

# Report on Assignment 8

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## Summary

In this project we have to implement different Regression models on the dataset provided. The dataset consists of training and testing images each of which has two folders the background images and the face images. We have to train different regression models using the training data supplied and have to classify the testing images into either face or background.

The different regression models to be implemented:

1. Logistic Regression
2. Bayesian Logistic Regression
3. Dual Logistic Regression
4. Dual Bayesian Logistic Regression
5. Kernel Logistic Regression
6. Relevance Vector Logistic Regression

## Algorithmic Approach

### Pre-processing:

1. All the images are read into matrices, then the face and the background data are combined into a single matrix. Then the WTrain and WTest matrices are generated, these are initialized values (0 or 1) in the order in which the above face and background matrices are combined.
2. Different methods and their corresponding cost functions are provided so we have to transform the data to fit the requirements.
3. Different parameters like VarPrior, InitialPhi, InitialPsi, Lambda, Nu are set accordingly and supplied to the fit function of the corresponding method.
4. Here the fminunc function will utilize the parameters supplied and compute the gradient vector and hessian matrix using the cost function.
5. This gives the prediction values on the test images.
6. We then convert these values using a step operation where the values greater than 0.5 are equated to one and those less than 0.5 are equated to 0.
7. These values are now compared with the ground truth WTest.
8. During this process the Miss Detection Rate and the False Alarm Rate are calculated for Face and Background respectively.

# Results

The models were run in different colorspace:

## RGB

Results:

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1. Logistic Regression:

Miss Detection Rate: 0.025126  
False - Alarm Rate: 0.156977

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2. Bayesian Logistic Regression:

Miss Detection Rate: 0.050251  
False - Alarm Rate: 0.203488

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3. Dual Logistic Regression:

Miss Detection Rate: 0.025126  
False - Alarm Rate: 0.162791

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4. Dual Bayesian Logistic Regression:

Miss Detection Rate: 0.025126  
False - Alarm Rate: 0.156977

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5. Kernel Logistic Regression:

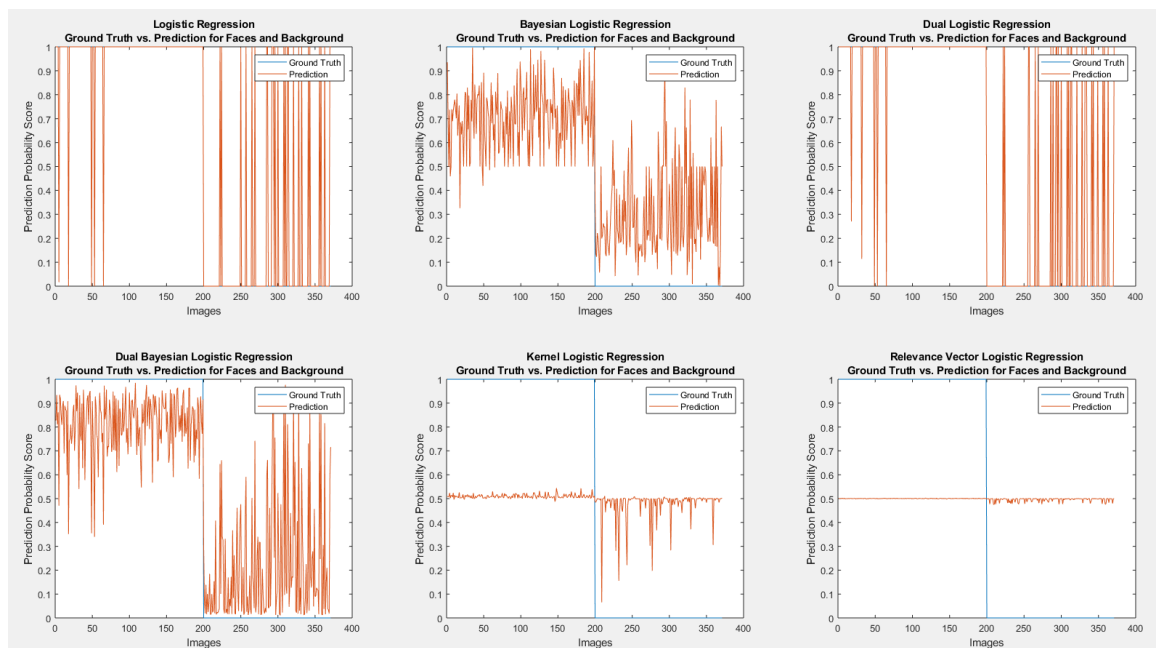
Miss Detection Rate: 0.040201  
False - Alarm Rate: 0.313953

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6. Relevance Vector Logistic Regression:

Miss Detection Rate: 0.256281  
False - Alarm Rate: 0.151163

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### Observation:

We can observe that the Logistic Regression method classifies the face data better than the background data. The Bayesian does a good job but has less accuracy compared to the previous method. The Dual LR Works similar to the LR method and there is an increase in accuracy in the Dual BLR. Finally, the Kernel and the RVLr perform way worse than its predecessors. All their face and background predictions lie close to 0.5 with high degree of ambiguity

## HSV

### Results:

#### 1. Logistic Regression:

Miss Detection Rate: 0.190955  
False - Alarm Rate: 0.215116

#### 2. Bayesian Logistic Regression:

Miss Detection Rate: 0.060302  
False - Alarm Rate: 0.209302

#### 3. Dual Logistic Regression:

Miss Detection Rate: 0.195980  
False - Alarm Rate: 0.203488

#### 4. Dual Bayesian Logistic Regression:

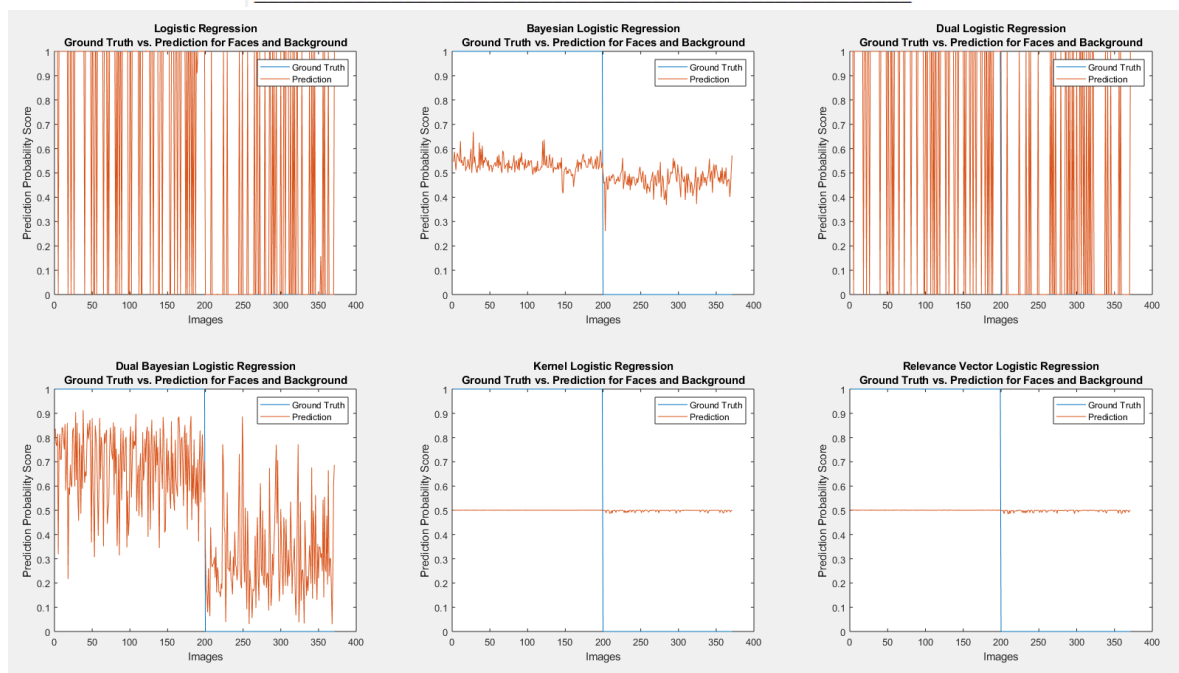
Miss Detection Rate: 0.145729  
False - Alarm Rate: 0.139535

#### 5. Kernel Logistic Regression:

Miss Detection Rate: 0.175879  
False - Alarm Rate: 0.063953

#### 6. Relevance Vector Logistic Regression:

Miss Detection Rate: 0.180905  
False - Alarm Rate: 0.063953



### Observations:

There is a high degree of ambiguity in the logistic regression model, we can see that the model classifies the data as both face and background in both datasets. The Bayesian LR method performs better, we can see that it classifies max face data as face and max background data as background. The dual LR performs similar to the LR method classifying the data as both in face and background sets. The Dual BLR works similar to the BLR but with variable prediction accuracy. Finally, the Kernel and RLVR show negligent change in the face data and very small change in the background data. These are by far the worst among the methods again.

## YCbCr

### Results:

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#### 1. Logistic Regression:

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Miss Detection Rate: 0.060302  
False - Alarm Rate: 0.122093

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#### 2. Bayesian Logistic Regression:

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Miss Detection Rate: 0.065327  
False - Alarm Rate: 0.186047

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#### 3. Dual Logistic Regression:

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Miss Detection Rate: 0.185930  
False - Alarm Rate: 0.122093

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#### 4. Dual Bayesian Logistic Regression:

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Miss Detection Rate: 0.115578  
False - Alarm Rate: 0.139535

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#### 5. Kernel Logistic Regression:

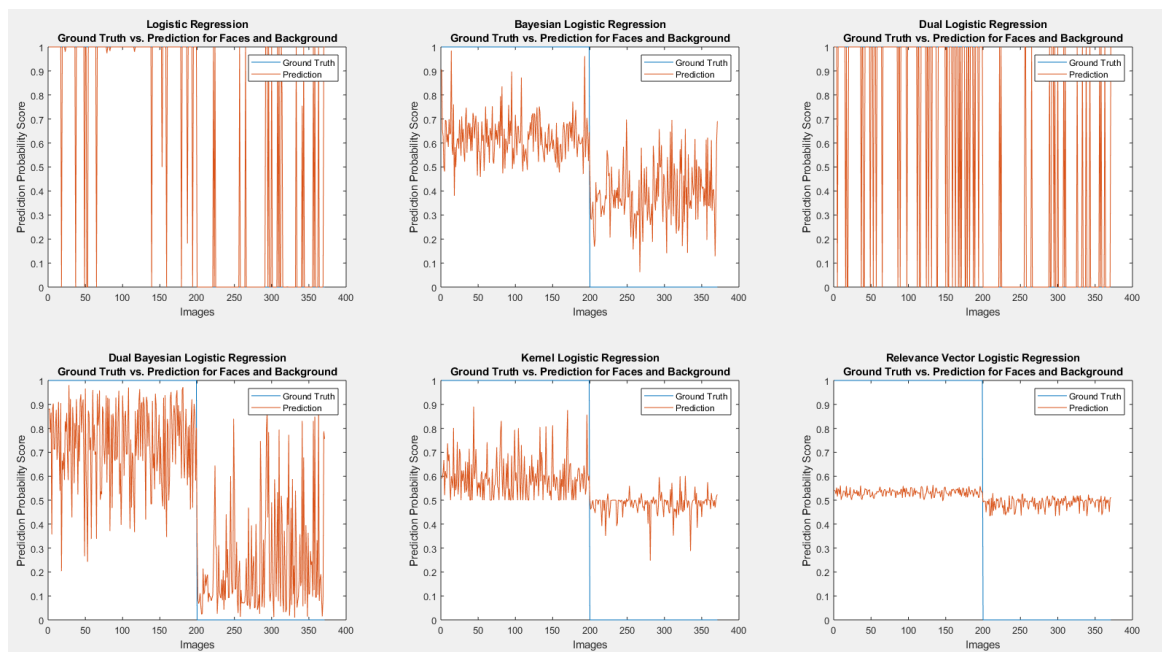
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Miss Detection Rate: 0.005025  
False - Alarm Rate: 0.447674

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#### 6. Relevance Vector Logistic Regression:

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Miss Detection Rate: 0.005025  
False - Alarm Rate: 0.308140

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### Observations:

In this colorspace the first four methods show high degree of ambiguity in the prediction result classifying the data as face and background in both the datasets. While the last two methods, the kernel and RVLRL work efficiently classifying the data as face and background in both the datasets.

## Gray

### Results:

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1. Logistic Regression:  
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Miss Detection Rate: 0.231156  
False - Alarm Rate: 0.186047

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2. Bayesian Logistic Regression:  
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Miss Detection Rate: 0.522613  
False - Alarm Rate: 0.098837

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3. Dual Logistic Regression:  
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Miss Detection Rate: 0.165829  
False - Alarm Rate: 0.174419

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4. Dual Bayesian Logistic Regression:  
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Miss Detection Rate: 0.221106  
False - Alarm Rate: 0.197674

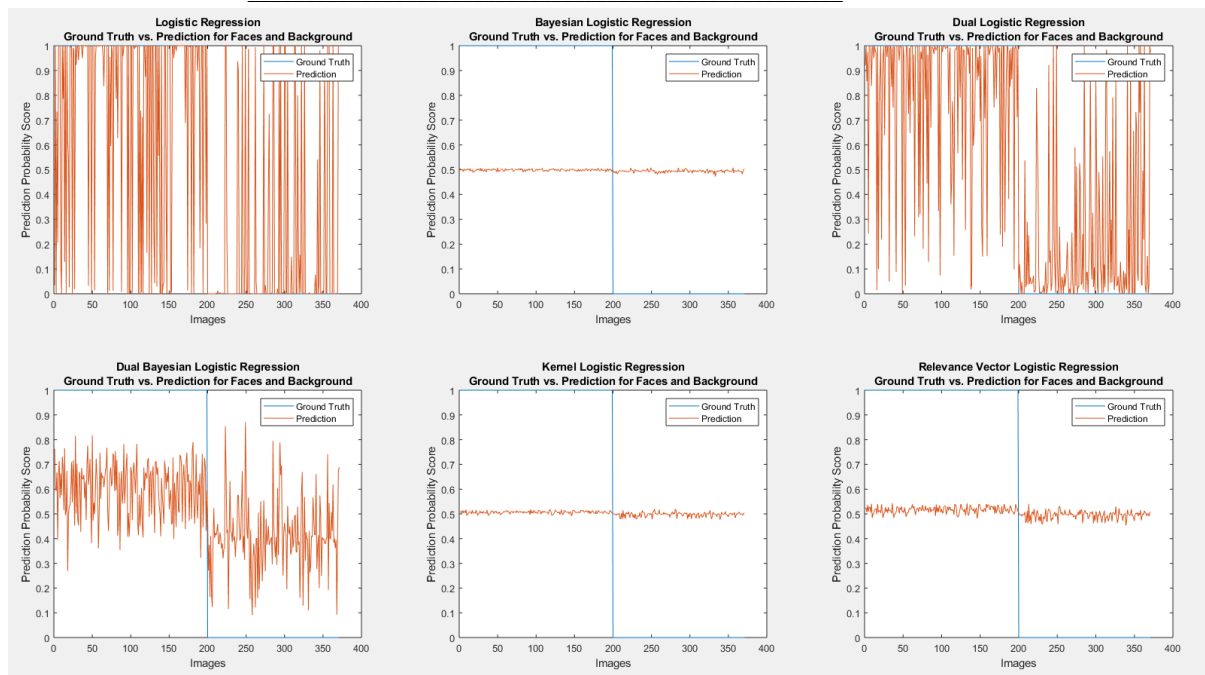
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5. Kernel Logistic Regression:  
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Miss Detection Rate: 0.100503  
False - Alarm Rate: 0.383721

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6. Relevance Vector Logistic Regression:  
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Miss Detection Rate: 0.115578  
False - Alarm Rate: 0.366279

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### Observations:

In this final colorspace we can observe that LR and Dual LR work worst showing very high degree of ambiguity in classifying the data as face and background in both the datasets. The BLR and Kernel and RVLr do a fairly decent job at predicting the face and background in the respective datasets and the Dual BLR although predicts most of the face images correctly, it shows high degree of ambiguity in the background prediction!