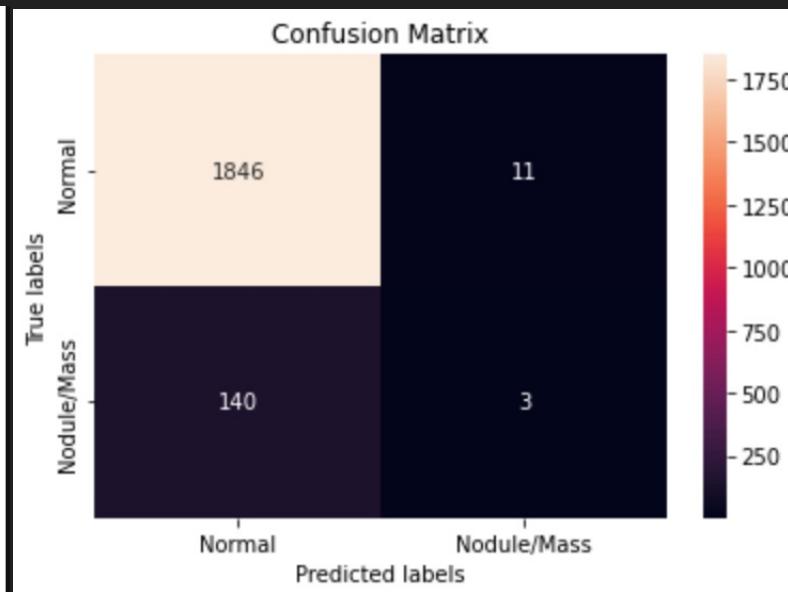
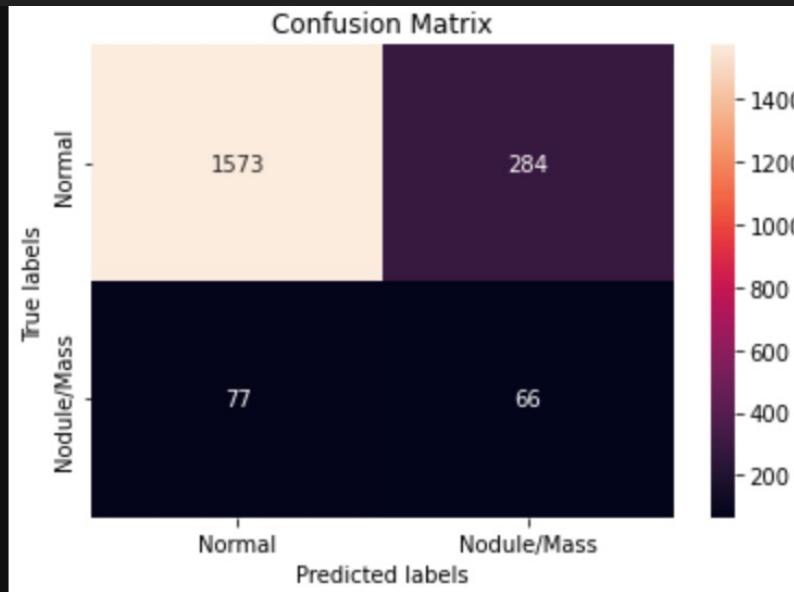
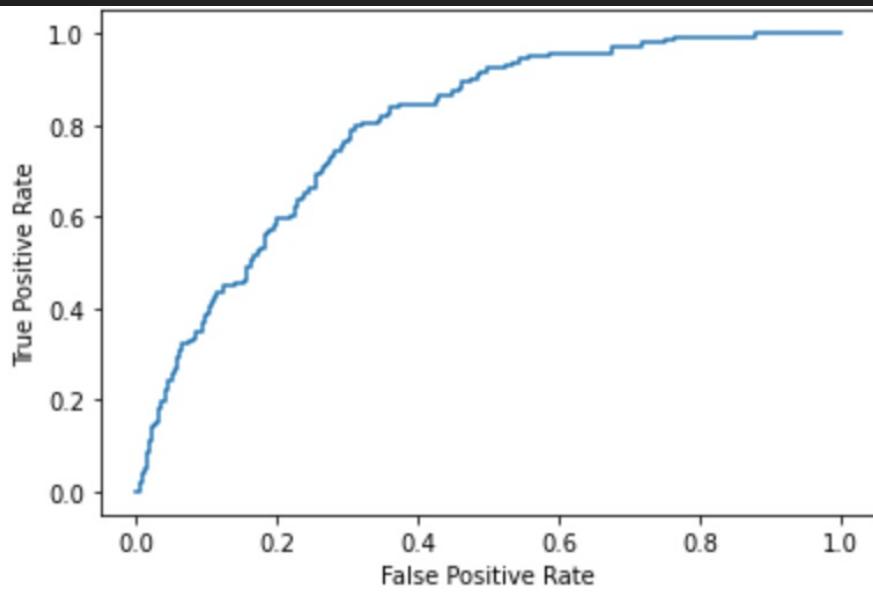


Behold AI Task



Initial Model & Analysis

Recall = 0.46

Precision = 0.19

F1 score = 0.27

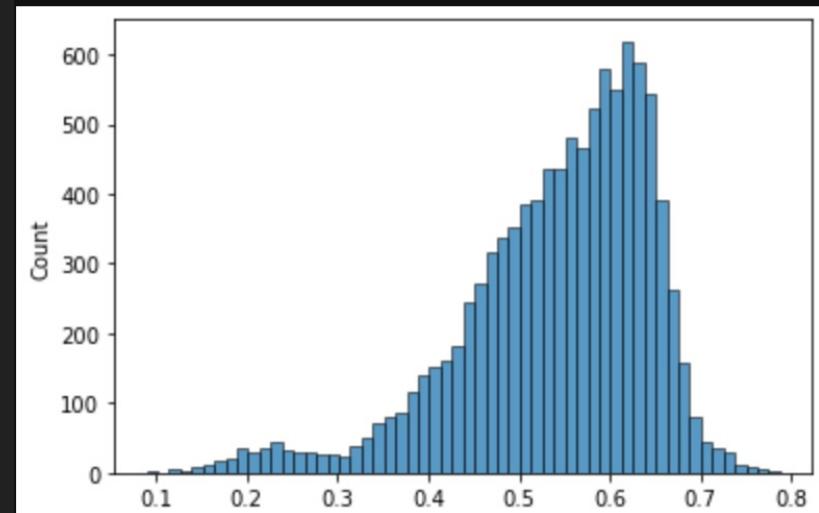
AUC = 0.795

Confusion matrix at 0.1 & 0.5 threshold

Data Augmentation

- Brightness variation:
 - Analysed mean brightness variance in training dataset
 - Used standard deviation (std) of brightness
 - Used torchvision class to randomly vary brightness within range
[$\max(0, 1 - \text{mean brightness std})$, $1 + \text{brightness std}$]

```
transforms.ColorJitter(brightness=img_brt_std),
```



- Horizontal Flipping
 - Random horizontal flipping ($p=0.5$)

Loss Function & Optimiser

- ML Problem: Imbalanced dataset.
- Business problem: False Positive preferable to False Negative
- Reweighted Binary Cross Entropy to optimize for recall (of nodule/mass class)
 - Positive weight = ratio of positive examples to negative examples

```
self.BCELoss = nn.BCEWithLogitsLoss(pos_weight=self.pos_weight)
```
- Experimented with SGD w/ momentum vs Adam

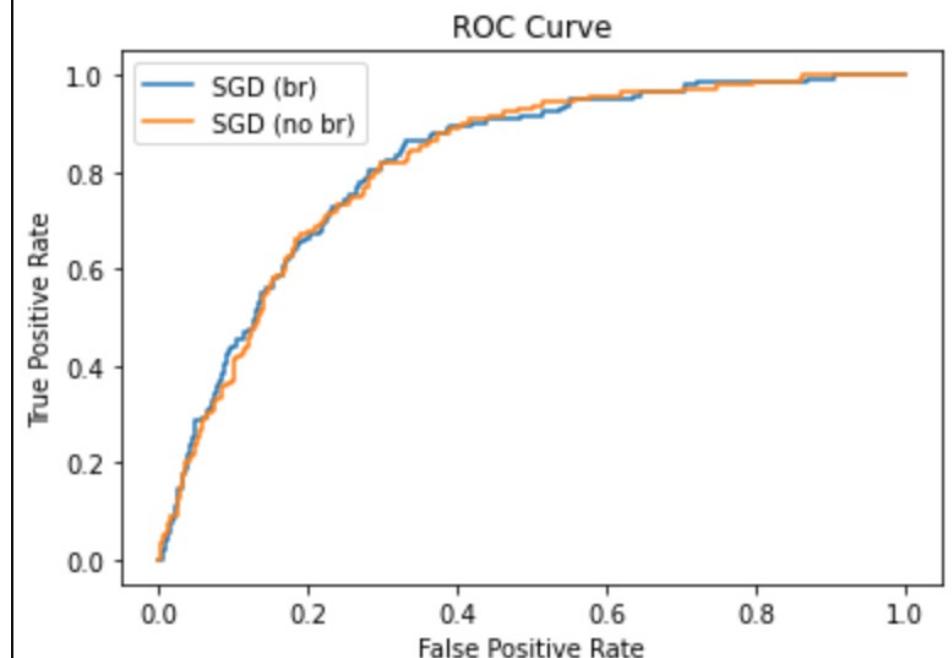
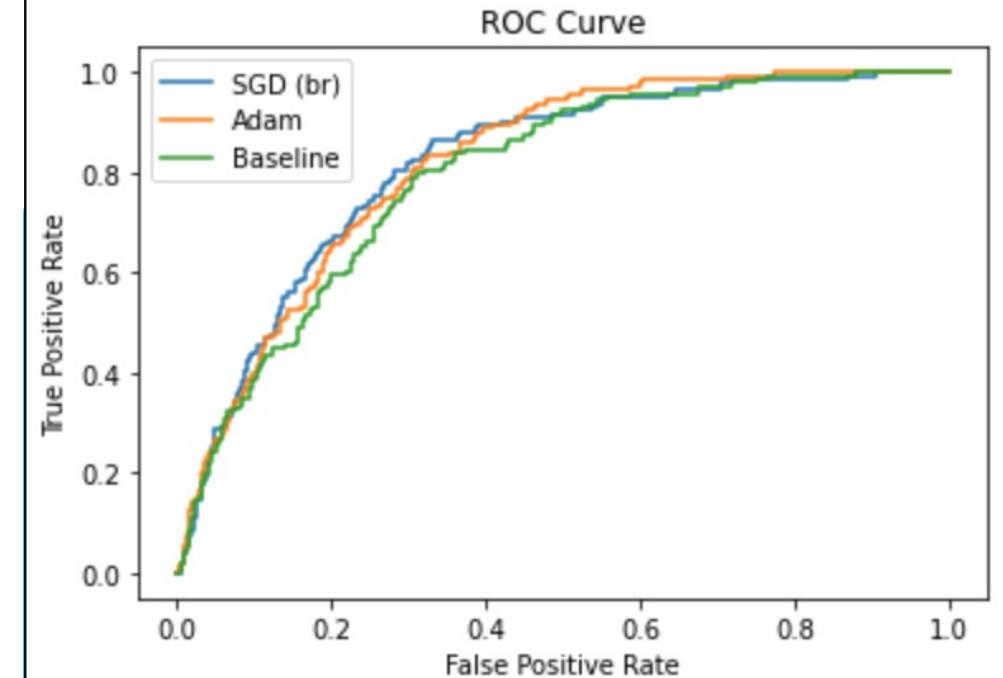
Results

Adam vs SGD

- Trained both models for 20 epochs.
- Adam AUC (brightness) = AUC 0.818
- SGD (no brightness) AUC = 0.815
- SGD (brightness) = 0.821

Brightness Variation

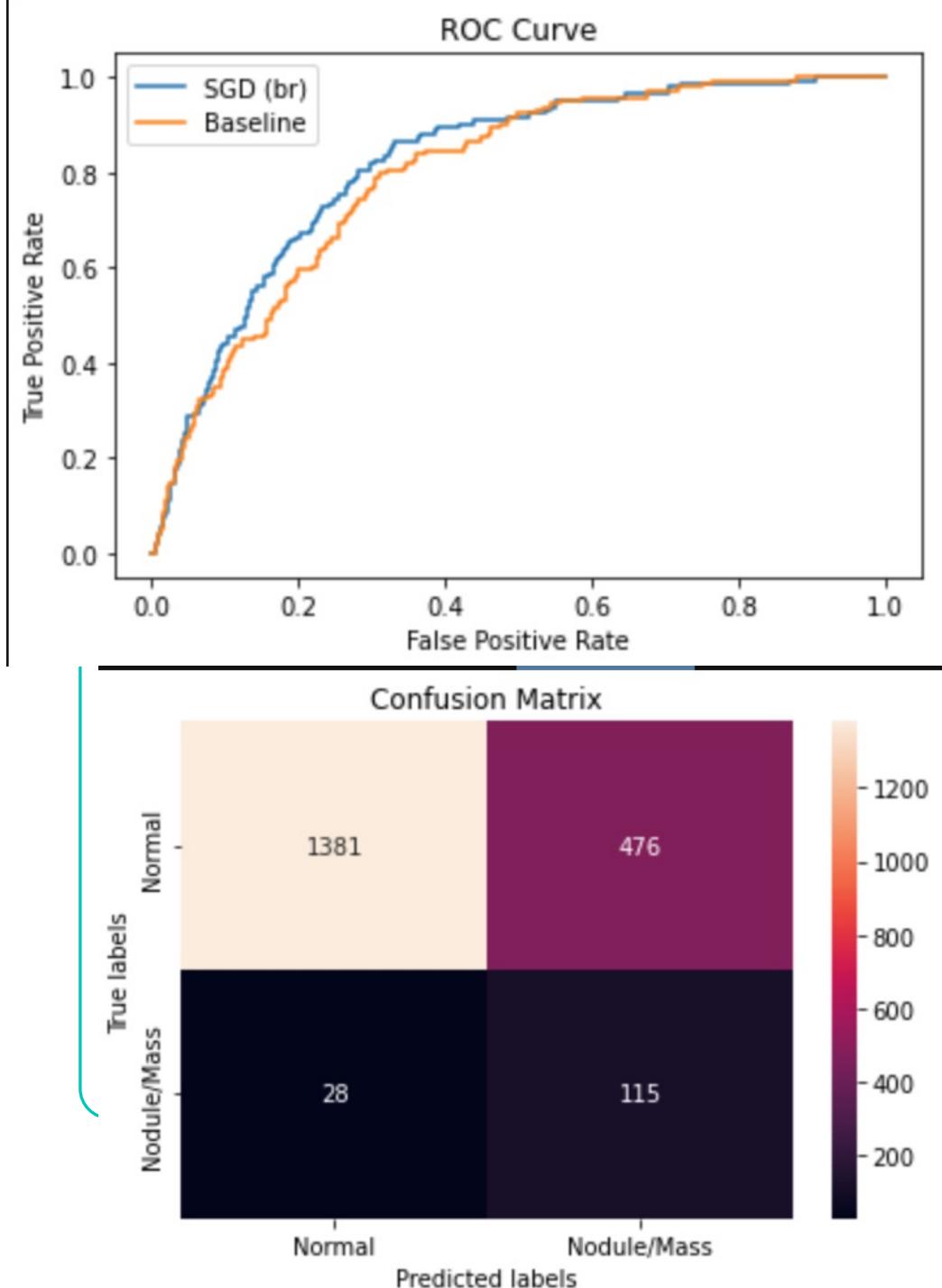
- Brightness Variation of 1 std slightly improves model performance
- Doesn't have significant impact



Results

Best model

- Chose SGD w/ Brightness
- Because of the weighting to positive examples, the ROC curve has a higher gradient initially.
- Similar AUC overall to Adam.
- But better recall at lower thresholds. Given consideration of FP cost vs FN cost, I believe this makes it preferable.
- Threshold tuned to 0.16.
 - Nodule/Mass Recall = 0.80
 - Nodule/Mass Precision = 0.2
 - Nodule Mass F Score = 0.61 (beta = 3)
 - Maximised F-score for beta = 3



What to label?

- Prioritise labelling of images closest to decision boundary (i.e. that the model is most uncertain about).
- From the chosen model, the decision boundary/threshold was empirically chosen as 0.16.
- Choose the 300 assets closest to the decision boundary.