

# TWO ALTERNATIVE FORCED CHOICE CLASSIFIERS

Symbolic and Evolutionary Artificial Intelligence's Project

Stefano Petrocchi



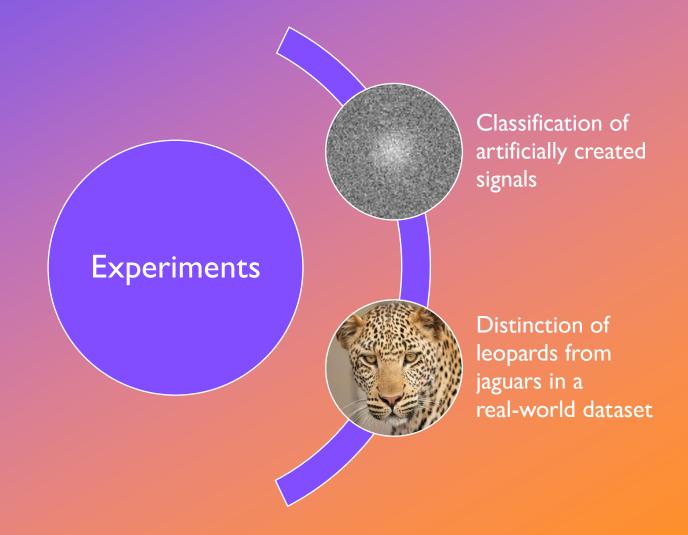


#### Introduction

# Objectives

Verify that the two alternative forced choice (TAFC) problem is simpler than the single image classification problem

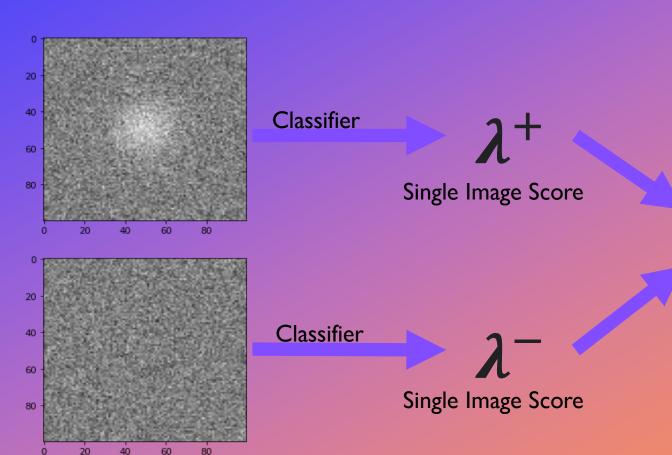
Verify that a Siamese network is better suited to the TAFC than a single image classifier





#### Two Alternative Forced Choice

#### Single Image Classifier:



$$s = \begin{cases} 1 & if \ \lambda^+ > \lambda^- \\ 0 & if \ \lambda^+ < \lambda^- \end{cases}$$

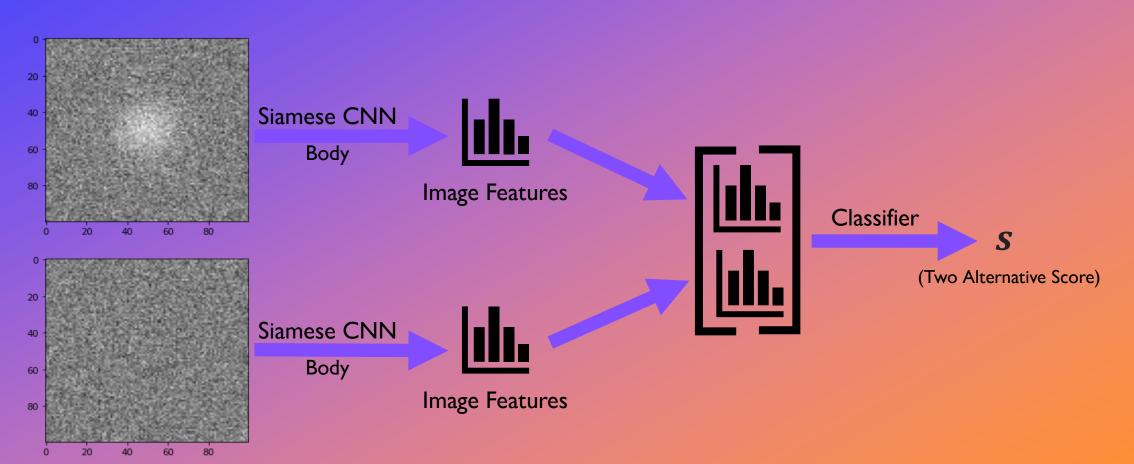
Two Alternative Score



0

#### Two Alternative Forced Choice

#### Siamese Classifier:





#### Objectives

# Single Image Classification:

Classify single images by the presence or absence of the signal

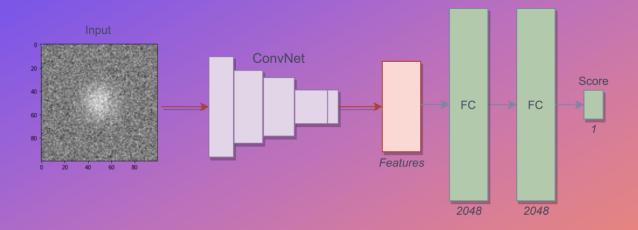
Two Alternative Images
Classification:

Two images are presented to the network that must distinguish which of the two input contains the signal

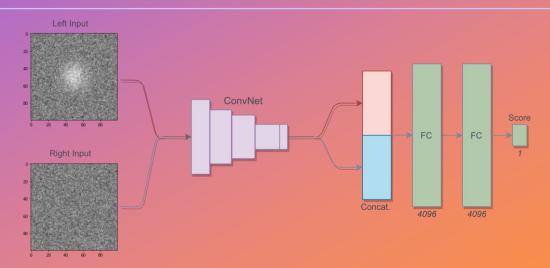


# Models

Single Image Classifier



Siamese Network Koch et al. 2015

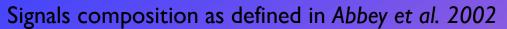


0



O

# Artificial Signals Experiment





#### **Parameters**

(Decided Heuristically)

Background

Noise

Signal

value ∈ [0.2, 0.8]

 $\mu = 0$ 

 $\sigma \in [0.01, 0.2]$ 

 $\overline{\mu} = 0$ 

 $\sigma \in [5,25]$ 

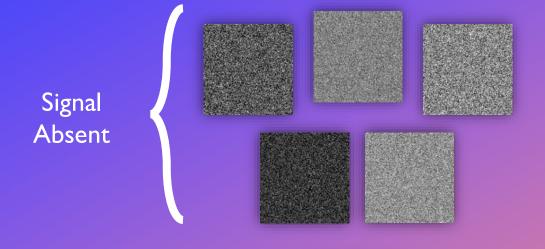
 $\alpha \in [0.005, 0.05]$ 



0

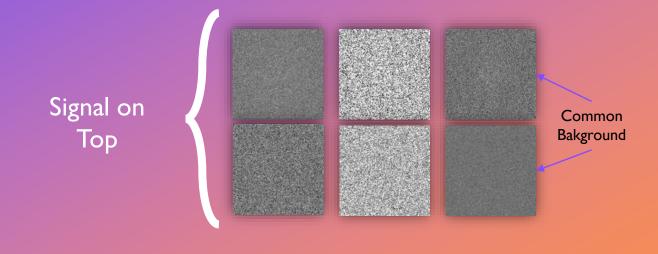
# Artificial Signals Experiment

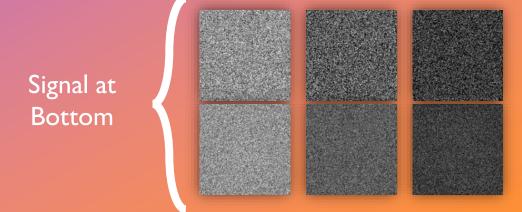
#### Single Image Classification:





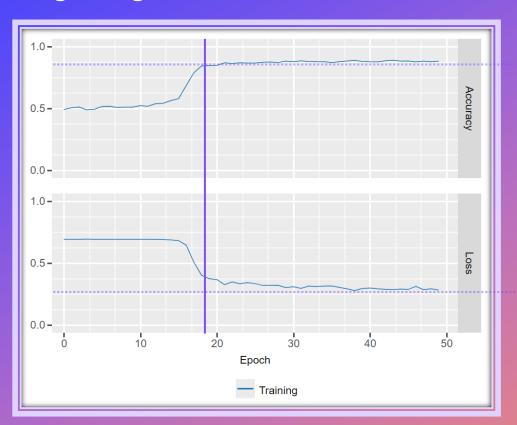
#### Two Alternative Images Classification:



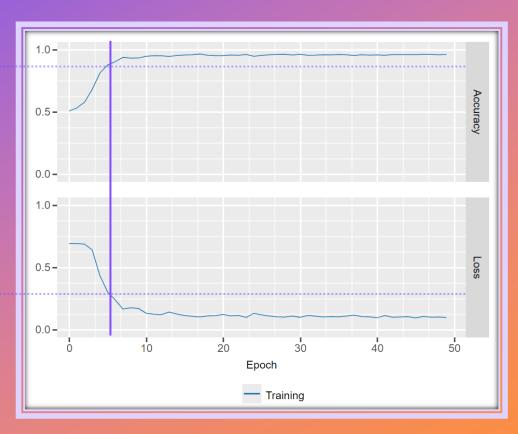




#### Single Image Model:



#### Siamese Model:

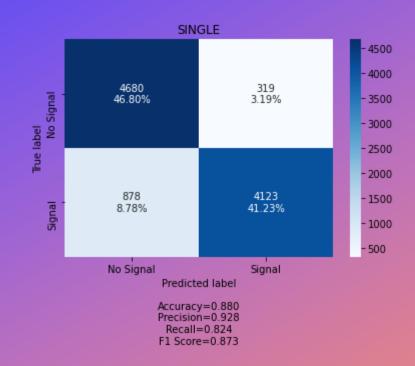


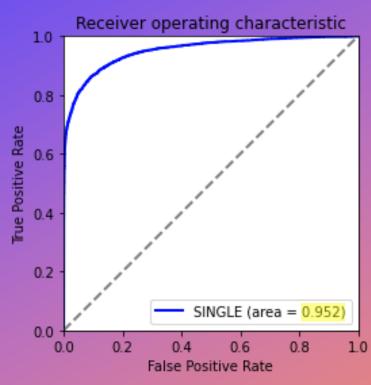
Training performed over 50 epochs (composed by 128 batches of 32 samples) to allow comparison. The data is artificially generated so it is not necessary to use a validation set as overfitting cannot occur.



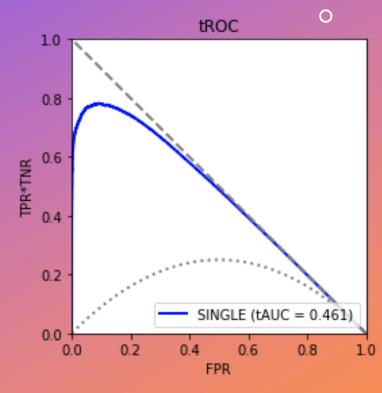
Single Input Results

To evaluate the results obtained from the single input model, a test set consisting of 10 000 single images was generated





It should be noted that in the TAFC problem, as expected for large test sets, the value of the single input model's accuracy is the same as the AUC for the single images classification



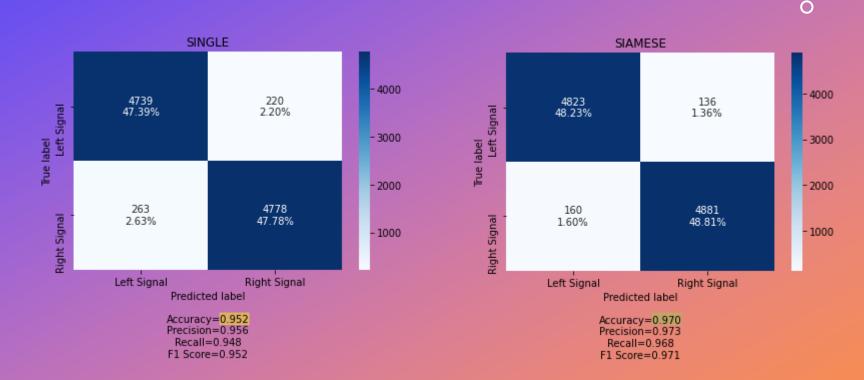
As described in Marafini's thesis, it is possible to obtain the *tROC* curve by replacing the TPR in the ordinates with TPR \* TNR



TAFC Results

Also for the TAFC problem, 10 000 pairs of images were generated to evaluate the results obtained from the Siamese model compared with the results obtained from the single input model

To do this, the images in pairs are supplied one at a time to the single input classifier and then the two scores obtained are compared, the image with the highest score is the one classified as containing the signal



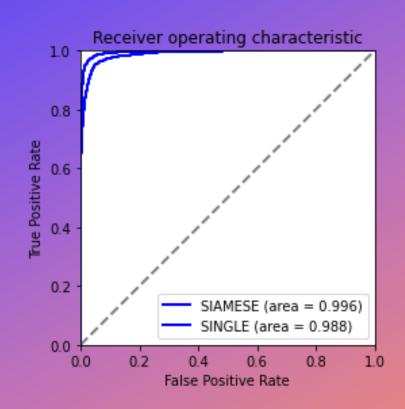
By comparing the accuracy value of the Siamese model with that of the single input model, it can be seen that this has a better accuracy. This is probably due to the fact that the Siamese model is able to directly compare images as opposed to the single model that sees them individually.

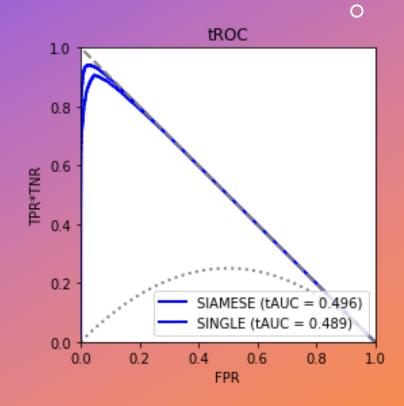


To draw the ROC and tROC curves also for the single input model, the confidence in the classification of a pair of images was interpreted with the following formula:

$$c = \begin{cases} \frac{1 + (r - l)}{2} & \text{if } r \ge l \\ \frac{1 - (l - r)}{2} & \text{if } r < l \end{cases}$$

Given that the optimal confidence is 1 if  $r \ge l$  and 0 otherwise





It can be seen that both for the ROC curve and for the tROC curve the value of AUC and tAUC is better for the Siamese model than for the single input model, confirming the hypotheses.

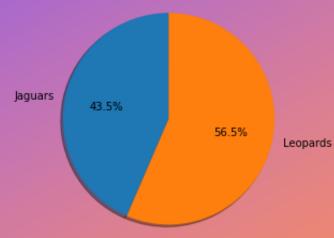


#### Dataset

The results obtained from the previous experiment with artificial data were validated using a realworld dataset consisting of photos of leopards and jaguars.

#### Models

In this experiment 4 different models were compared: two single input networks and two Siamese networks, in which one is defined from scratch and the other uses a pretrained ResNet-50



The dataset is composed of 1418 images divided for 60% in training set, for 20% in validation set and for the remaining 20% in test set, using a stratified split.



Training

Two different trainings were carried out, one for the single image classification and one for the TAFC one

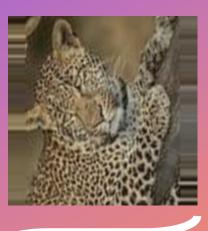
Strong augmentation was applied to the training batches to compensate for the small size of the dataset

Batches for the TAFC are made by random permutations of training set images, while validation set is decided initially and remains constant

Early stopping with 5 epochs patience was used for all models

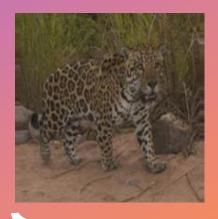














Leopards

Jaguars



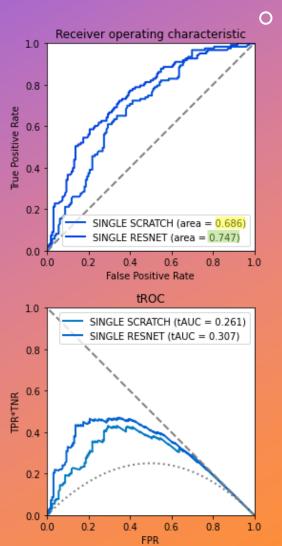
Single Input Results To the side are reported the results obtained from the single input models obtained using the test set

It can be seen that the from scratch model is not able to achieve optimal accuracy, since the problem is difficult and the data is small

The pretrained model, on the other hand, manages to achieve better results, even if not optimal

It is interesting to note that the AUC values of the two models are similar to their accuracy in the TAFC problem in the next slide





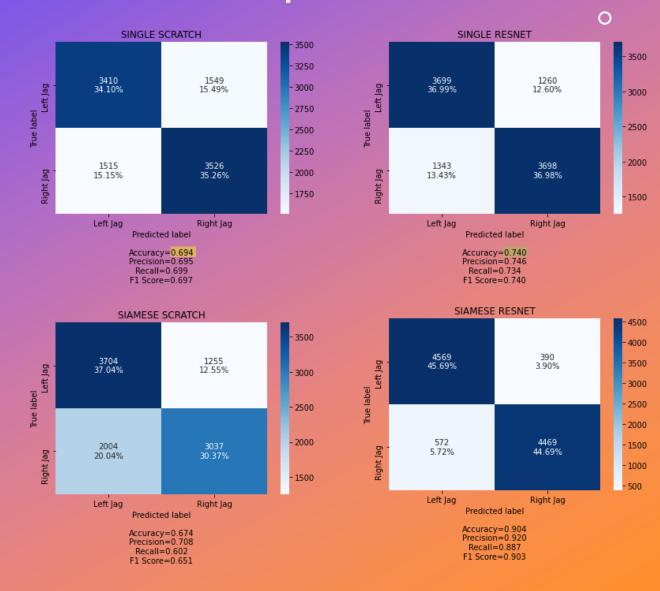


#### TAFC Results

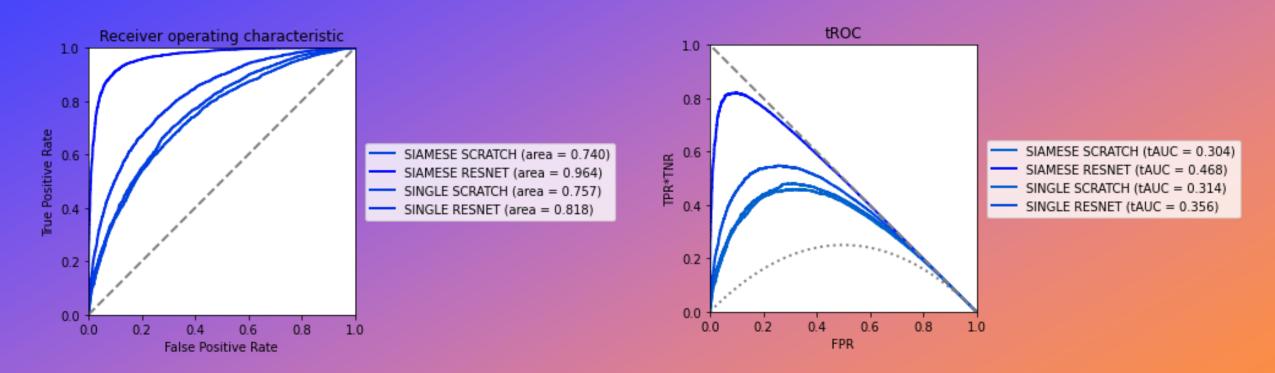
The test set for the TAFC problem consists of 10 000 pairs of images (note that about a 40 000 different test set images permutations are possible)

It is interesting to note that the Siamese scratch model performs worse than its single input counterpart

As expected, the pretrained Siamese version has the best results, confirming, together with the experiment with synthetic data, that the Siamese model is better than a single model for the TAFC problem







The ROC and tROC curves confirm the considerations made previously. It is possible to notice how the pretrained Siamese model stands out in terms of effectiveness, as it can consider both inputs at the same time and reduce overfitting thanks to the many possible permutations of the images and to the weights already initialized.



+

0

# QUESTIONS?