PART C:

Preface:

First we need to generate the ROI/Thumbnail using the *generate _roi(input img, salmap, thresh, alpha, beta)* function.

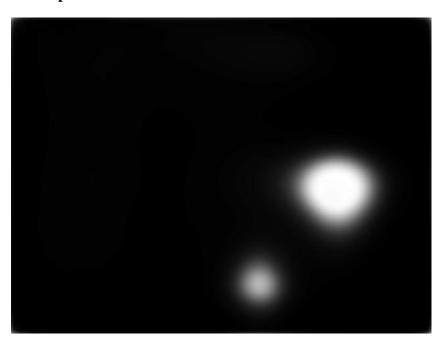
```
def generate_roi(salmap, thresh, alpha, beta):
  image=salmap
  # height, width, number of channels in image
  height = image.shape[0]
  width = image.shape[1]
  aspect=width/height
  threshMap = cv2.threshold(image, int(2.55*thresh), 255,cv2.THRESH_BINARY)[1]
  # find contours and get the external one
  contours, hier = cv2.findContours(threshMap, cv2.RETR_TREE, cv2.CHAIN_APPROX_SIMPLE)
  for c in contours:
    # get the bounding rect
    x, y, w, h = cv2.boundingRect(c)
    if w>15 and h>15:
       roi.append([x,\,y,\,w,\,h])
       # cv2.rectangle(image_orig, (x, y), (x+w, y+h), (0, 0, 255), 2)
  # print(roi)
  for r in roi:
    h_opti=round((r[2]/aspect))
     w_opti=round(r[2])
    # draw a red rectangle to visualize the bounding rect
    cv2.rectangle(image\_orig, (x, y), (x+w, y+h), (0, 0, 255), 2)
    y1 = int(r[1]/4)
    Rs=0
    roi_final=[]
    for j in range(r[3]-h_opti):
       dr=alpha*np.sum(threshMap[r[1]+j:r[1]+j+h\_opti, r[0]:r[0]+r[3]])-beta*(h\_opti*r[3])
         roi\_final = [r[0], r[1] + j, r[3], h\_opti]
  return roi_final
```

Examle:

Input Image:



SalMap:



Calculation of Aspect Ratio:

height = image.shape[0]
width = image.shape[1]
aspect=width/height

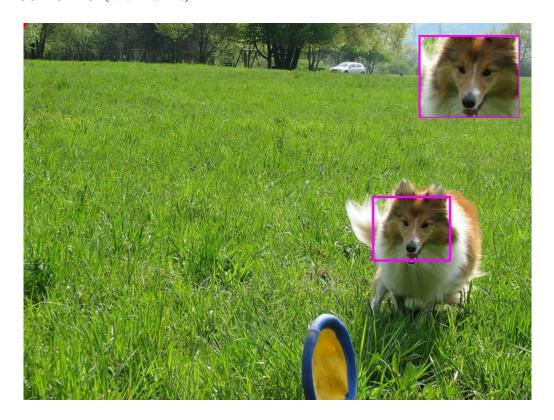
ROI Score:

$$R_s = \alpha \left(\sum_{P \in \mathbf{R}} S_P P \right) - \beta(N)$$

equivalent code:

```
\label{eq:resonant_resonant_resonant_resonant} Rs = 0 \\ \mbox{roi\_final} = [] \\ \mbox{for } j \mbox{ in } range(r[3]-h\_opti): \\ \mbox{dr=alpha*np.sum(threshMap[r[1]+j:r[1]+j+h\_opti, r[0]:r[0]+r[3]])-beta*(h\_opti*r[3])} \\ \mbox{if } dr > Rs: \\ \mbox{Rs=dr} \\ \mbox{roi\_final} = [r[0],r[1]+j,r[3],h\_opti] \\ \mbox{}
```

OUTPUT ROI (With Max Rs)



PARAMETERS:

 $-\alpha = 0.8, \, \beta = 0.3$

 $-\alpha = 1.0, \beta = 0.1$

 $-\alpha = 0.6, \beta = 0.4$

Precision/Recall Equations:

precision=TP/(TP+FN)
recall=TP/(TP+FP)

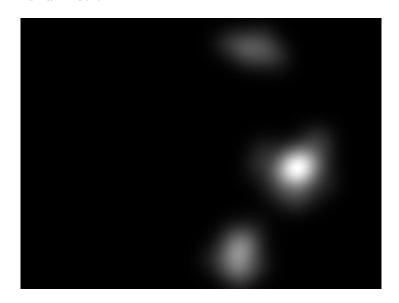
where:

TP=True Positive FN=False Negative FP=False Positive

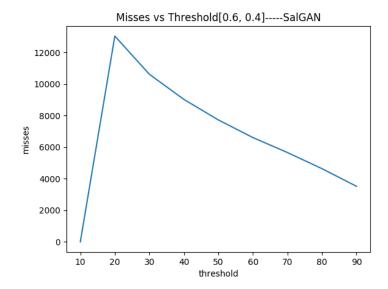
Corresponding Score Move Code:

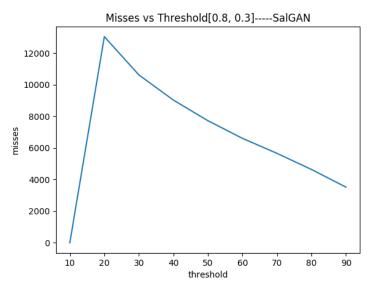
```
area = cv2.contourArea(c)
  temp = mask[r[1] : r[1] + r[3], r[0] : r[0] + r[2]]
  FP=abs(roi[2]-r[2]) * abs(roi[3]-r[3])
  precision=TP/(TP+FN)
  recall=TP/(TP+FP)
  return precision,recall
def salience_score_roi(map, roi):
temp=map[roi[1]:roi[1]+roi[3],roi[0]:roi[0]+roi[2]]
  score=np.mean(temp)
  tot=np.sum(temp>0)
  missed=tot-score
  return score, missed
for j in range(3):
  sc=[]
  ms=[]
  th=[]
  for i in range(10,100,10):
     try:
       roi_coords = generate_roi(image_sal, i, param[j][0], param[j][1])
        score,missed = salience_score_roi(image_mask, roi_coords)
       sc.append(score)
       ms.append(missed)
        th.append(i) \\
     except:
       pass
```

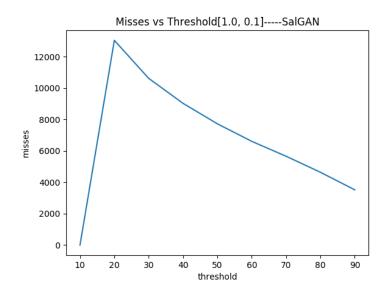
Human Truth:

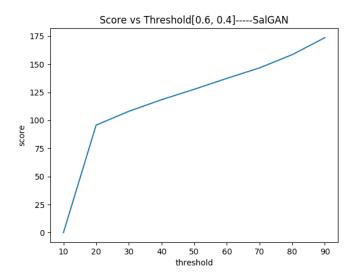


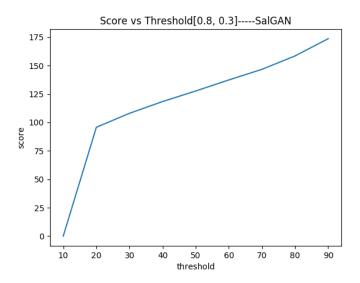
PLOTS (SalGAN):

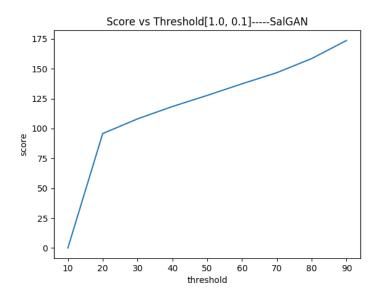




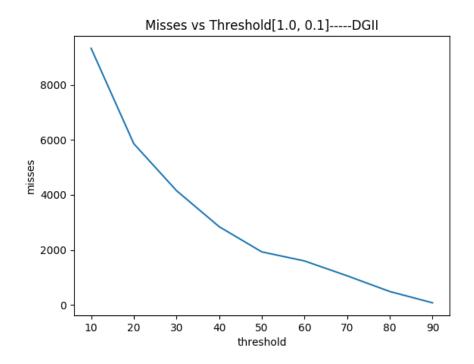


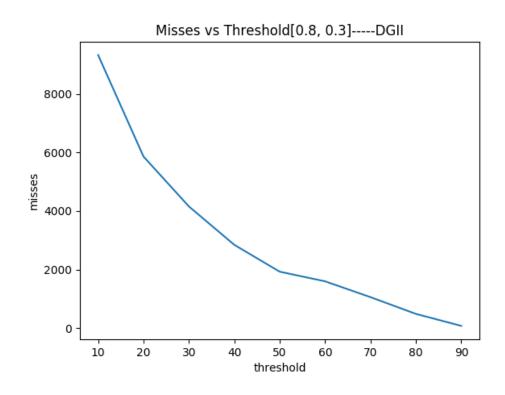


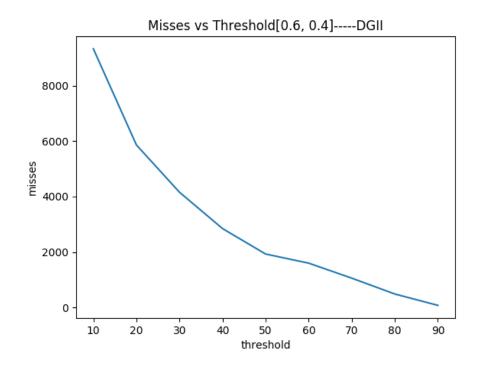


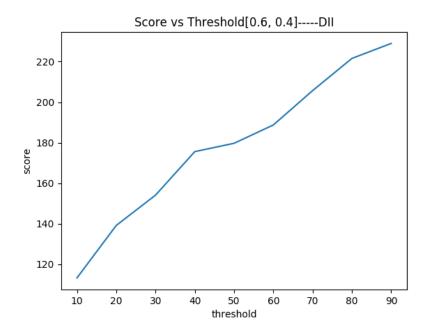


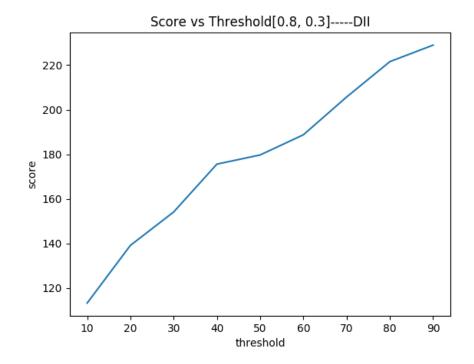
PLOTS (DGII):

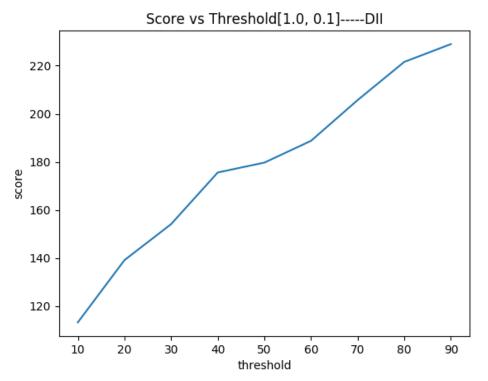












Discussio n:

For **SalGAN** and DGII, obviously the SalGan has better AUC than DGII as evidentfr om the plots. However, for the other salincy maps,

there no visual or numerical conclusion from plots, hence not showed.

The score to threshold maps shows that, for the the DGII odel, the plot is almost linear, however for salgan, the plot is almot exponential. It can thus be concluded that for salgan, the increase in threshold has more incremental effect on score than DGII. I would prefer SalGAN for its better AUC and Scor/Threshold values.