# Generative Adversarial Networks (GAN) Saliency Detection, Depth Based Saliency Comparison

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## What is saliency prediction?













(a) RGB

(b) Depth

(c) Salient

(a) RGB

(b) Depth

(c) Salient

Figure: Example saliency from the NLPR dataset

Figure: Example saliency from the Olesova dataset

Saliency prediction algorithms identify eyecatching objects in a scene/quantify how salient each pixel of a scene is.

#### What is a GAN

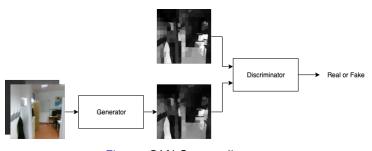


Figure: GAN System diagram

GANs are a design of neural network based on two networks:

- Generator To create fake images based off some input
- Discriminator To classify real and fake images

#### Discriminator

A discriminator is based on a convolutional neural network to classify images.

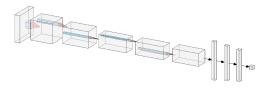


Figure: Convolutional neural network

A convolutional neural network is based on image convolution to downsample an image into a more processable amount of data. with each convolution more feature maps are generated. Once the images are small enough the data is linearised and densely connected.

## **CNN** Example

```
image = cv2.imread('uniLogo.png')
image=cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
image=image/255
plt.imshow(dmage)
plt.axis('off')
plt.show()
print("Shape: ",image.shape)
print("Size: ",image.size,image.size/image.size*100, "%")
```



Shape: (225, 225, 3) Size: 151875 100.0 %

Figure: Load image code

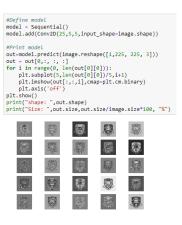


Figure: Tensorflow single Convolution code

#### Generator

A generator is based on a convolutional neural network followed by a de-convolutional neural network.



Figure: Convolutional to de-convolutional neural network

## Project aims

There are two main aims for this project:

- Create a novel saliency prediction algorithm using a GAN.
- Compare saliency prediction with RGB and RGB-D

## Methodology

- Choose GAN generator and discriminator design
- Train 3 GANs to each accept: RGB, Depth and RGB-D separately
- Use other combination methods on RGB and depth data to compare ours to.
- Quantify model efficiency

#### Results

Two different methods were used to quantify model accuracy:

- F-measure
- Receiver operating characteristic curve (ROC)

Both of these are based on True positive rate, false positive rate, true negative rate and false negative rate.

#### F-Measure

F-Measure is a numeric value calculated as the harmonic mean of precision and recall:

$$F = 2 \cdot \frac{\frac{TP}{TP + FP} \cdot \frac{TP}{TP + FN}}{\frac{TP}{TP + FP} + \frac{TP}{TP + FN}} \tag{1}$$

- TP True positive rate
- P False positive rate
- FN False negative rate

F-Measure ranges from 1-0 with 1 meaning 100% true positive rate and 0% false positives/negatives.

### Receiver operating characteristic

ROC is a line generated by varying a threshold on an image and comparing that to ground truth.



Figure: Varying threshold on saliency map

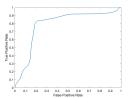


Figure: Single ROC

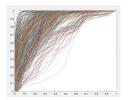


Figure: All ROC in dataset

#### Results

Model	F-Measure	AUC
Ours(RGB)	0.7248	0.9227
GBVŠ(RGÉ)	0.3163	0.7052
Itti(RĠB)	0.3097	0.7380
LMH ´	0.1878	0.7646

Model	F-Measure	AUC
GP	0.5607	0.8668
GBVS (RGB)	0.5113	0.8927
Itti (RĠB)	0.4221	0.8597
Ours (*)	0.3950	0.7731
LMH`´	0.3208	0.7921

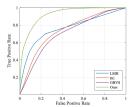


Figure: Olesova results

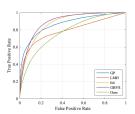


Figure: NLPR results

## Results - Depth

Combination	F - Measure	AUC
RGB	0.7248	0.9227
RGBDepth	0.6866	0.8723
RGB + Depth	0.6121	0.8674
RGB*Depth	0.4144	0.8227
Depth .	0.4131	0.6821

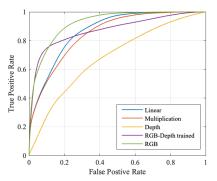


Figure: Our combination methods compared

#### **Future Work**

#### A few ideas:

- Design neural network to take multiple inputs (3/4 channel)
- Run network on a larger epoch system
- Train/Test on similar saliency datasets
- Implement the network in a system such as compression

#### Conclusion

We set out with the following aims:

- Create a novel saliency prediction algorithm using a GAN.
- Compare saliency prediction with RGB and RGB-D

# Any Questions?