# **Breast Cancer Detection**

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Term: Spring 2023

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Course: CS6220 Data Mining

## Why choose this topic

#### Real life problem

Breast cancer is a **significant public health concern**, affecting millions of women worldwide. Early detection of breast cancer is critical for successful treatment and improving the chances of survival.

A good chance to explore data mining techniques and ML concepts



## Intro

What is breast cancer?

Why we need early detection?

How to use ML to detection breast cancer in my Project?

(current survey and my approach)



## **Surprise Fact:**

→ Rate:

1 out of 10 women will face breast cancer in the life time.

## 52%

decline in prostate cancer from 1993-2015





decline in breast cancer in women from 1989-2015

39%

## 52%

decline in colorectal cancer in both genders from 1970-2015





decline in lung cancer in men from 1990-2015

45%

19%

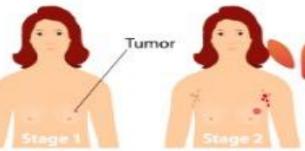
decline in lung cancer in women from 2002-2015

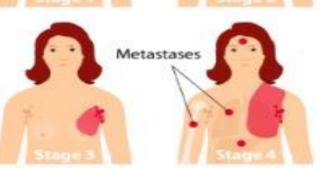


# 2 AAAAA

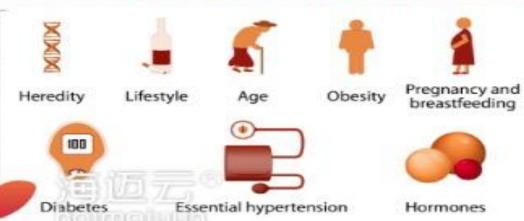
1 out of 10 women have a breast cancer

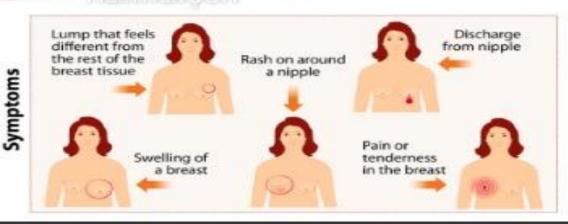
Risk





#### **All About Breast Cancer**





# Why we need early detection?

Stage	5-Year Relative Survival Rate
Stage 0	100%
Stage 1	100%
Stage 2	93%
Stage 3	72%
Stage 4	22%
	Source: American Cancer Society

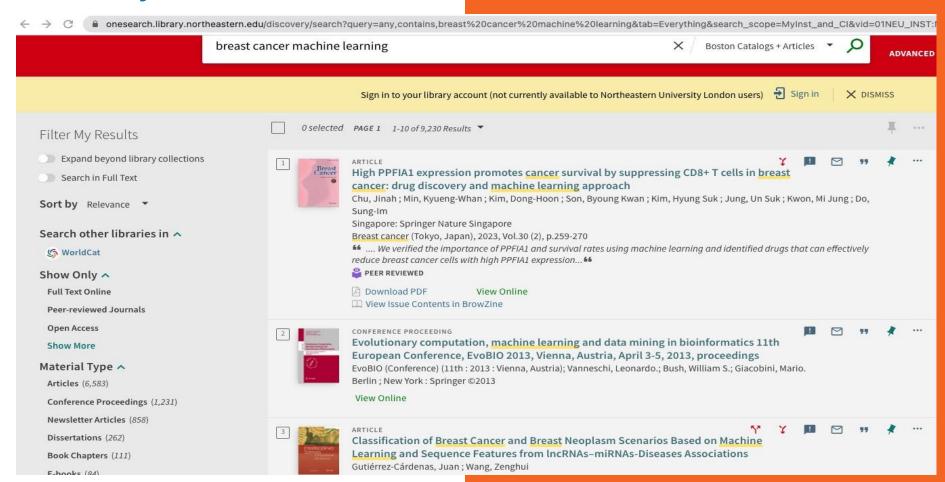
### **Currently Research Outcome:**

Machine Learning Method

Medical area



### **Currently Research Outcome:**



### **Currently Research Topic:**

**Images Analysis** 

Classification

Models for prediction or diagnosed the stages for breast cancer.

**Decision Tree for the treatments or further detection** 

Computing

**Data Visualization** 

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## . My Approach

- → Data Cleaning and Transform
- **→** Data Visualization
- → Evaluation Metric in Machine Learning



### **Choose Dataset:**

My skills and knowledge,

Currently resource,

Reasonable techniques



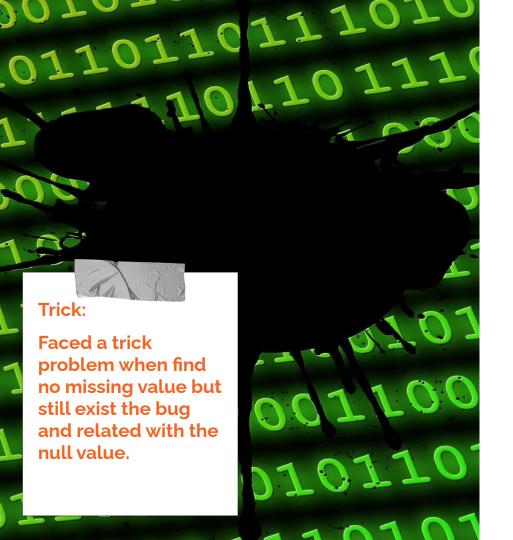
Where is my path?

Google drive?

Local device?

Cloud?

• • •



Missing data or null data:

Play around and check.

Focus.

Meet different people.

Research online.

**Learn from the practice.** 

Take a break.



Some techniques and knowledge will come after these. otherwise you will find life becomes a blur.

purehappylife.com



## **HeatMap**

```
# just learn the orginal idea for heatmap implementation, can see it is a type of data plot function.
import numpy as np
import matplotlib.pyplot as plt
# Create some random data
data = np.random.rand(5, 5)
# Create a heatmap of the data
fig. ax = plt.subplots()
    ax.imshow(data)
     a color bar
     = ax.figure.colorbar(im, ax=ax)
# Set the x and y axis tick labels
ax.set xticks(np.arange(data.shape[1]))
ax.set yticks(np.arange(data.shape[0]))
ax.set_xticklabels(['A', 'B', 'C', 'D', 'E'])
ax.set yticklabels(['1', '2', '3', '4', '5'])
# Rotate the tick labels and set their alignment
plt.setp(ax.get xticklabels(), rotation=45, ha="right",
        rotation mode="anchor")
# Loop over data dimensions and create text annotations.
for i in range(data.shape[0]):
    for j in range(data.shape[1]):
        text = ax.text(j, i, f'{data[i, j]:.2f}',
                       ha="center", va="center", color="w")
```

Display the relationship between different values in a matrix or table of data,

**Color encoding** 

Correlation vs coefficients between two variables,

Displaying the correlation between different features in a multidimensional dataset.



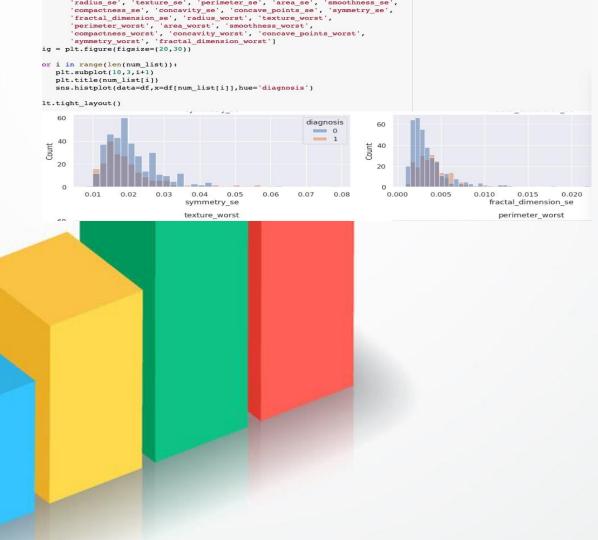
#### Tip

Stories become more credible when they use concrete details such as the specific complex moves Alberto learned through Translate and his 30 goals in 21 games performance stats.

```
# using Pandas library and Seaborn library in Python to create a heatmap of a correlation matrix.
# 1)Selects a subset of columns from a Pandas DataFrame called 'df' using the variable 'features mean'.
# 2)Computes the correlation between these columns using the Pandas corr() method and stores the resulting correlation matrix in the variable 'corr'.
# 3)creates a square figure with a size of 14 by 14 using the Seaborn plt.figure() function.
# uses the Seaborn heatmap() function to create a heatmap of the correlation matrix with the following arguments:
#cbar=True: show a colorbar next to the heatmap to indicate the range of values.
#square=True: make the cells of the heatmap square.
#annot=True: show the correlation coefficients as annotations inside the cells.
#fmt='.2f': format the annotation values as floating-point numbers with two decimal places.
#annot kws={'size': 15}: set the font size of the annotations to 15.
#xticklabels=features mean: label the x-axis ticks with the names of the selected columns.
#yticklabels=features mean: label the y-axis ticks with the names of the selected columns.
                                                                                            0.99
                                                                                                                      U.Io
                                                                                                                             U.DU
                                                                                                                                    0.09
#cmap='coolwarm': use the 'coolwarm' colormap to represent the corr
                                                                                                  U.32
                                                                                                                                           U.OZ
                                                                                                                                                 U.IO
corr = df[features mean].corr()
plt.figure(figsize=(14,14))
                                                                                                  -0.02
                                                                                                                      1.00
                                                                                                                             0.66
                                                                                                                                   0.52
                                                                                                                                          0.55
                                                                                                                                                 0.56
                                                                       smoothness mean
sns.heatmap(corr, cbar = True, square = True, annot=True, fmt= '.2
           xticklabels= features mean, yticklabels= features mean,
           cmap= 'coolwarm')
                                                                                           0.51
                                                                                                  0.24
                                                                                                         0.56
                                                                                                               0.50
                                                                                                                      0.66
                                                                                                                             1.00
                                                                                                                                                 0.60
                                                                      compactness mean
                                                                                           0.68
                                                                                                  0.30
                                                                                                         0.72
                                                                                                               0.69
                                                                                                                      0.52
                                                                                                                                   1.00
                                                                                                                                                 0.50
                                                                         concavity mean
                                                                                                  0.29
                                                                                                                      0.55
                                                                                                                                    0.92
                                                                                                                                          1.00
                                                                                                                                                 0.46
                                                                     concave points mean
                                                                                                                      0.56
                                                                                                                             0.60
                                                                                                                                   0.50
                                                                        symmetry mean
                                                                                                                                          0.46
                                                                                                                                                 1.00
```

## Histogram:

A powerful tool to see what going on in the data set



## **K-Fold Cross Validation**

Machine Learning Model Evaluation Method

## **K-Fold Cross Validation**

Detect whether the model is overfitting

Improve the accuracy of evaluation

```
| model = svm.SVC()
    classification model(model,data,prediction var,outcome var)
[ ] model = KNeighborsClassifier()
    classification model(model,data,prediction var,outcome var)
[ ] model = RandomForestClassifier(n estimators=100)
    classification model(model,data,prediction var,outcome var)
[ ] model=LogisticRegression()
    classification model(model,data,prediction var,outcome var)
data X= data[prediction var]
    data_y= data["diagnosis"]
def Classification model gridsearchCV(model, param grid, data X, data y):
        clf = GridSearchCV(model,param grid,cv=10,scoring="accuracy")
        clf.fit(train X,train y)
        print("The best parameter found on development set is :")
        print(clf.best params )
        print("the best estimator is ")
        print(clf.best estimator )
        print("The best score is ")
        print(clf.best score )
[ ] param grid = {'max features': ['auto', 'sgrt', 'log2'],
                  'min samples split': [2,3,4,5,6,7,8,9,10],
                  'min_samples_leaf':[2,3,4,5,6,7,8,9,10] }
    model= DecisionTreeClassifier()
   Classification_model_gridsearchCV(model,param_grid,data_X,data_y)
```



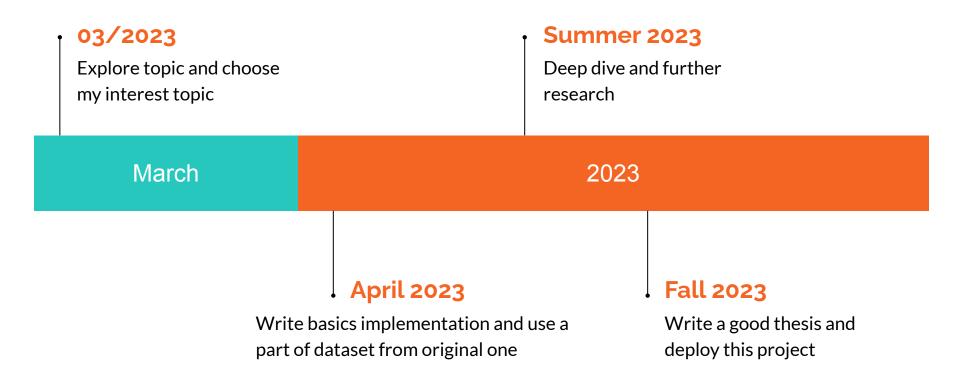
**Milestones** 

(no ending in this course)

Test and further outcome

(data distributed, data visolization, data transform, model adjust and evaluate)

## **Milestones**



## Reference for this project:

1.https://onesearch.library.northeastern.edu/discovery/fulldisplay?docid=cdi\_crossref\_primary\_10\_3991\_ijoe\_v18i05\_ 29197&context=PC&vid=01NEU\_INST:NU&lang=en&search\_scope=MyInst\_and\_Cl&adaptor=Primo%20Central&tab=Everything&query=any,contains,breast%20cancer%20machine%20learning&offset=10

2.https://www.kaggle.com/code/malik12345/breast-cancer-detection-using-ml/notebook

3.https://www.kaggle.com/datasets/nancyalaswad90/breast-cancer-dataset

4.https://www.kaggle.com/code/usakshaya/breast-cancer-prediction

5.https://machinelearningmastery.com/k-fold-cross-validation/

6.<u>https://en.wikipedia.org/wiki/Histogram</u>

7.https://towardsdatascience.com/different-ways-to-connect-google-drive-to-a-google-colab-notebook-pt-1-de034

8. https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9175124/

9.https://www.freecodecamp.org/news/how-to-handle-missing-data-in-a-dataset/

## Reference for this project:

- 10..https://www.v7labs.com/blog/overfitting
- 11.https://www.youtube.com/watch?v=qxpKCBV60U4
- 12.https://scikit-learn.org/stable/modules/cross\_validation.html
- 13.https://stackoverflow.com/questions/73443407/svm-problem-name-model-svc-is-not-defined
- 14.https://www.youtube.com/watch?v=tLhunN5Jhqs
- 15.https://www.youtube.com/watch?v=Hlk5psu5yFw

## Healthy guideline:



Leading a healthy lifestyle is recommended to protect your overall health and may help reduce your risk for certain cancers.

#### Here are a few tips to follow:











- Eat five or more servings of fruits and vegetables each day.
- Get regular physical activity.
- Maintain a healthy weight.
- · Limit alcohol intake to no more than one drink per day.
- Do not smoke. Or, quit smoking.

#### Check:

## Scheduling Exams

While living a healthy life can help reduce your risk for cancer, women can be diagnosed with breast cancer at any age. Detecting breast cancer at an early stage, when treatment is more likely to be successful, still provides the best hope for survival. This is why it is so important for you to schedule regular exams. Below you will find some general guidelines for breast cancer early detection methods. You should always consult with your doctor to create a screening schedule that is most appropriate for you.

EXAM	AGE	FREQUENCY
Breast Self-Awareness	18+	Regularly/Monthly
Well-Woman Exam	21+	Yearly
Mammogram	40+	Yearly

## Thanks!

