# **Breast Cancer Classification**

**Team Name: Prophet** 

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# Background

Depend on digital Biomedical Photography Analysis----Histopathological images

Classify images between Benign, Benign without call back and Malignant

★ Convolutional Neural Network

### Data Pre-process

#### **Data Source** Clean Data ■ 2620 mammography images Select useful data ■ Normal, benign and malignant cases Drop missing data Verified information tables Data **Split Data Process Image** ■ Rotate the images – avoid overfitting ☐ Training data: 80% ■ Validation data: 20% Resize the images Label the images

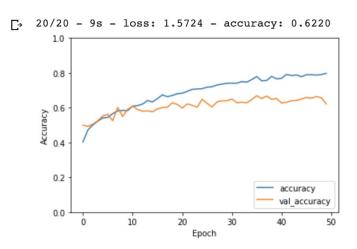
#### Model & Core Algorithm

- Model: Convolutional Neural Network
  - Automatically detecting the important features without any human supervision
- Core Algorithm :
  - Create the convolutional base
  - Add Dense layers on top
  - Compile and train model
  - Evaluate the model

```
model = models.Sequential()
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(256, 256, 3)))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Dropout(0.2))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Dropout(0.3))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.Dropout(0.3))
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu', input dim=X train.shape[1],
                       kernel_regularizer=regularizers.12(0.01)))
model.add(layers.Dropout(0.5))
# 3 classes with softmax activation for multi-class
model.add(layers.Dense(3, activation='softmax'))
model.compile(optimizer='adam',
             loss=tf.keras.losses.SparseCategoricalCrossentropy(),
```

# Result Analysis

- ☐ Training Accuracy: 79.74%
- □ Validation Accuracy: 62.20%
- ☐ Test Accuracy: 54.14%



### Former Model Analysis 1

**Problem: Overfitting** 

**Solution:** 

- Expand data volume
  - Rotate original images
- Add regularization
- Dropout features

```
plt.plot(history.history['accuracy'], label='accuracy')
 plt.plot(history.history['val accuracy'], label = 'val accuracy')
 plt.xlabel('Epoch')
 plt.ylabel('Accuracy')
 plt.ylim([0.5, 1])
 plt.legend(loc='lower right')
 test loss, test acc = model.evaluate(X test, y test, verbose=2)
10/10 - 4s - loss: 2.0941 - accuracy: 0.6355
   0.9
   0.6
                        Epoch
```

## Former Model Analysis 2

**Problem: Accuracy fluctuating** 

**Solution:** 

Increase batch size

```
[83] plt.plot(history.history['accuracy'], label='accuracy')
     plt.plot(history.history['val_accuracy'], label = 'val_accuracy')
     plt.xlabel('Epoch')
     plt.ylabel('Accuracy')
     plt.ylim([0, 1])
     plt.legend(loc='lower right')
     test_loss, test_acc = model.evaluate(X_test, y_test, verbose=2)

    ↑ 10/10 - 5s - loss: 1.7174 - accuracy: 0.6194

        1.0
        0.8
      Accuracy
9.0
        0.2
                                            accuracy
                            20
                              Epoch
```

# Further Optimization

Limited volume of training data



Find larger dataset



Add more rotated images

Poor accuracy



Create another prediction model



Add boost process

Not suitable for complex condition



Add different layers (10 scale or even more)