

# OSDA Big Homework: Neural FCA

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December 13, 2022

**Used dataset:** openly available ([\[1\]](#)), with 6 attributes (5 binary and 1 numerical) and 2 target columns, with 120 rows.

**The main goal** of this dataset is to help predict two diseases of urinary system: an acute inflammations of urinary bladder (or cystitis in medical terms) and an acute nephritis.

## Dataset attributes and their values:

1. Temperature of patient - int values in the segment [35.5,41.5];
2. Occurrence of nausea - "yes" or "no";
3. Lumbar pain - "yes" or "no";
4. Urine pushing (continuous need for urination) - "yes" or "no";
5. Micturition pains - "yes" or "no";
6. Burning of urethra, itch, swelling of urethra outlet - "yes" or "no".

## Our targets for binary classification:

1. Inflammation of urinary bladder (cystitis) - "yes" or "no";
2. Nephritis of renal pelvis origin - "yes" or "no".

In this work I will focus only on the prediction of cystitis.

# Binarization strategy and prediction quality measure

**Binarization strategy.** According to [2], let's divide all data in column "Temperature of patient" into 2 groups (answers the question whether the patient has a temperature):

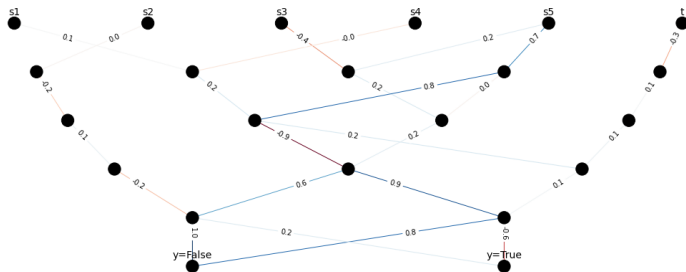
- "no" if temperature  $\in [35.5, 37.2]$ ;
- "yes" if temperature  $\in [37.3, 41.5]$ .

**Prediction quality measure.** I prefer to use the **F1 score** because it maintains a balance between precision and recall for the classifier and it gives a better measure of the incorrectly classified cases than the accuracy metric.

# First results

**Important comment.** Here and further I take the minimum possible count of best concepts and train neural networks 50000 epochs.

7 best concepts gives **F1 score**  $\approx 0.67$ .

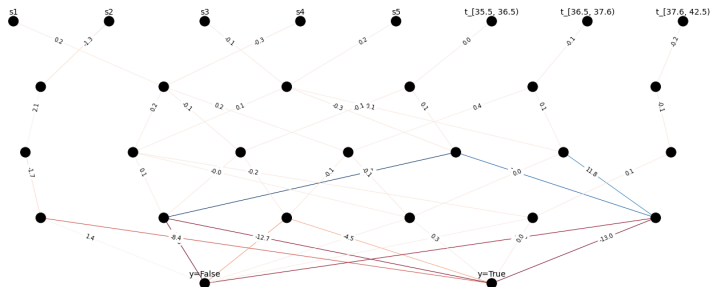


# Another attribute binarization

Based on [3], let's divide "Temperature of patient" into 3 groups:

- $t \in [35.5, 36.4]$ ;
- $t \in [36.5, 37.5]$ .
- $t \in [37.6, 41.5]$ .

12 best concepts gives **F1 score**  $\approx 0.47$ .



Neural network with fitted edge weights for model 2

# Another technique to select best concepts

Let's try to **select best concepts according to accuracy**. So, the results are:

- based on model 1 with **9** best concepts (let's name new model as model 3.1) - **F1 score**  $\approx 0.51$ , **accuracy**  $\approx 0.38$ ;
- based on model 2 with **14** best concepts (let's name new model as model 3.2) - **F1 score**  $\approx 0.59$ , **accuracy**  $\approx 0.42$ .

Neural network graphs see in Appendix 1.

# The efficiency of various nonlinearities to put in the network

I decide to use **Leaky ReLU** and **hyperbolic tangent** instead of base ReLU. So, the results are

1. based on model 1 for both nonlinearities we get **F1 score = 1** (let's name them model 4 and model 5 for Leaky ReLU and hyperbolic tangent respectively - here we get **7** best concepts);
2. based on model 2 for both nonlinearities we get **F1 score = 1** (similarly name them model 6 and model 7 - here we get **12** best concepts).

Neural network graphs see in Appendix 2.



# Comparison of the prediction quality of the proposed model with State-of-the-Art approaches

I compare proposed model with results of **Logistic Regression** and **Random Forest Classifier** of default parameters.

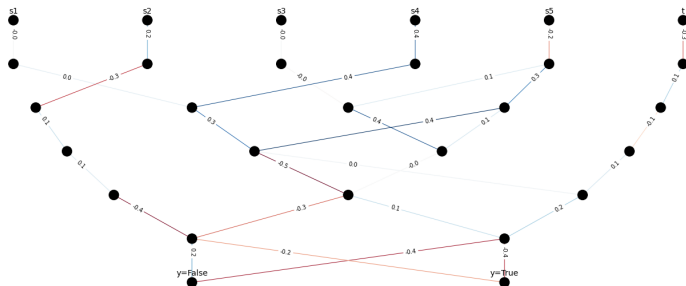
In both cases I get **F1 score = 1** **BUT** these models work **faster** than proposed ones.

# References:

1. <https://archive.ics.uci.edu/ml/datasets/Acute+Inflammations>
2. <https://www.tuasaude.com/en/how-to-tell-if-you-have-a-fever/>
3. [https://en.wikipedia.org/wiki/Human\\_body\\_temperature](https://en.wikipedia.org/wiki/Human_body_temperature)
4. My code and dataset can be find in my GitHub repository <https://github.com/thecrazymage/Neural-FCA>

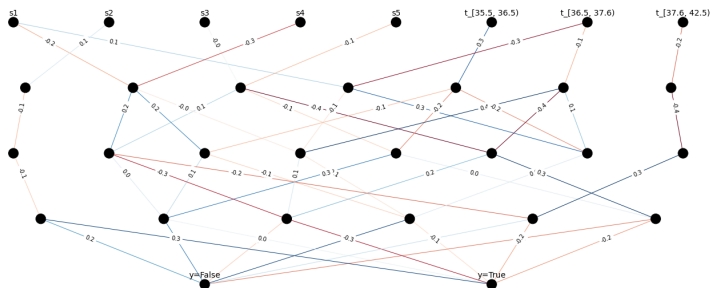
*Thank You For Your Attention!*

# Appendix 1



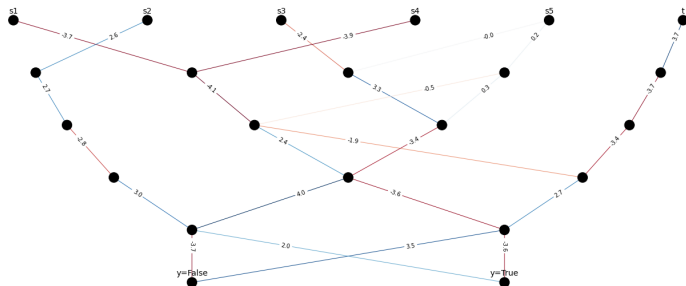
Neural network with fitted edge weights for model 3.1

## Appendix 1



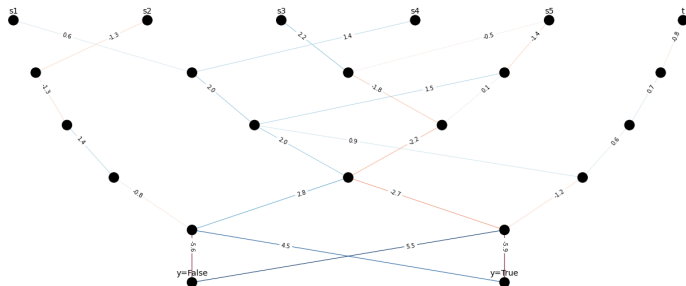
Neural network with fitted edge weights for model 3.2

## Appendix 2



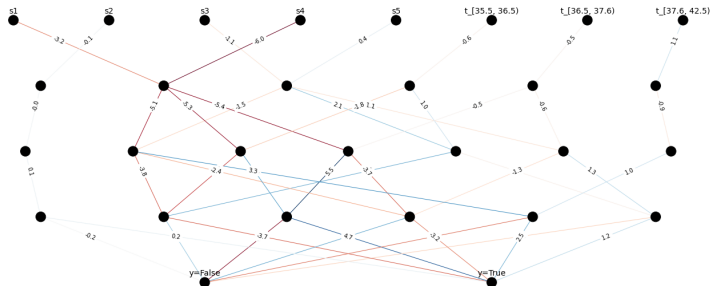
Neural network with fitted edge weights for model 4

## Appendix 2



Neural network with fitted edge weights for model 5

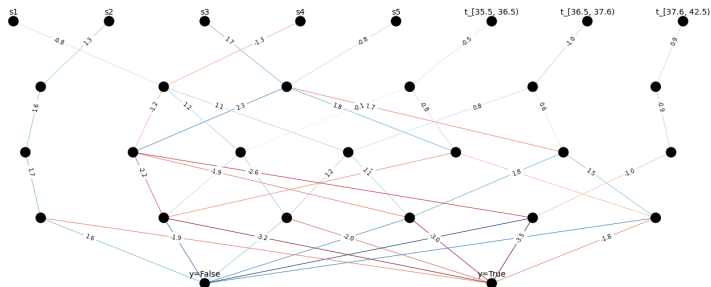
## Appendix 2



Neural network with fitted edge weights for model 6



## Appendix 2



Neural network with fitted edge weights for model 7