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MINISTRY OF HIGHER EDUCATION AND SCIENTIFIC RESEARCH  
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DEPARTMENT OF COMPUTER SCIENCE



MASTER'S INTELLIGENT SYSTEMS ENGINEERING

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**KNOWLEDGE REPRESENTATION 02 (RC2)**

PROJECT REPORT

**BAYESIEN NETWORK  
IMPLEMENTATION on the  
BREAST CANCER DATASET**

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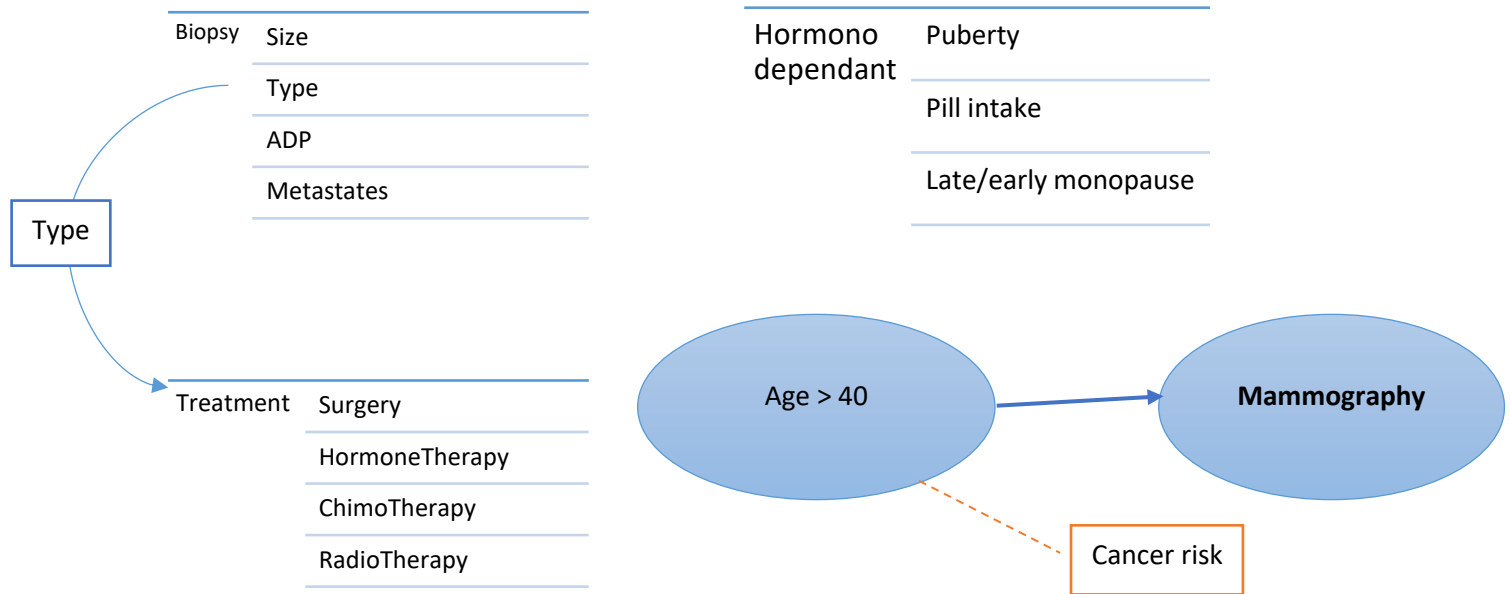
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## 1. BUILDING THE BAYESIEN NETWORK :

### 1.1. CONSULTING THE EXPERT :

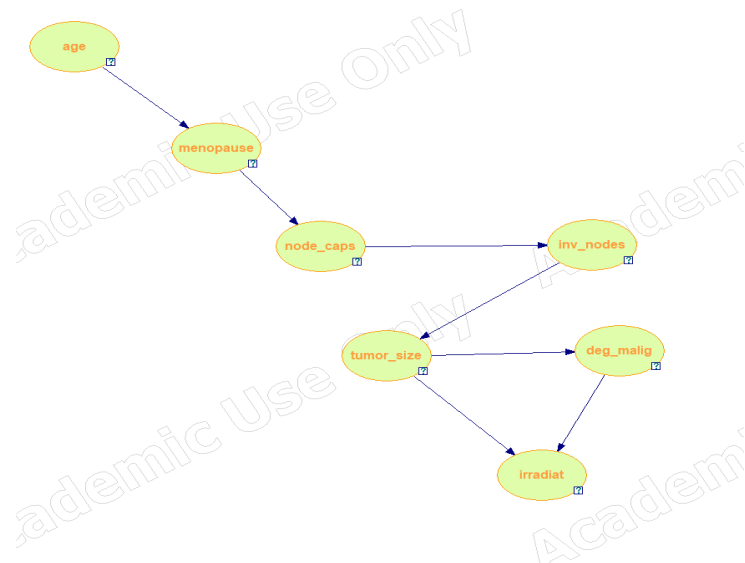
The doctor's goal was to provide us with a comprehensive understanding of cancer diagnosis in general, covering everything from its causes to detection methods, determining its stage and size, and identifying suitable treatment options.

Here is a look on some of the information that was provided by the expert :



### 1.2. BUILDING THE INITIAL NETWORK

After a long discussion with the doctor we agreed finally on a suitable network structure that would be as efficient for cancer detection without having a redundancy in the information that would influence or mislead the results :

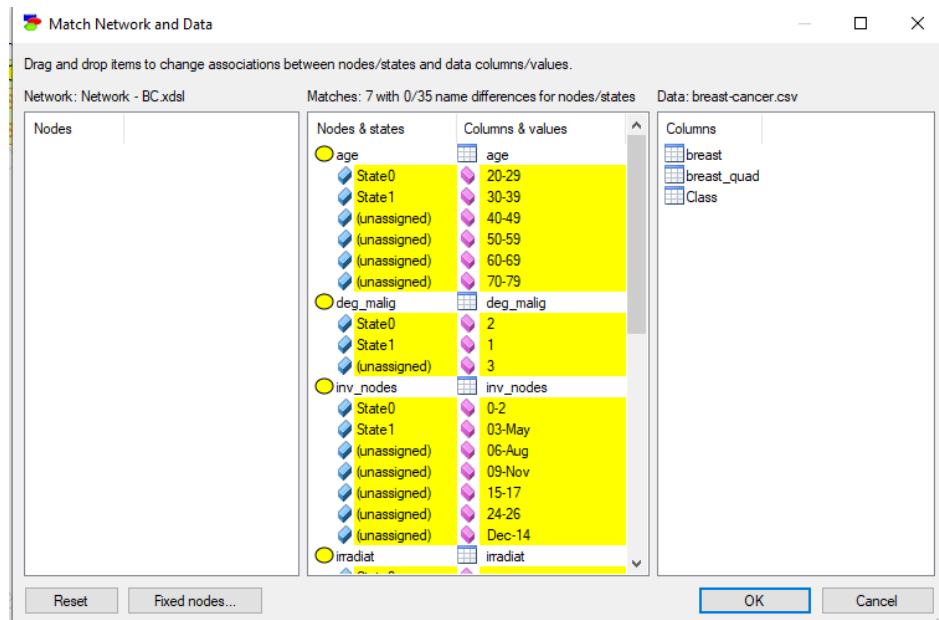


## 2. LOADING DATA AND LEARNING PARAMETERS

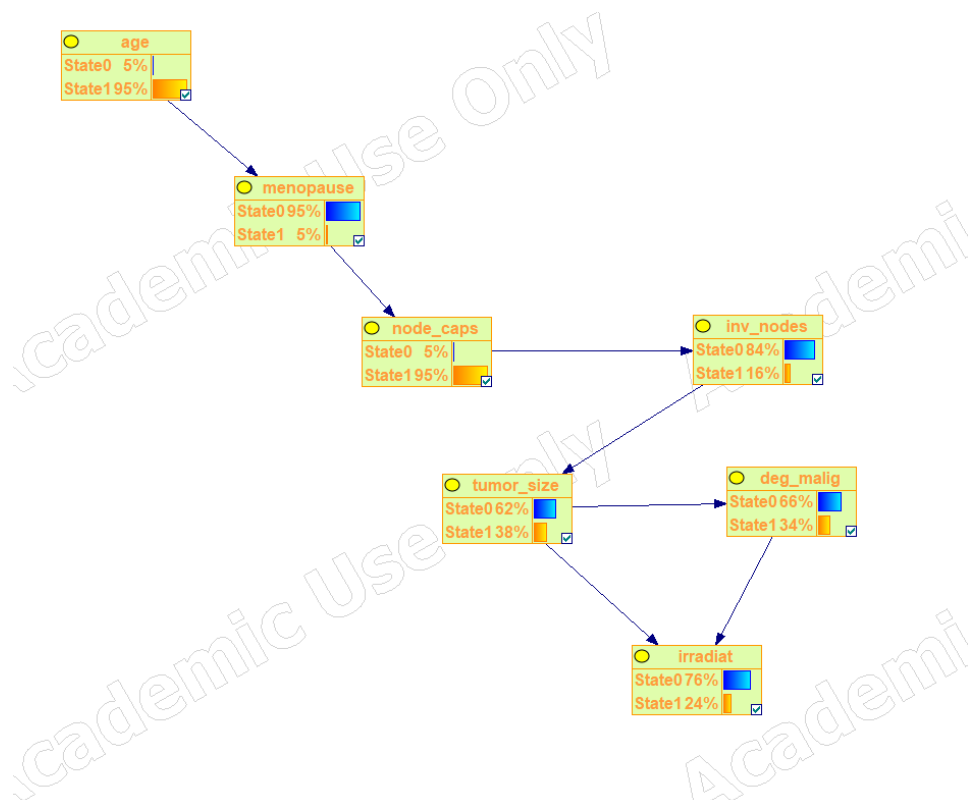
### 2.1. ON GeNIe

#### 2.1.1. Loading data

After loading the data we went to the parameter learning phase , we entered the data and checked if both the nodes and the data headers are under the same name and executed .

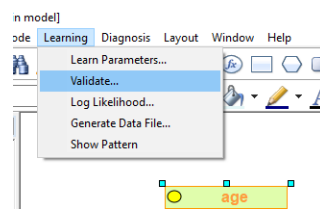


We got the following values :

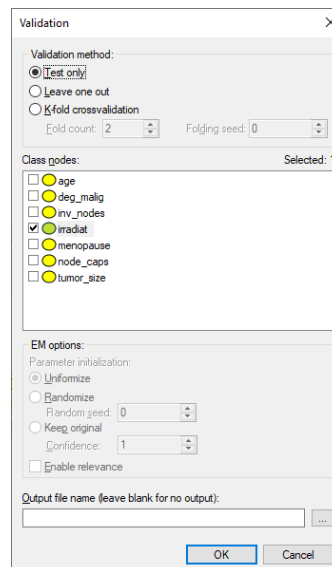


### 2.1.2. Testing

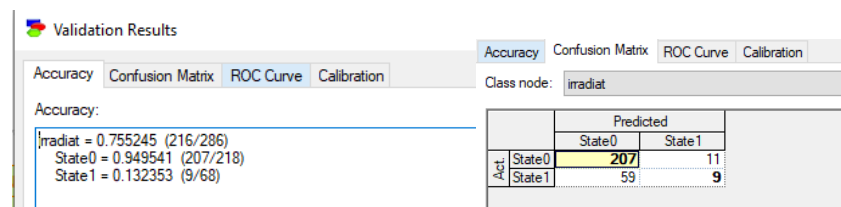
Started by using the Validate method found in GENIE's software ,after loading all the data



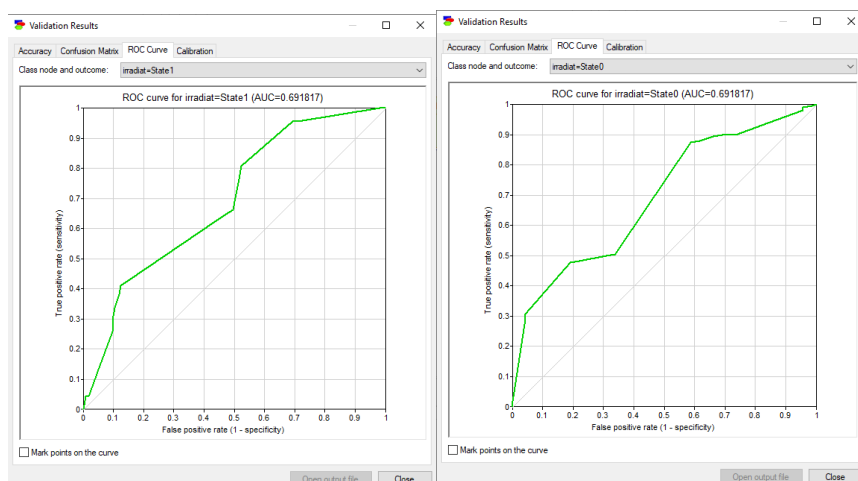
And then choosing the target value of which we would like to do the training on



Having the following results :



GENIE takes by default the 80% 20% principle , where 80% of the datais for the training and 20% of it is for the testing .



## 2.2. Using PGMPY

### 2.2.1. loading libraries

```
import pandas as pd
import numpy as np
from pgmpy.models import BayesianModel
from pgmpy.models import BayesianNetwork
from pgmpy.estimators import ExpectationMaximization
from pgmpy.inference import VariableElimination
from pgmpy.metrics import correlation_score ,log_likelihood_score
import matplotlib.pyplot as plt
import networkx as nx
```

-The 'pandas' 'numpy' libraries are for the manipulation of the data , from reading , splitting to attributing treatments .

-The 'pgmpy' library we got the Bayesian model , the Expectation Maximization algorithm , matrices calculating methods .

-Where we used the 'nx' and 'matplotlib' for the visualisation of the result .

### 2.2.2. Defining network structure

```
# Define the structure
model = BayesianNetwork([
    ('age', 'menopause'),
    ('menopause', 'node_caps'),
    ('node_caps', 'inv_nodes'),
    ('inv_nodes', 'tumor_size'),
    ('tumor_size', 'deg_malign'),
    ('tumor_size', 'irradiat'),
    ('deg_malign', 'irradiat')
])
```

The definition of the network structure was based in the previous implementation of the bayesian network .

### 2.2.3. Reading the data and calling the EM algorithm

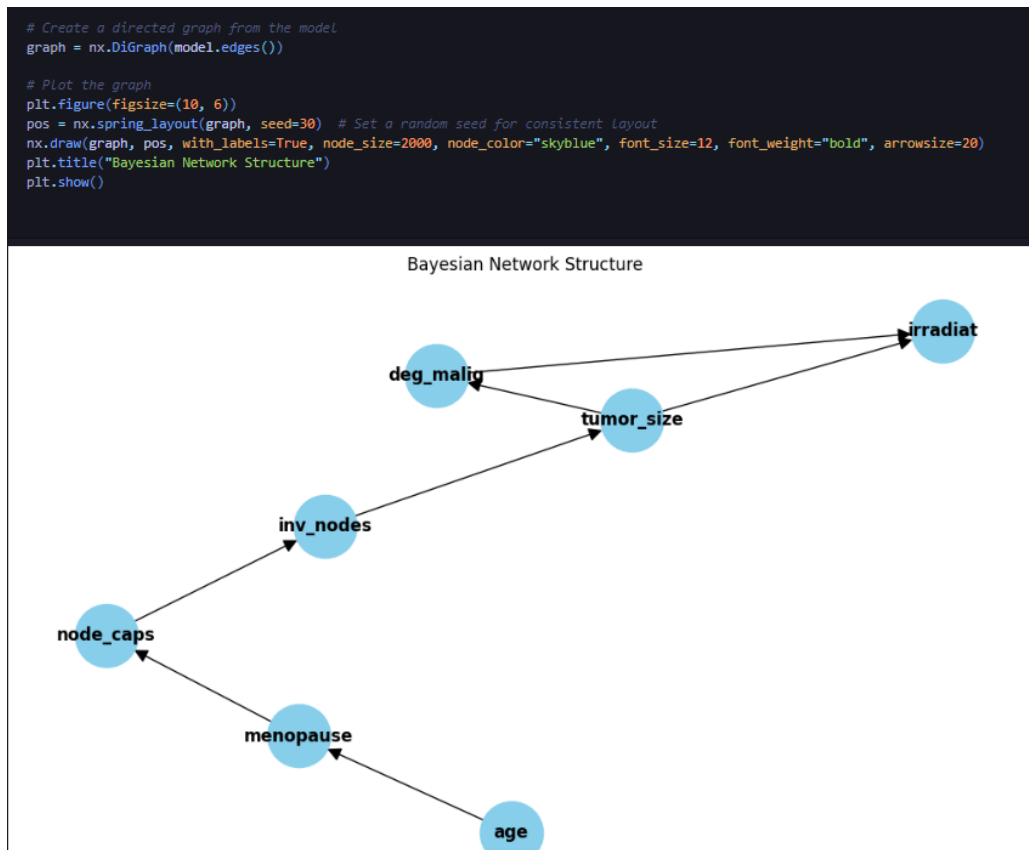
```
# Load the data from the CSV file into a pandas DataFrame
data = pd.read_csv(r"K:\STUDIES\----- M1 AI\S2\RC2\RC2-LAB\TP5\breast-cancer.csv")

# Instantiate the ExpectationMaximization Estimator
estimator = ExpectationMaximization(model, data)
```

Calling the EM algorithm and feeding it the defined model structure and the the headed csv file using the pandas library .

### 2.2.4. Plotting the network structure

Using the nx and matplotlib libraries to define the structure of the network and plotting it .



### 2.2.5. CPD values

```
# Get the CPD values
cpd_values = estimator.get_parameters()

# Print the CPD values
for cpd in cpd_values:
    print("CPD for node:", cpd.variable)
    print(cpd)
```

```
Could not render content for 'application/vnd.jupyter.widget-view+json'
{"model_id":"fe3090d66a414afdbceffd0fe5ad7305","version_major":2,"version_minor":0}
```

```
CPD for node: menopause
+-----+-----+-----+-----+
| age      | age(20-29) | ... | age(60-69) | age(70-79) |
+-----+-----+-----+-----+
| menopause(ge40) | 0.0 | ... | 0.9649122807017544 | 1.0 |
+-----+-----+-----+-----+
| menopause(1t40) | 0.0 | ... | 0.03588771929824561 | 0.0 |
+-----+-----+-----+-----+
| menopause(premeno) | 1.0 | ... | 0.0 | 0.0 |
+-----+-----+-----+-----+
CPD for node: deg_malign
+-----+-----+-----+
| tumor_size | tumor_size(0-4) | ... | tumor_size(Oct-14) |
+-----+-----+-----+
```

### 2.2.6. Validation

For this process splitting the data into training data and testing data is necessary and then performing the testing phase .

```
# Split the data into train and test sets (80% for train, 20% for test)
train_data, test_data = train_test_split(data2, test_size=0.2, random_state=42)
```

Dropping the 'irradiat' class as it is our target on the testing slice of data .

```

test_data.columns

Index(['Class', 'age', 'menopause', 'tumor_size', 'inv_nodes', 'node_caps',
      'deg_malign', 'breast', 'breast_quad', 'irradiat'],
      dtype='object')

test_data.drop(columns=['Class', 'breast_quad', 'breast'], inplace=True)
test_features = test_data.drop(columns=['irradiat'])
test_actual_values = test_data['irradiat']

test_features.columns

Index(['age', 'menopause', 'tumor_size', 'inv_nodes', 'node_caps',
      'deg_malign'],
      dtype='object')

```

Testing the test data on calculating the accuracy of the result .

```

# Make predictions on test data
predictions = []
test_actual_values = test_data[['age', 'menopause', 'tumor_size', 'inv_nodes', 'node_caps', 'deg_malign', 'irradiat']] # Corrected column selection
for index, row in test_features.iterrows():
    row_df = pd.DataFrame([row], columns=test_features.columns) # Convert dictionary to DataFrame with correct column names
    prediction = model2.predict(row_df)
    # Assuming prediction is a dictionary with 'irradiat' as one of the keys
    predictions.append(prediction['irradiat'])

# Evaluate performance
accuracy = accuracy_score(test_actual_values['irradiat'], predictions) # Assuming 'irradiat' is the target column
print("Accuracy:", accuracy)

```