# Bone Fracture Detection via X-Ray Image Processing



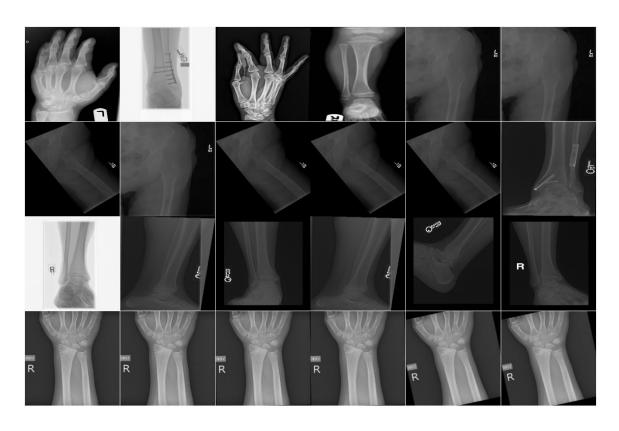
11.04.2025



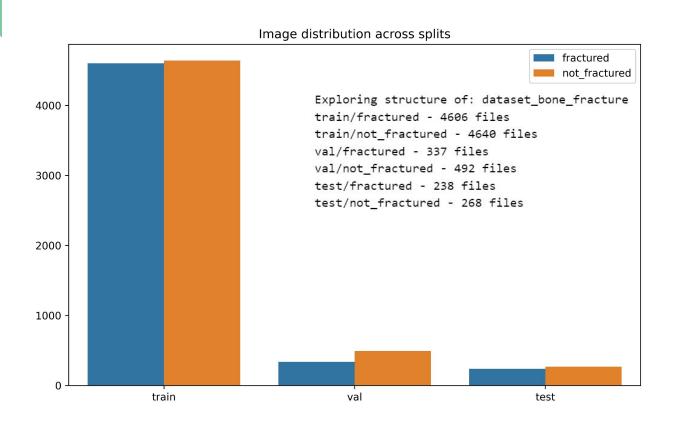
#### Overview

- **Dataset**: a collection of radiographic images (IMG\_SIZE=(224,224), in greyscale) covering all anatomical body regions (limbs, hips, knees, ...)
- Categories: train, test, and validation
- Subcategories: fractured and non-fractured
- Challenges:
  - Dataset contains corrupted images (removed via cleaning)
  - Random file names (unimportant when using ImageDataGenerator)
  - Multiple copies of the same image
  - Rotated images

# Sample Image Grid



#### Image Distribution across Splits



#### Preprocessing and Augmentation

- Constants: IMG\_SIZE=(224,224), BATCH\_SIZE=32
- Random Transformations: rotation, shift, zoom, flips
- Normalization: rescales pixel values to [0,1] for better Neural Network performance
- **ImageDataGenerator** is applied for training, validation, and testing a binary classification task
  - shuffle=False for test data ensures prediction
  - Class labels are inferred from subfolder names (frac, non-frac)

#### **B**uild a Model

- Architecture: sequential
- Convolutional Layers: 32 filters, ReLU activation, 3x3 Kernel size,
   224x224x3 input shape
- **Pooling Layers:** MaxPooling(2,2) downsamples feature maps
- **Flatenning**: to create 1D dense layers
- **Dropout (0.5)** drops 50% of neurons during training
- Dense sets up fully connected layers
- Final Layer: I neuron (binary classification) and sigmoid activation

## Model Summary

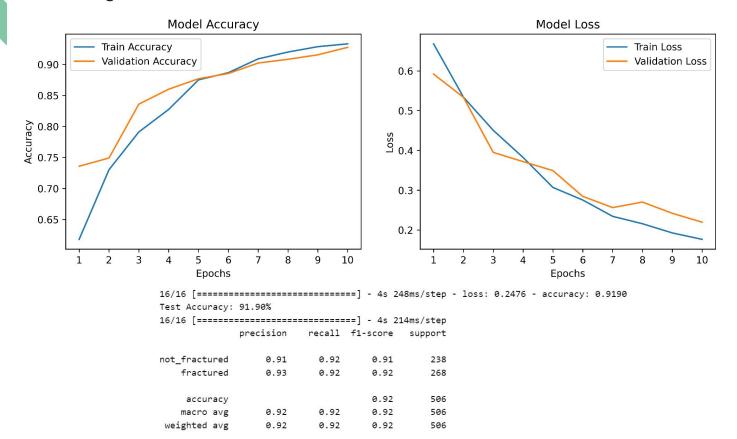
Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 222, 222, 32)	896
<pre>max_pooling2d (MaxPooling2 D)</pre>	(None, 111, 111, 32)	0
conv2d_1 (Conv2D)	(None, 109, 109, 64)	18496
<pre>max_pooling2d_1 (MaxPoolin g2D)</pre>	(None, 54, 54, 64)	0
flatten (Flatten)	(None, 186624)	0
dropout (Dropout)	(None, 186624)	0
dense (Dense)	(None, 128)	23888000
dense_1 (Dense)	(None, 1)	129

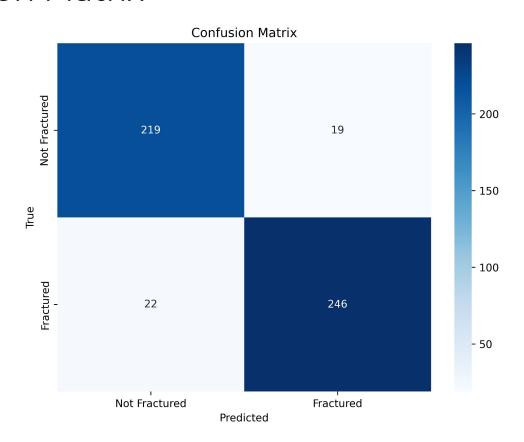
### Training Phase

```
Epoch 2/10
Epoch 3/10
Epoch 4/10
Epoch 5/10
Epoch 6/10
Epoch 7/10
Epoch 8/10
Epoch 9/10
Epoch 10/10
```

#### Accuracy and Loss



### Confusion Matrix



#### Further Improvement

#### Use of Hashing to remove duplicates

Avoids data leakage, class imbalance, and wasted computation

#### Use of Transfer Learning

 MobileNetV2 is lightweight and efficient using a pre-trained model for better performance of medical imaging

#### Rotation Invariance Strategies

- Add rotation invariance strategies or augmentation tricks for rotated duplicates
- **Wise choice of filename** for better reproducibility, debugging, and interpretation

# Thank you!