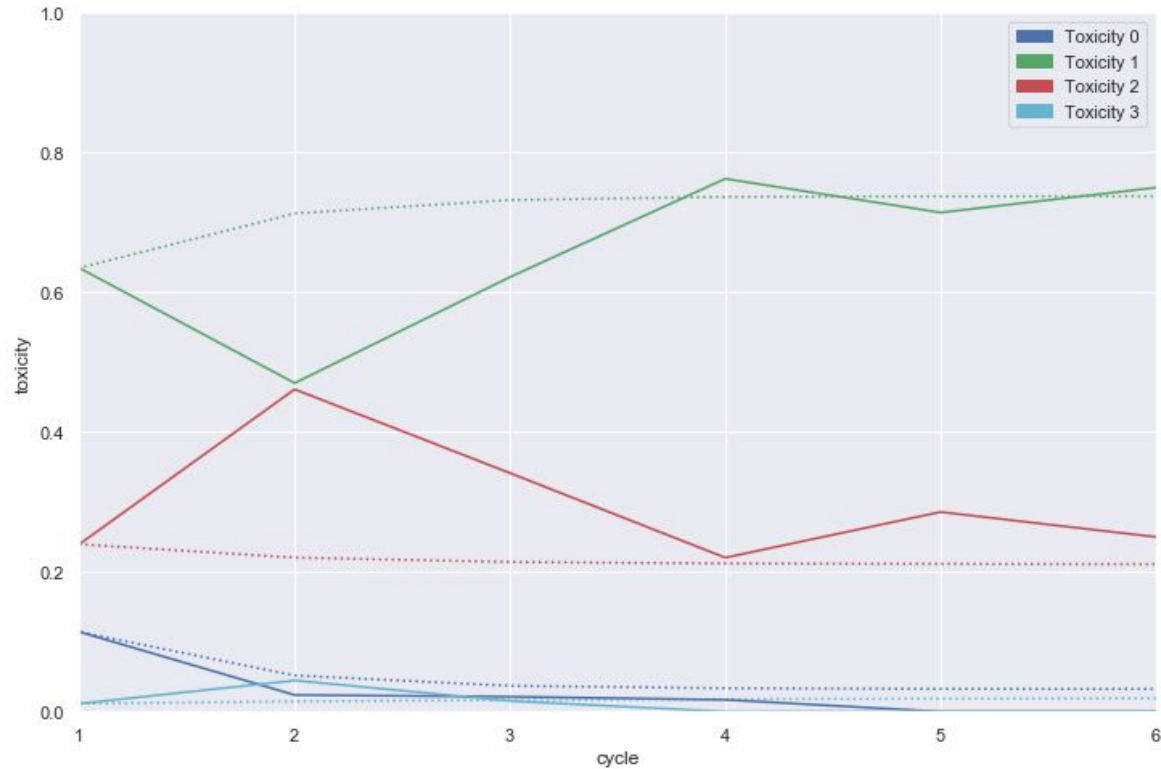


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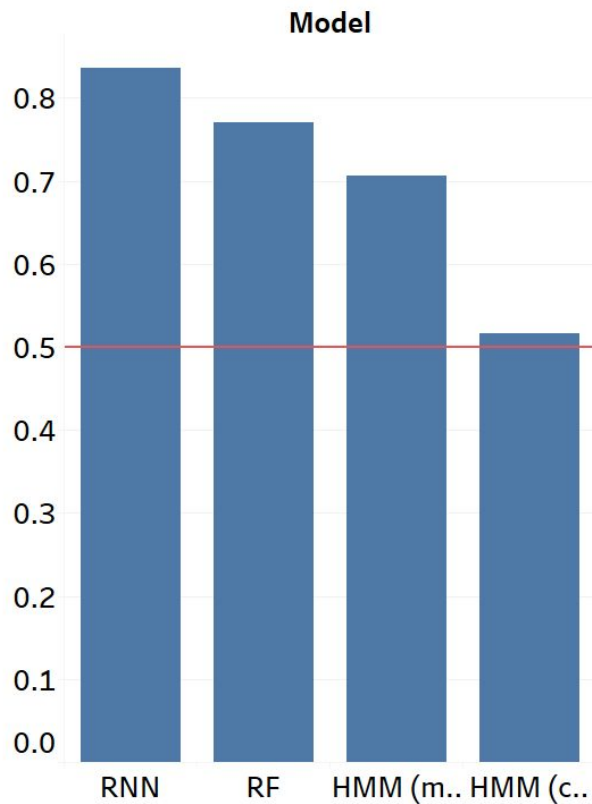
# Markov Model



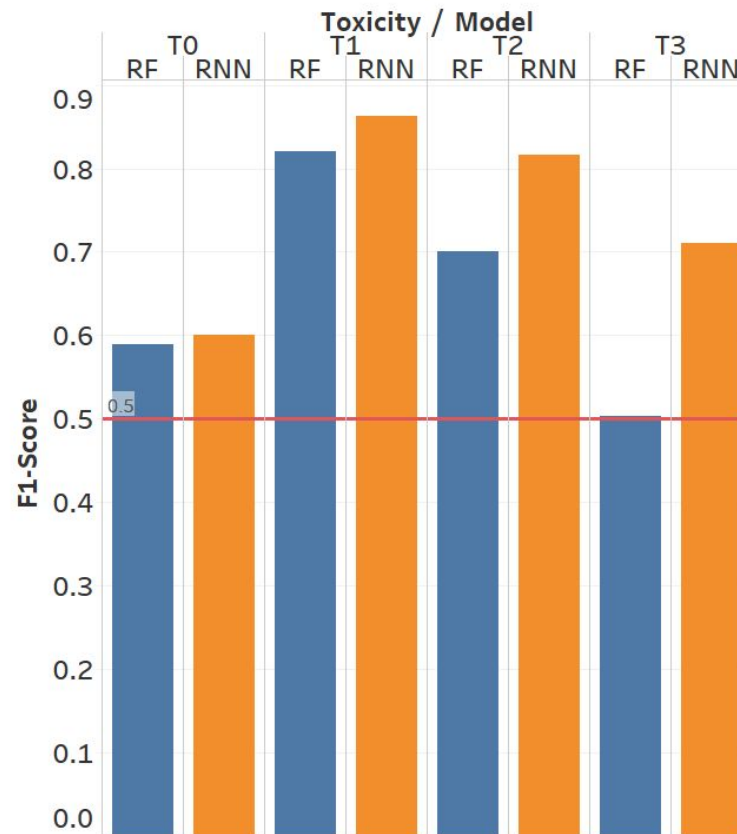
Palliative treatments distribution

# Classifier

Accuracy



F1-Score



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# Conclusion

- Our classifiers can predict the toxicity outcome of the chemotherapy outcomes with around **0.8 - 0.85** accuracy.
- The RNN model performed better than all other models because it considers all the patients' treatments
- In contrast to the MM, the classifiers are more tailored for an individual patient.
  - Both the MM and the classifiers are a complement to each other.



## Future Work

- Improve the accuracy further by integrating more data regarding the cancer characteristics and patients' comorbidity.
- Create a dashboard and/or reporting system which can be helpful to the physician as a second opinion to decide which regime is suitable for an individual patient.

# Ongoing Work

Our aim is to use a **synthetic data** during the development of the Application.



## DEMO

<http://localhost:8000/>

<https://breast.predict.nhs.uk/tool>

Toxicity Predictor

← → ↺ ⌂ <http://mock-toxicity.com>

GP Names ▼ Patient Name/ID 🔍

If no patient is selected this section shows the help menu

Can be either name or CHI or other identifier

Timeline of the Patient

**Gender** gender ▼

Age at diagnosis Text

weight Text

height Text

SIMD Text

**Treatment**

Intention Text

Regime Text

Duration of Treatment Text

Add drug information

Future work

**General Condition**

Performance status PS ▼

initial toxicity toxicity: 0-4 ▼

latest toxicity toxicity: 0-4 ▼

Analyse

**Initial State**

Site Site ICD-10 ▼

Tumour size (mm) Text

Nodes Text

Relapse/Metastasis Text

**Cormobidities**

Other diagnosis Site ICD-10 ▼

Charlson quan score Text

Add the characteristic related to the cancer site

ER status Text

HER2 status Text

This is an example for the breast cancer

This section shows the predictor fields explanation

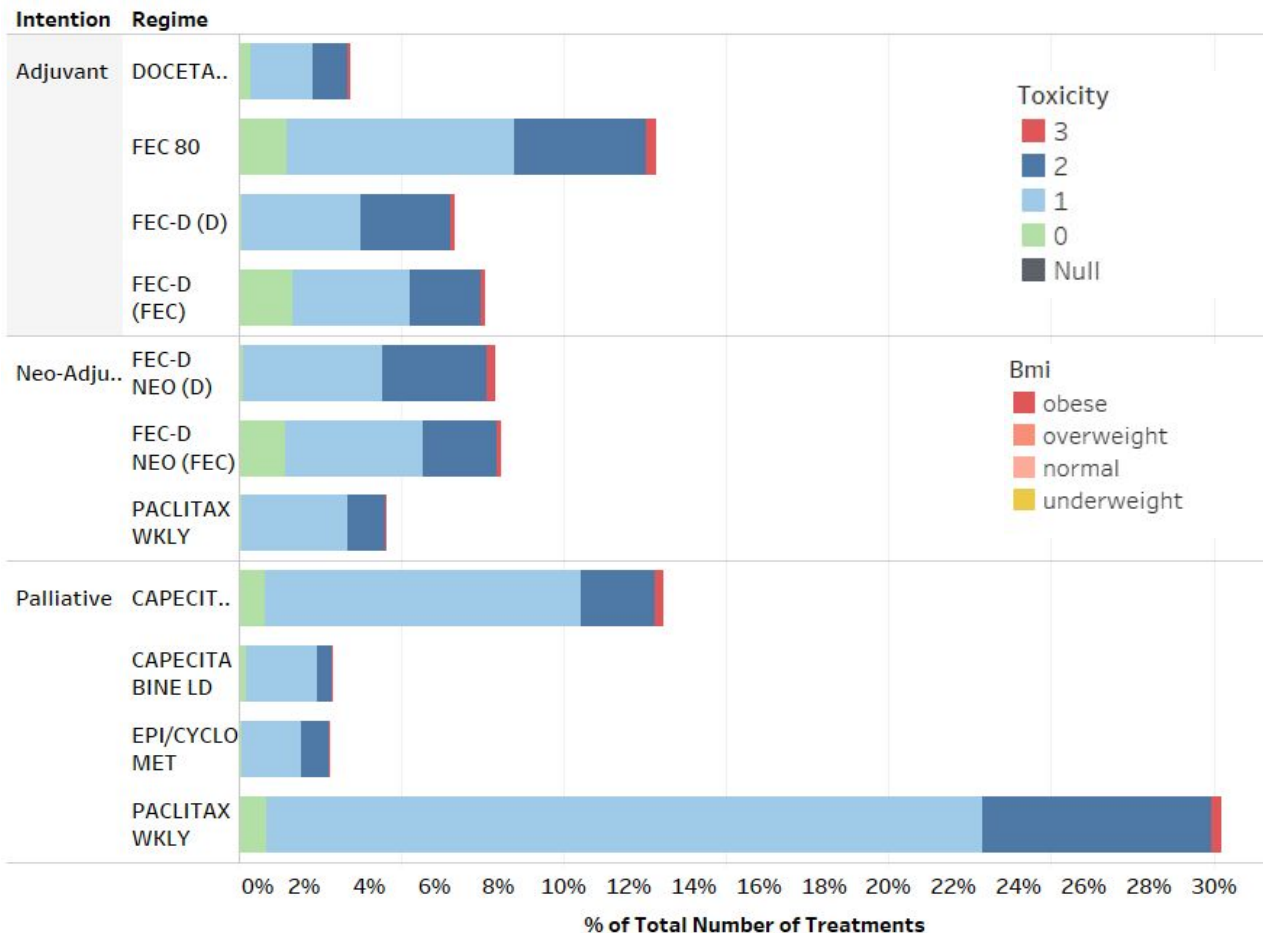


# Issues

- **Missing Values**
  - Solution: regression, removing some instances
- **Class imbalance**
  - Solution: duplication for some classes
- **Overfitting**
  - Solution: Cross Validation, using more data



# Toxicity proportion





# And something like this occasionally happens as well....

```
import {EventEmitter, Injectable} from '@angular/core';
import { HttpClient } from '@angular/common/http';
import {TreatmentModel} from '../shared/treatment.model';

@Injectable()
export class TreatmentService {
  // TODO: DYNAMIC UPDATE THE URL FROM CONFIG
  private API_URL = 'assets/treatments.json';

  // EVEN THOUGH MY HEART SCREAMS AT THIS PUBLIC ELEMENT... I NEED TO EMIT THE MAP,
  // LETTING IT FREE TO THE OTHER MODULE...
  // BECAUSE IT'S MORE CONVENIENT... I KNOW SOMEDAY IT WILL COME BACK AT ME
  // BUT FOR NOW, LET'S JUST HOPE FUTURE AGASTYA HANDLE IT WELL.
  public treatmentMap = new Map<string, TreatmentModel>();
  private mapEmitter: EventEmitter<Map<string, TreatmentModel>> = new EventEmitter();

  constructor(private http: HttpClient) {}
```