

UNIVERSITY OF  
CAMBRIDGE

# Bone Doctor

Team Bravo

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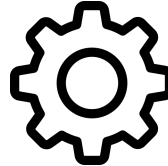
# Problem background

- Assist clinicians with analysing X-rays in hospitals
- Often untrained staff



X-ray

Patient comes  
X-ray is taken



Bone doctor

Produce information  
useful for the clinician



Review by a clinician

Decide on next steps

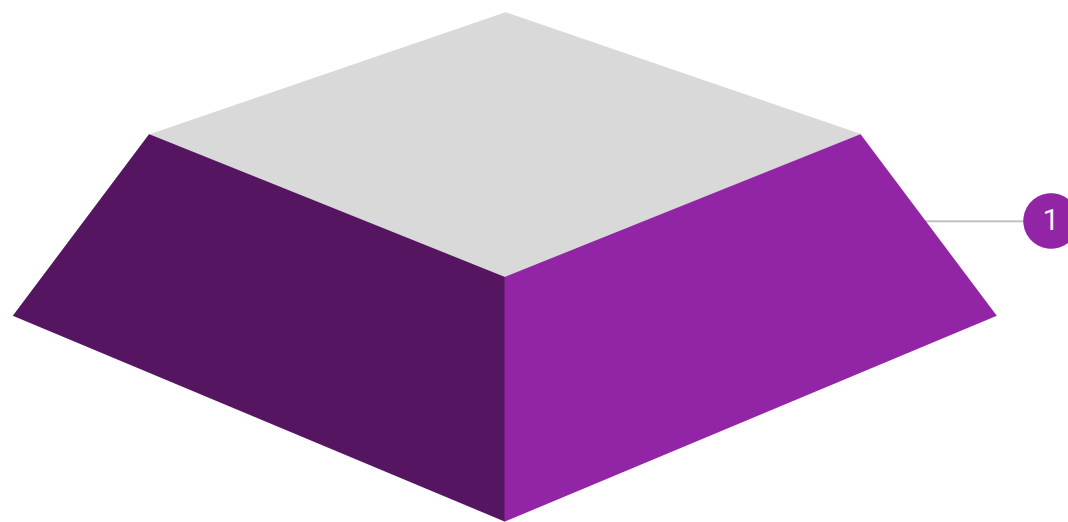
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# Features

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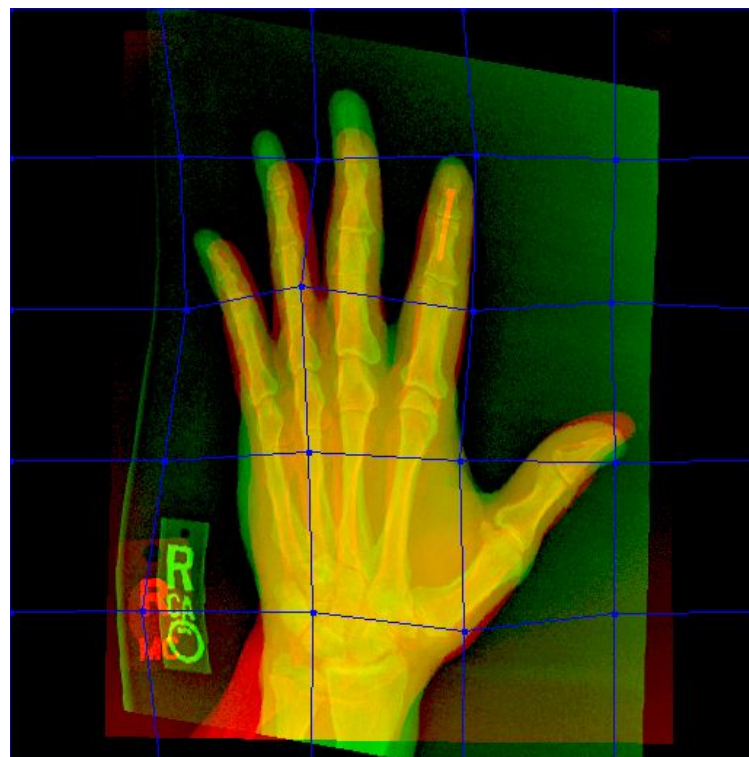
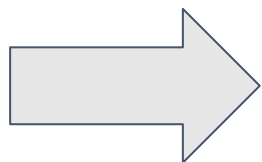
# Features



## **Refine the X-ray**

Make the X-ray more  
readable for the clinician

# Results



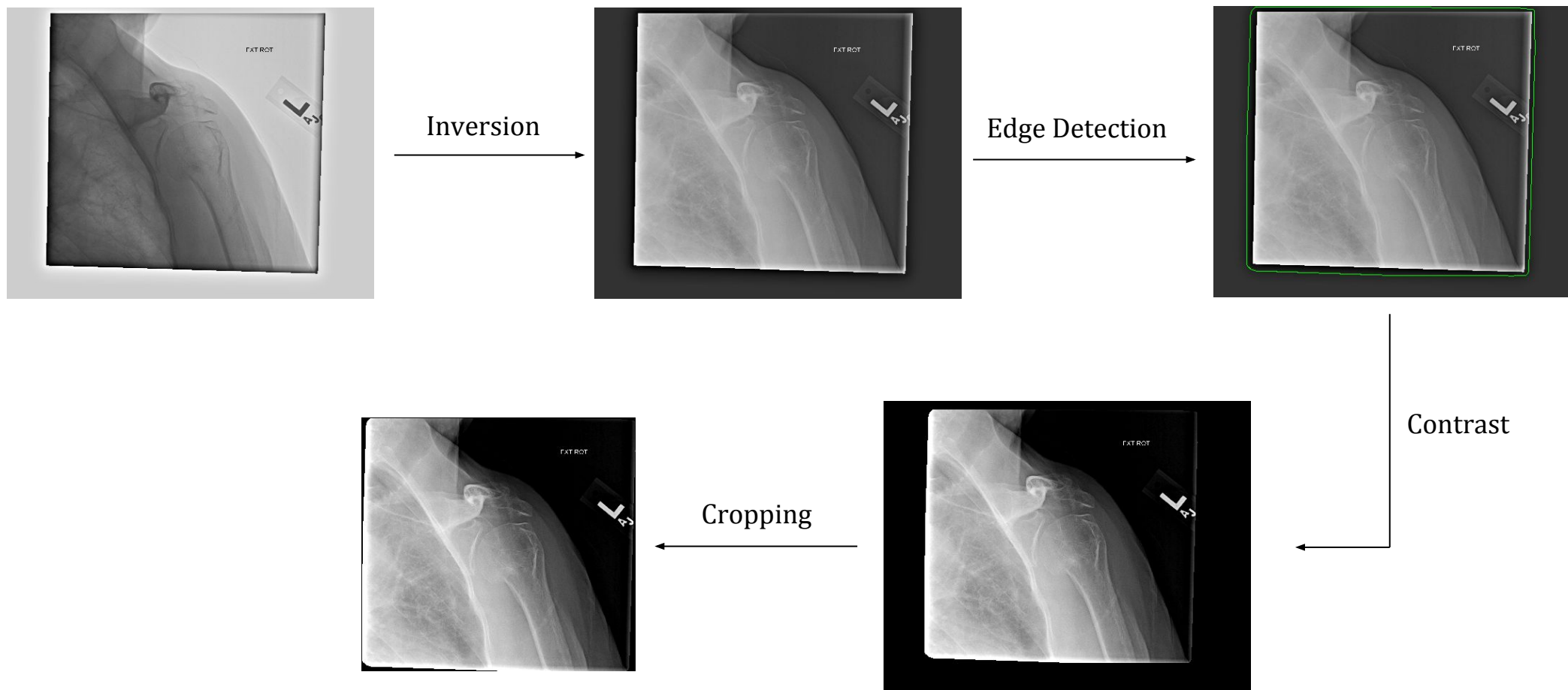
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# The MURA Dataset

- Classified body parts
- Classified normal / abnormal

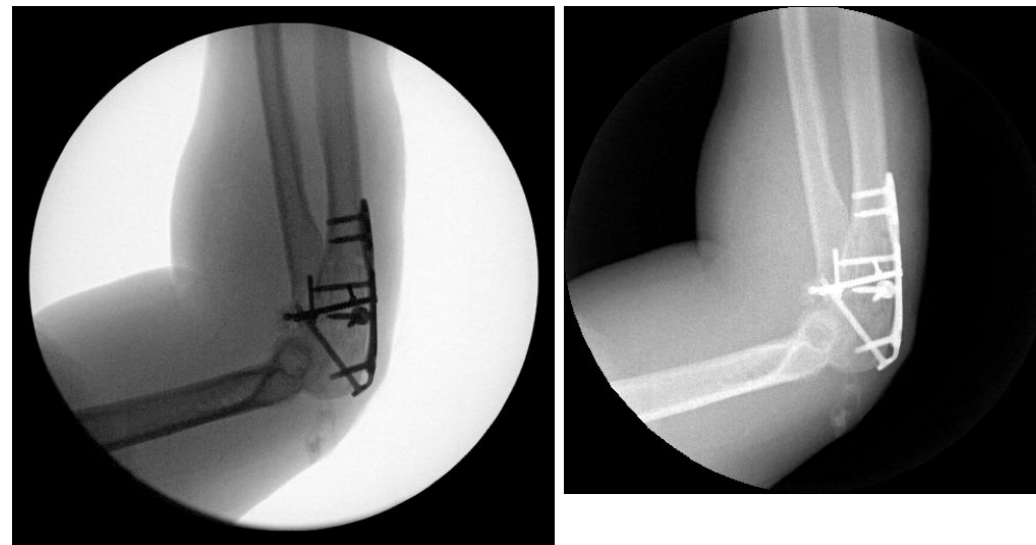
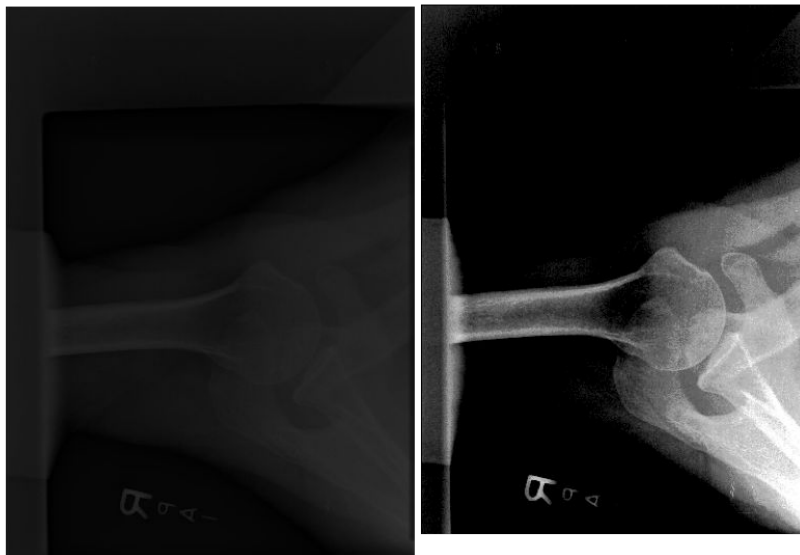


# Refining the X-ray



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## Refining the X-ray



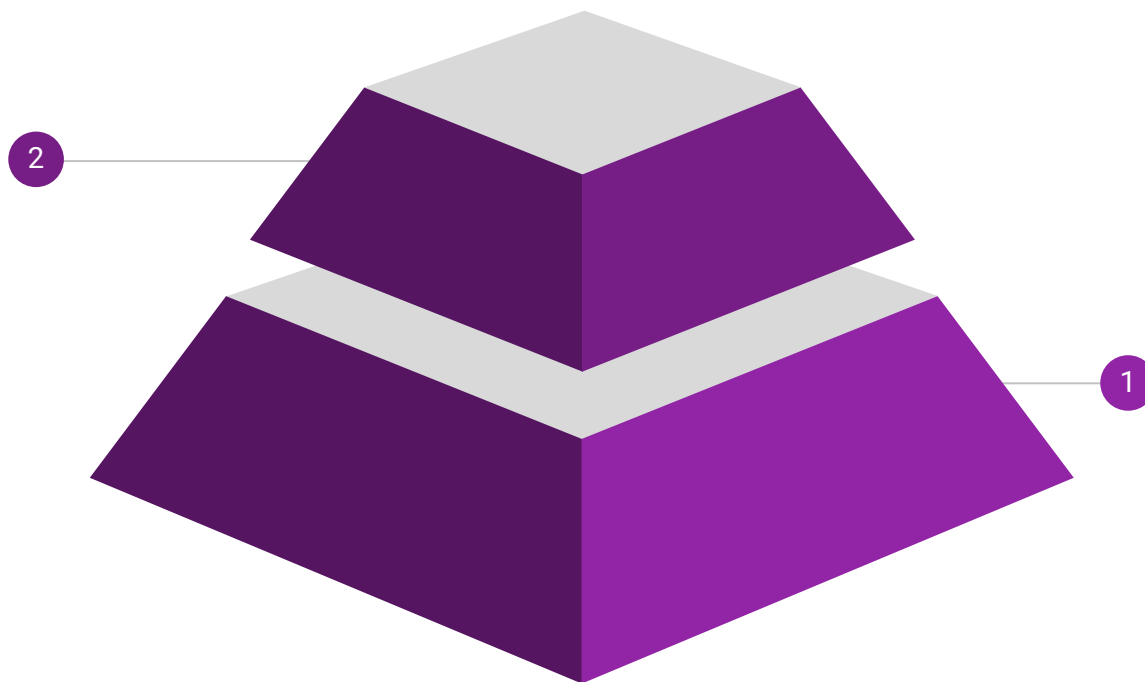


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# Features

## Find a fracture

Classify normal / abnormal,  
highlight abnormalities



## Refine the X-ray

Make the X-ray more  
readable for the clinician

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# Classification of Normal/Abnormal

- A binary classification model to distinguish between normal and abnormal X-ray images.



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# Difference Highlighting

- We use our matched image in order to highlight abnormalities in the original image.
- In order to do this, we found a way to encode each pixel in a training image into a hashset so that the highlighter only has to do a lookup to determine if a pixel is abnormal.

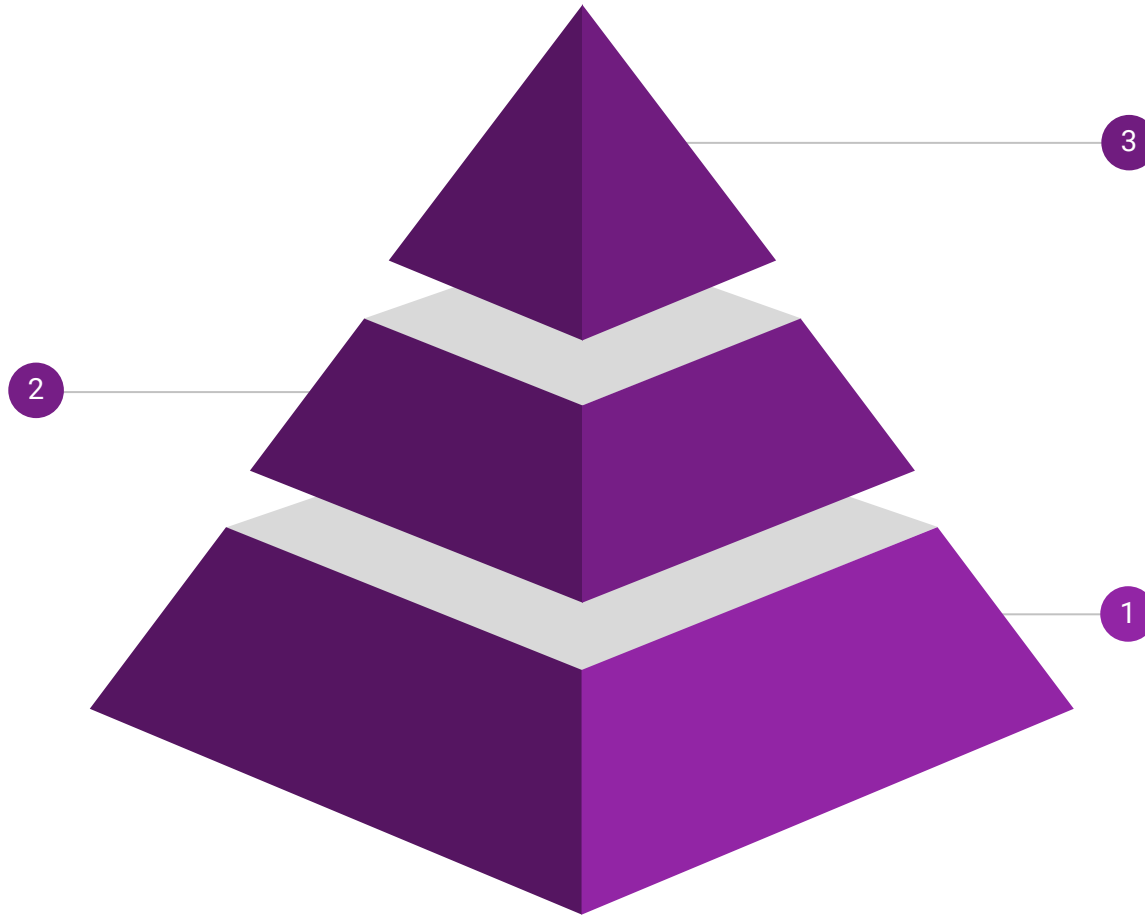


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# Features

## Find a fracture

Classify normal / abnormal,  
highlight abnormalities



## Compare with similar cases

Use X-rays with similar views  
of the same body part  
to compare differences

## Refine the X-ray

Make the X-ray more  
readable for the clinician

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# Clustering of Images

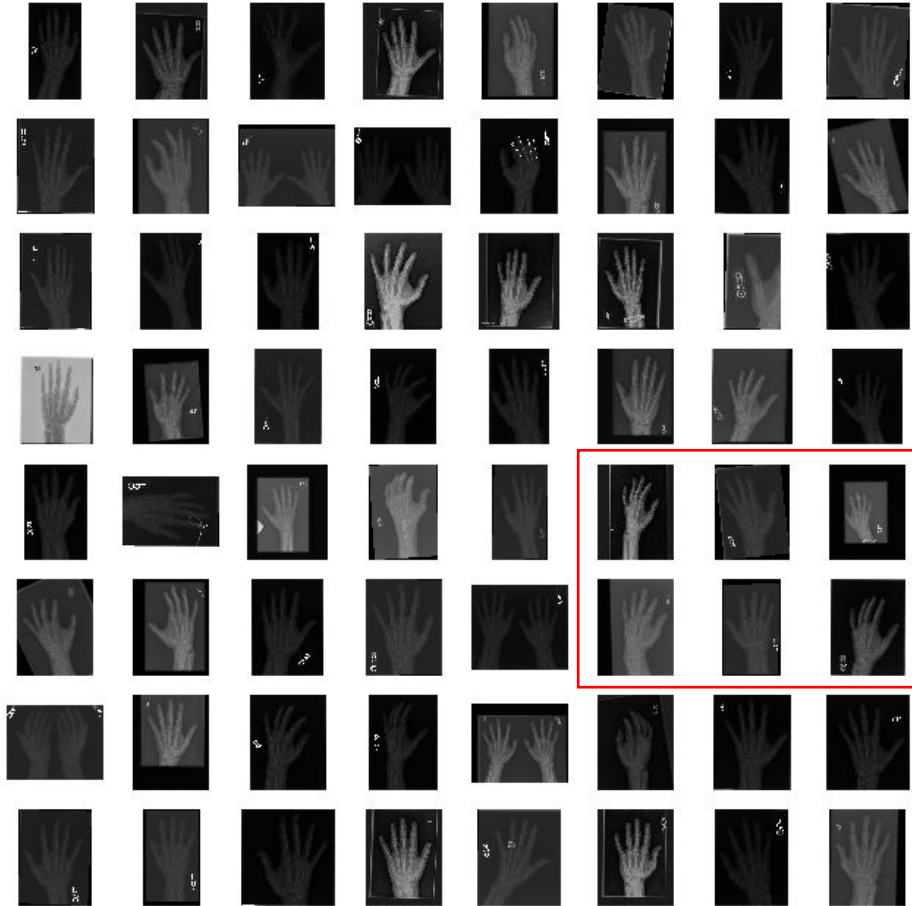
- Different views of each bone



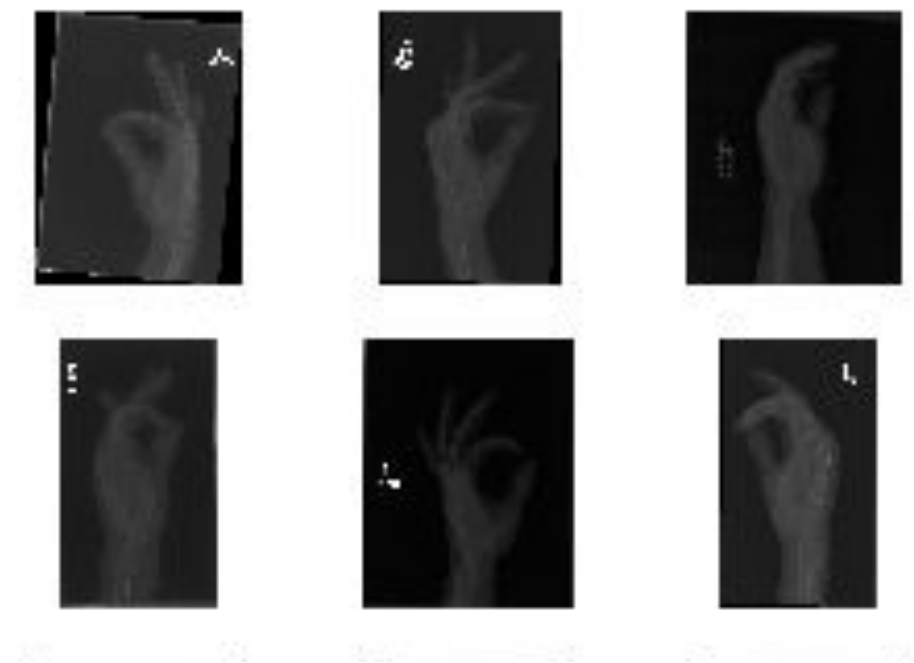
- Unsupervised learning
  - X-rays in dataset  $\rightarrow$  feature vectors
  - Input X-rays are assigned to pre-computed clusters
  - Cluster quality evaluated by silhouette score ( $\sim +0.1$ )
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# Clustering of Images



# Clustering of Images



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# Searching for Matching Images

- Using image hashing to locate the most similar subset of images.
- Given an input image, choose the most similar image from the subset by computing the cosine similarity of the images' feature vectors.
- Higher cosine similarity = greater visual similarity

Input Image



High Score



Low Score

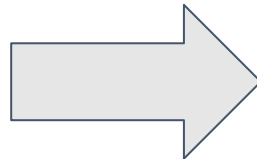




# Image hashing

- 40,000 images
- Searching all would be impossible in real time
- We instead use a hash function in order to compress the images
- A perceptual hash is one where visually similar images hash to the same thing or almost the same thing

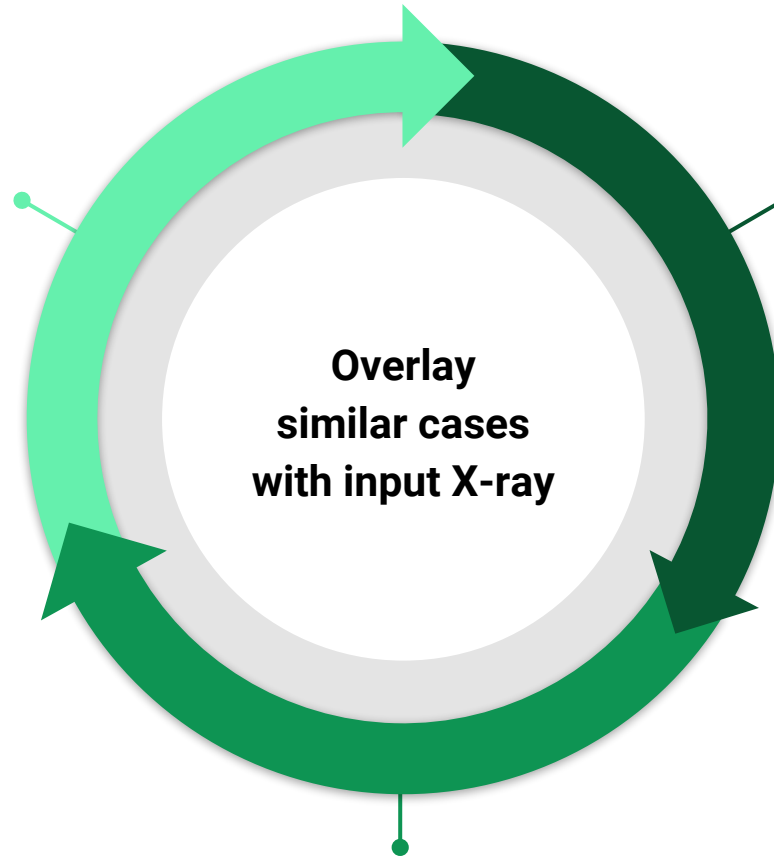
# Found similar images



# Overlaying of Images



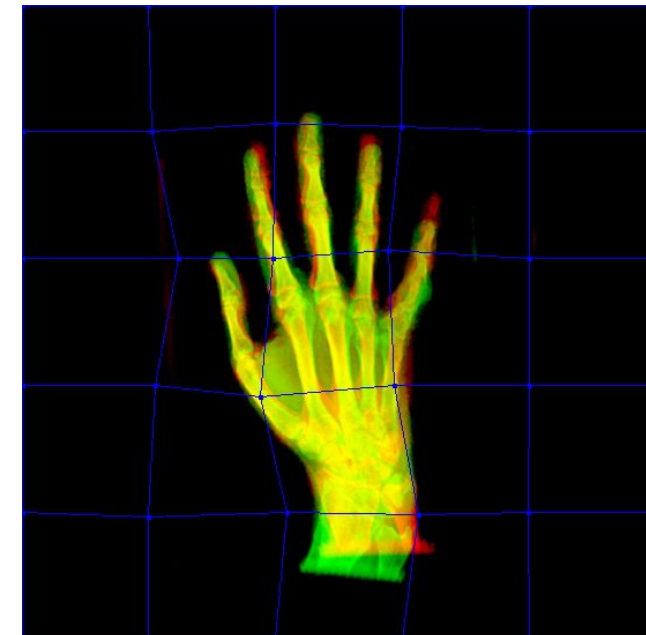
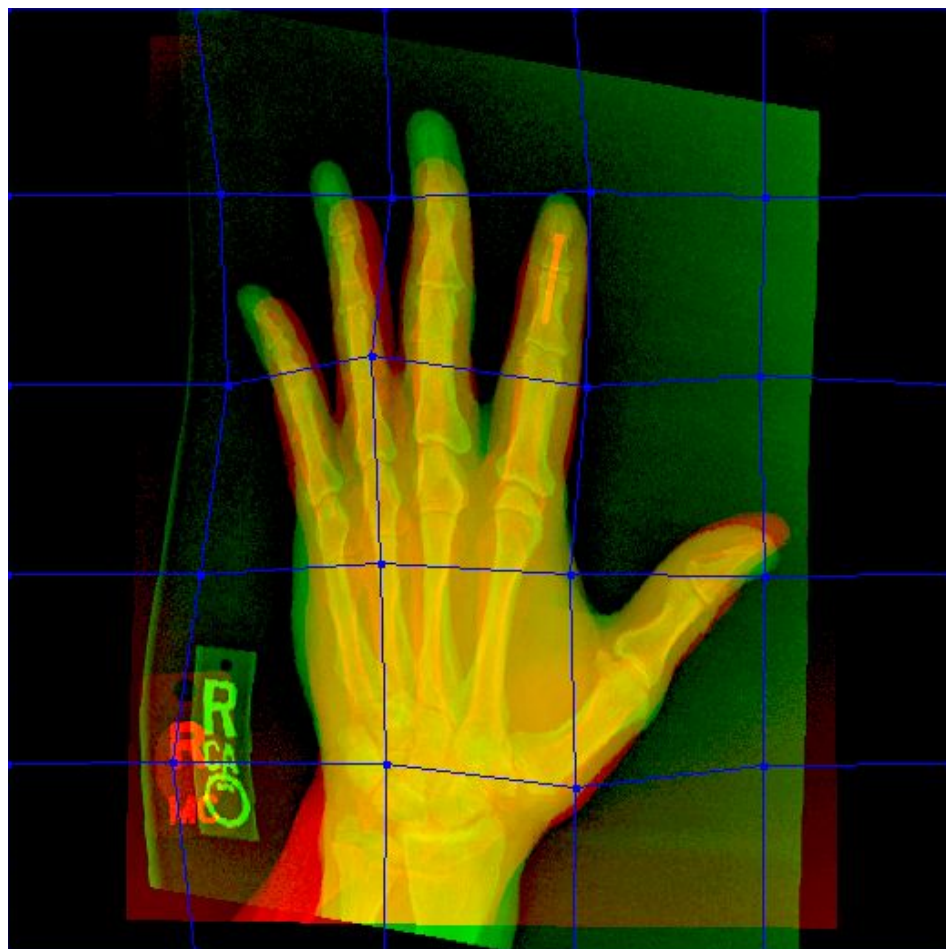
**Minimise**  
using the BOBYQA  
optimising algorithm



**Transform**  
Affine and Warp transforms

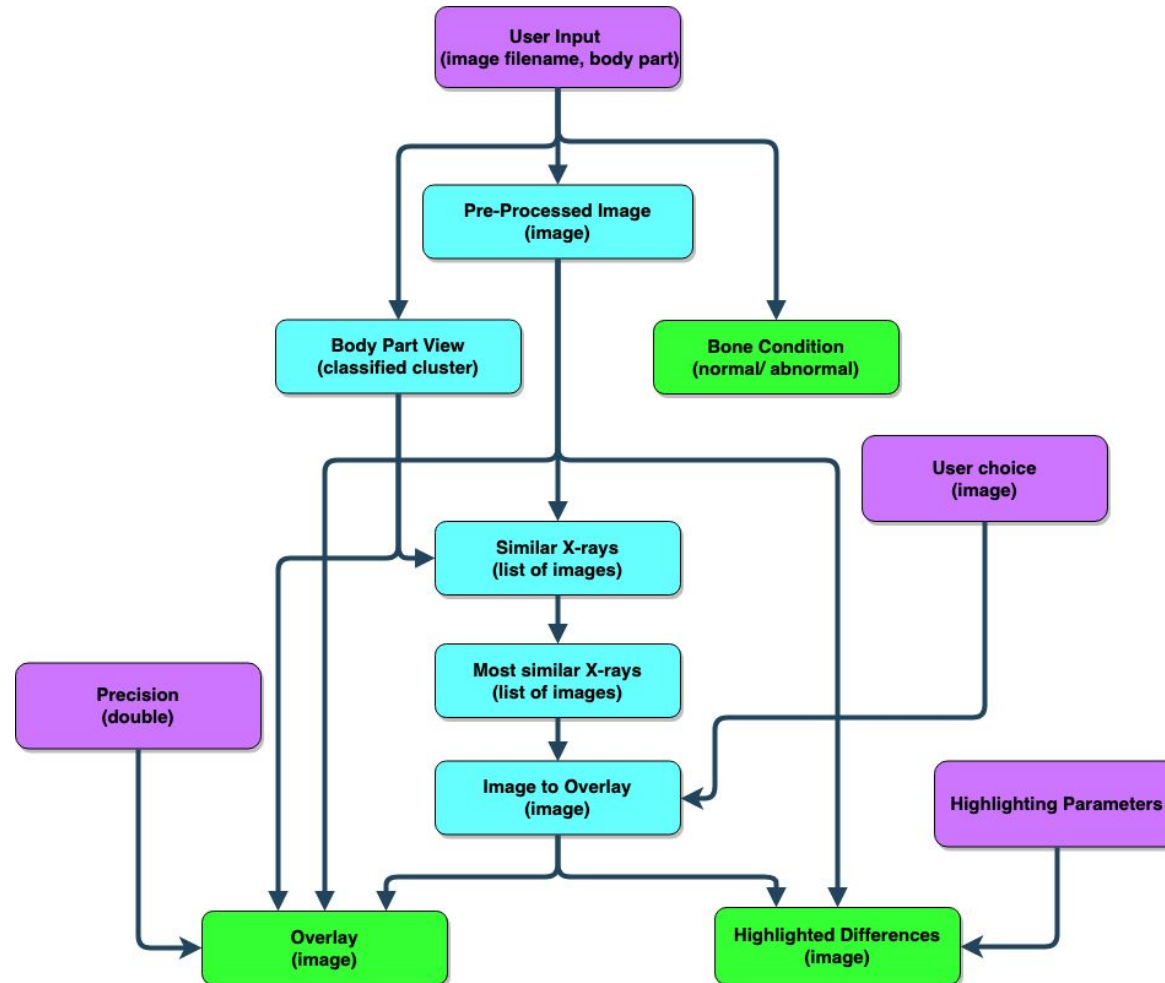
**Evaluate**  
Calculate difference  
between images

# Results

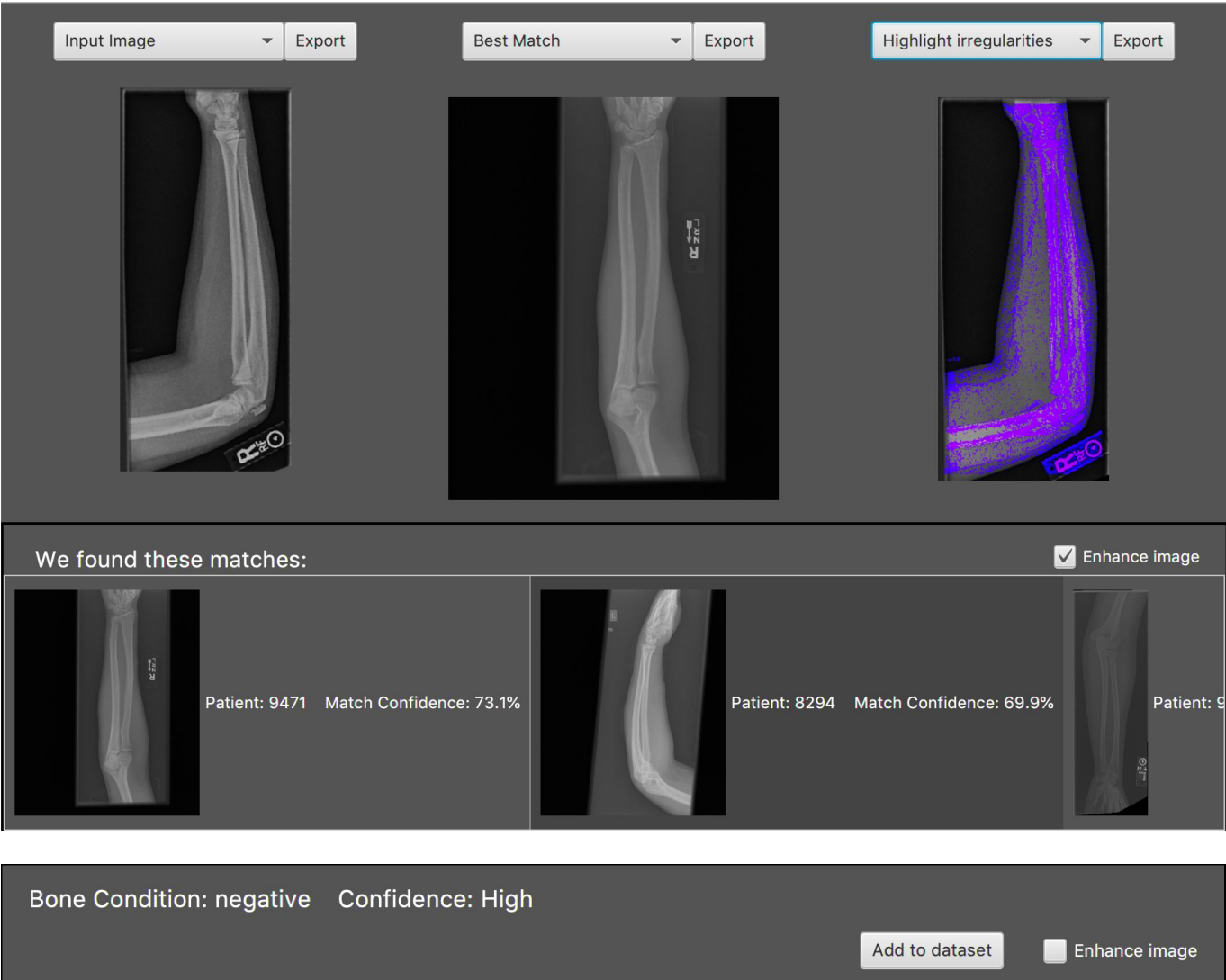
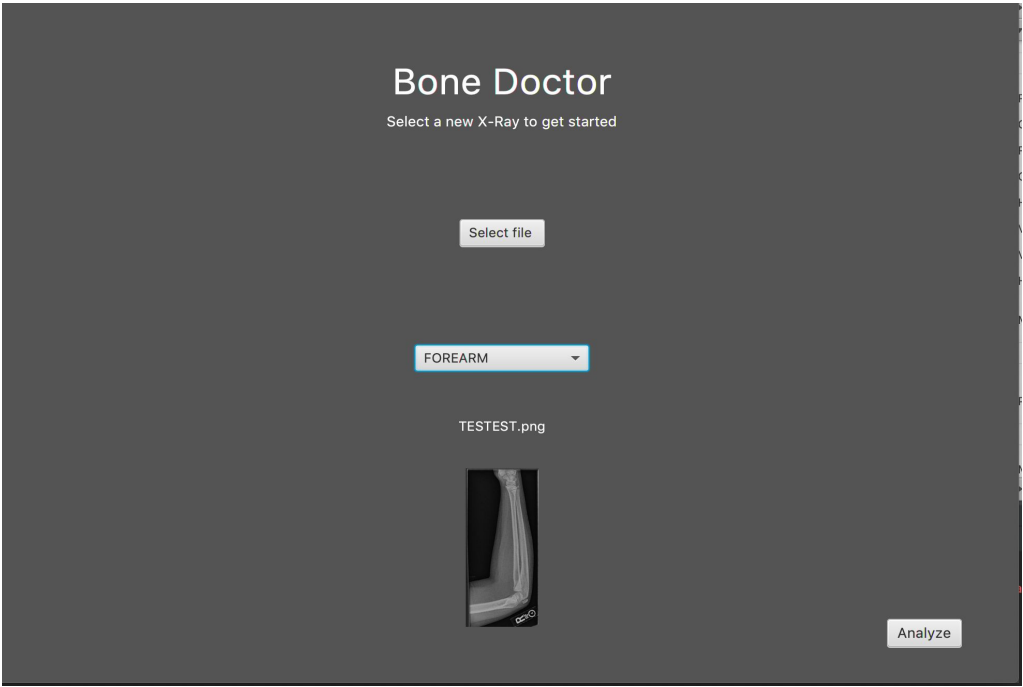


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# Connecting the components



# The User Interface





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# Bone Doctor

... taking healthcare three steps further

## Team Bravo

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The MURA dataset is released under [Stanford University School of Medicine MURA Dataset Research Use Agreement](#)

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