

# Bank Dataset

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Clear workspace

```
clear all; clc; close all;
```

## Initialize variables

Load tables

```
training_set = readtable('bank-sample.csv')
```

```
training_set = 2000x17 table
```

	age	job	marital	education	default	balance	housing
1	31	'management'	'single'	'tertiary'	'no'	0	'yes'
2	45	'entrepreneur'	'married'	'tertiary'	'no'	1752	'yes'
3	46	'services'	'divorced'	'secondary'	'no'	4329	'no'
4	35	'management'	'married'	'tertiary'	'no'	1108	'yes'
5	39	'management'	'married'	'secondary'	'no'	1410	'yes'
6	31	'management'	'single'	'tertiary'	'no'	499	'yes'
7	34	'entrepreneur'	'married'	'tertiary'	'no'	0	'yes'
8	39	'admin.'	'married'	'secondary'	'no'	26233	'no'
9	38	'blue-collar'	'married'	'secondary'	'no'	8444	'yes'
10	50	'management'	'married'	'tertiary'	'yes'	72	'no'
11	58	'blue-collar'	'married'	'secondary'	'no'	1075	'yes'
12	36	'management'	'divorced'	'secondary'	'no'	47	'no'
13	35	'blue-collar'	'married'	'secondary'	'no'	635	'yes'
14	72	'retired'	'married'	'secondary'	'no'	17739	'no'

⋮

```
test_set = readtable('bank-sample-test.csv');
```

Transform dates

```
training_set.date = datenum(strcat(num2str(training_set.day),training_set.month),'ddmmm');
```

```
test_set.date = datenum(strcat(num2str(test_set.day),test_set.month),'ddmmm');
```

Transform logarithmic variables

```
training_set.balance = log(max(training_set.balance,1));
training_set.duration = log(training_set.duration + 1);
```

```
test_set.balance = log(max(test_set.balance,1));
test_set.duration = log(test_set.duration + 1);
```

Transform categorical variables

```
training_set.job      = categorical(training_set.job);
training_set.marital  = categorical(training_set.marital);
training_set.education = categorical(training_set.education);
training_set.default  = categorical(training_set.default);
training_set.housing  = categorical(training_set.housing);
training_set.loan     = categorical(training_set.loan);
training_set.contact  = categorical(training_set.contact);
training_set.month    = categorical(training_set.month);
training_set.month    = reordercats(training_set.month, ["jan", "feb", "mar", "apr", "may", "jun", "jul", "aug", "sep", "oct", "nov", "dec"]);
training_set.poutcome = categorical(training_set.poutcome);

training_set.subscribed = categorical(training_set.subscribed);
```

```
test_set.job      = categorical(test_set.job);
test_set.marital  = categorical(test_set.marital);
test_set.education = categorical(test_set.education);
test_set.default  = categorical(test_set.default);
test_set.housing  = categorical(test_set.housing);
test_set.loan     = categorical(test_set.loan);
test_set.contact  = categorical(test_set.contact);
test_set.month    = categorical(test_set.month);
test_set.month    = reordercats(test_set.month, ["jan", "feb", "mar", "apr", "may", "jun", "jul", "aug", "sep", "oct", "nov", "dec"]);
test_set.poutcome = categorical(test_set.poutcome);

test_set.subscribed = categorical(test_set.subscribed);
```

## Explore data

```
subscribed = training_set(training_set.subscribed == "yes",:);
unsubscribed = training_set(training_set.subscribed == "no",:);
```

## Histograms with category info

```
nbins = 25;

for col = 1: width(training_set)
```

```

figure

if iscategorical(table2array(training_set(:,col)))
    Y(1,:) = countcats( table2array(unsubscribed(:,col)) );
    Y(2,:) = countcats( table2array( subscribed(:,col)) );

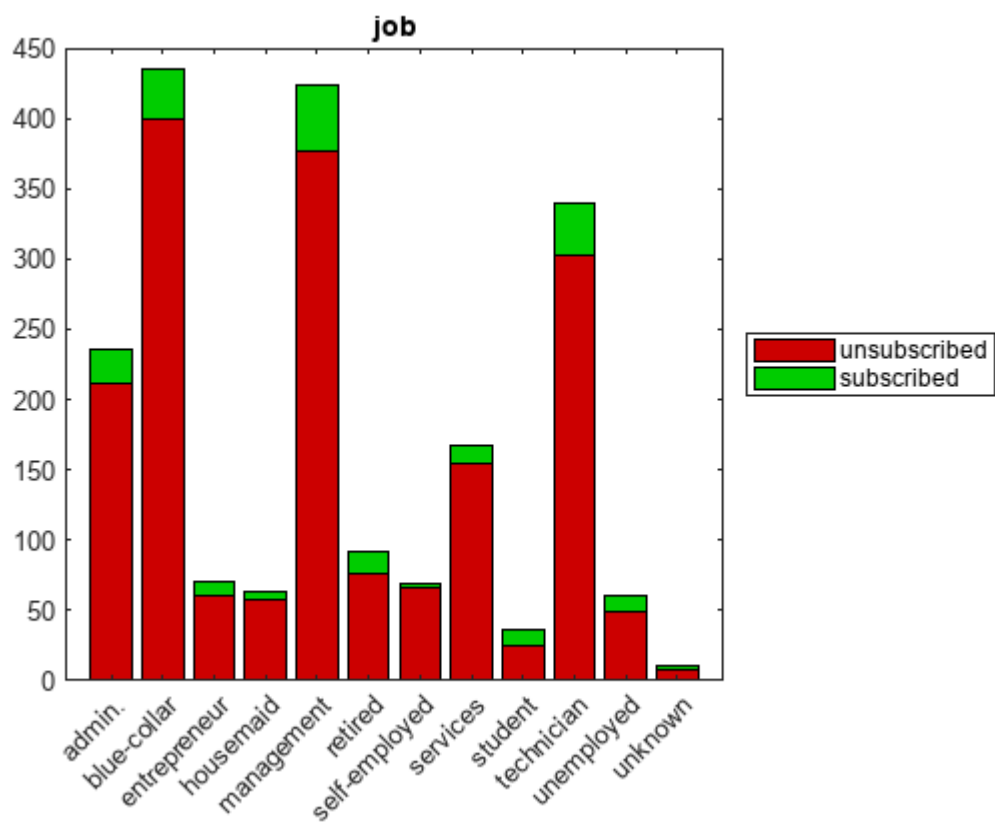
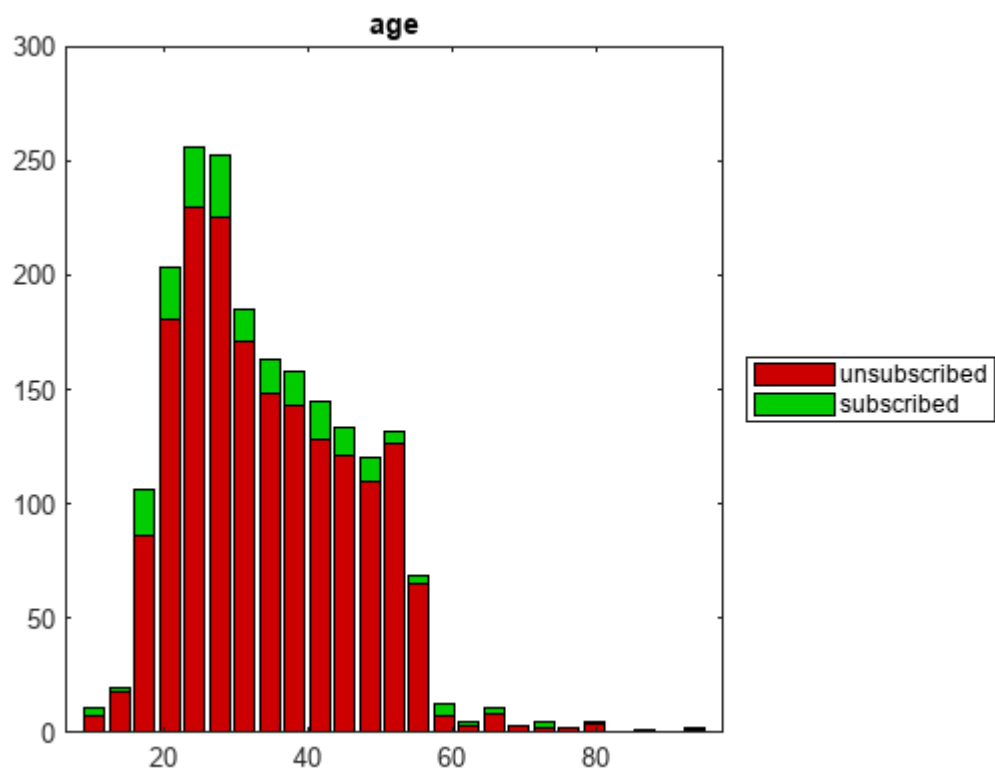
    ba = bar(categorical(categories(table2array(training_set(:,col)))), ...
        Y', 'stacked', 'FaceColor','flat');
    ba(1).CData = [0.8 0 0];
    ba(2).CData = [0 0.8 0];
    clear Y
else
    hist_edges = linspace(min(table2array(training_set(:,col))), ...
        max(table2array(training_set(:,col))), nbins+1);
    plot_edges = linspace(hist_edges(2)/2, (hist_edges(end-1)+hist_edges(end))/2, nbins);

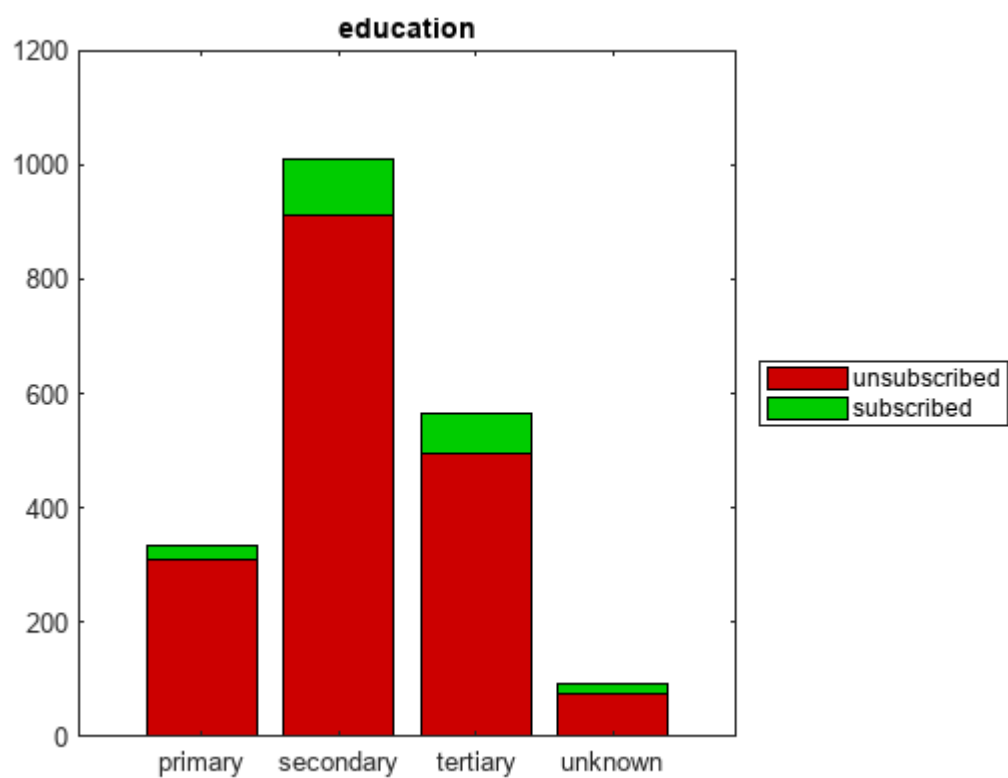
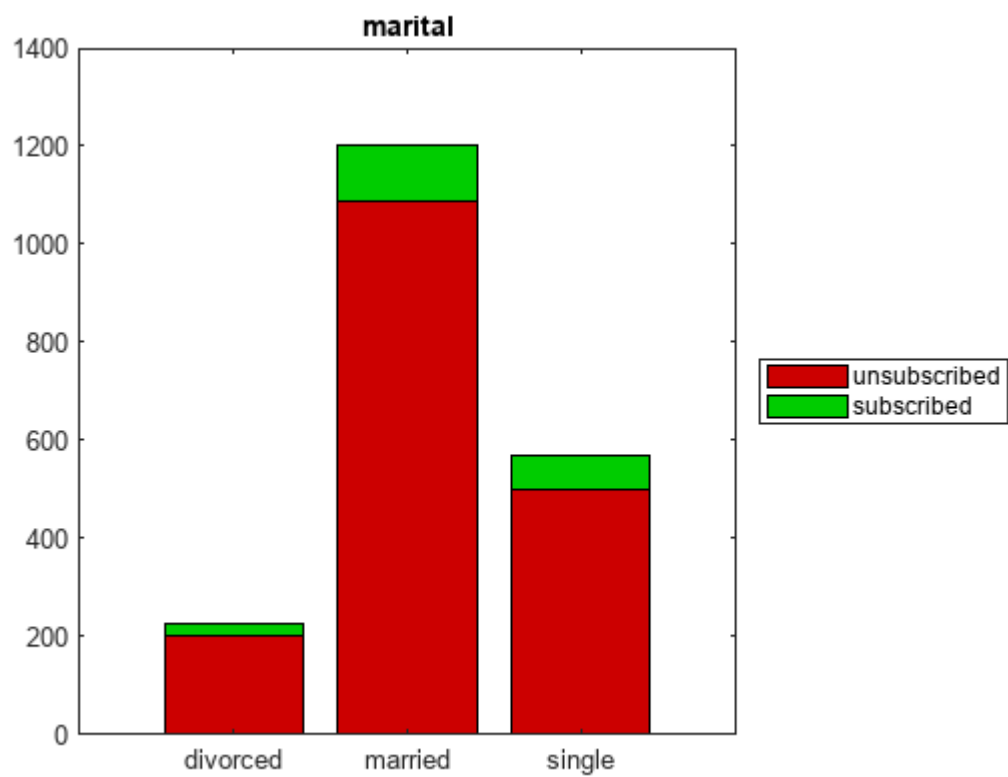
    X(1,:) = histcounts( table2array(unsubscribed(:,col)), hist_edges);
    X(2,:) = histcounts( table2array( subscribed(:,col)), hist_edges);

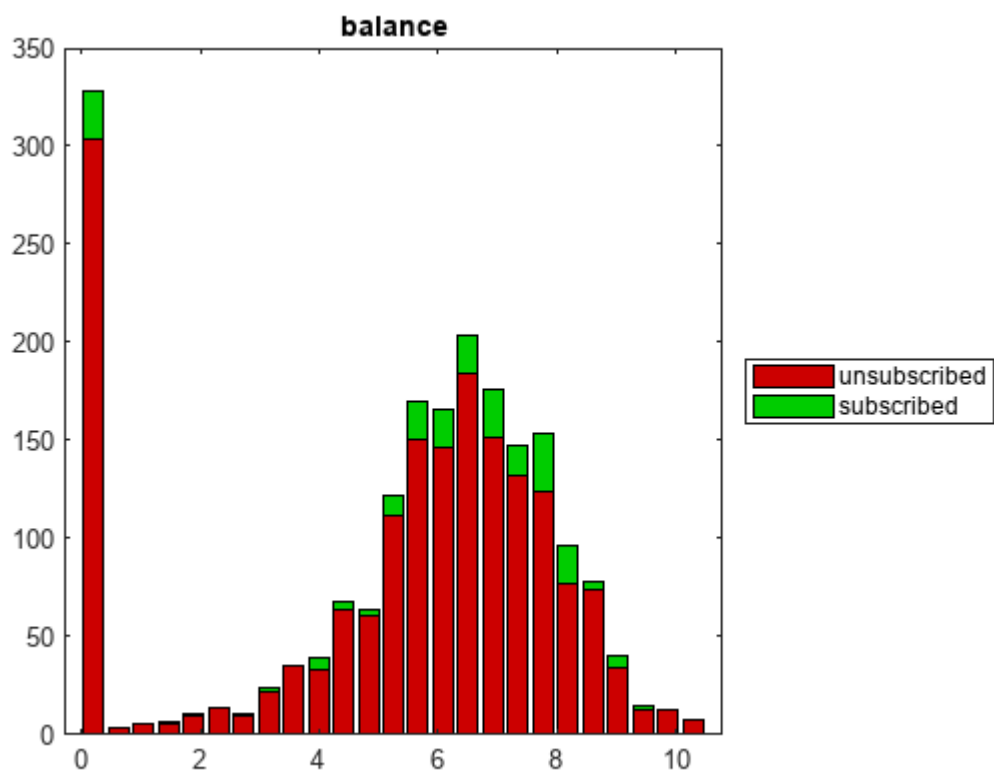
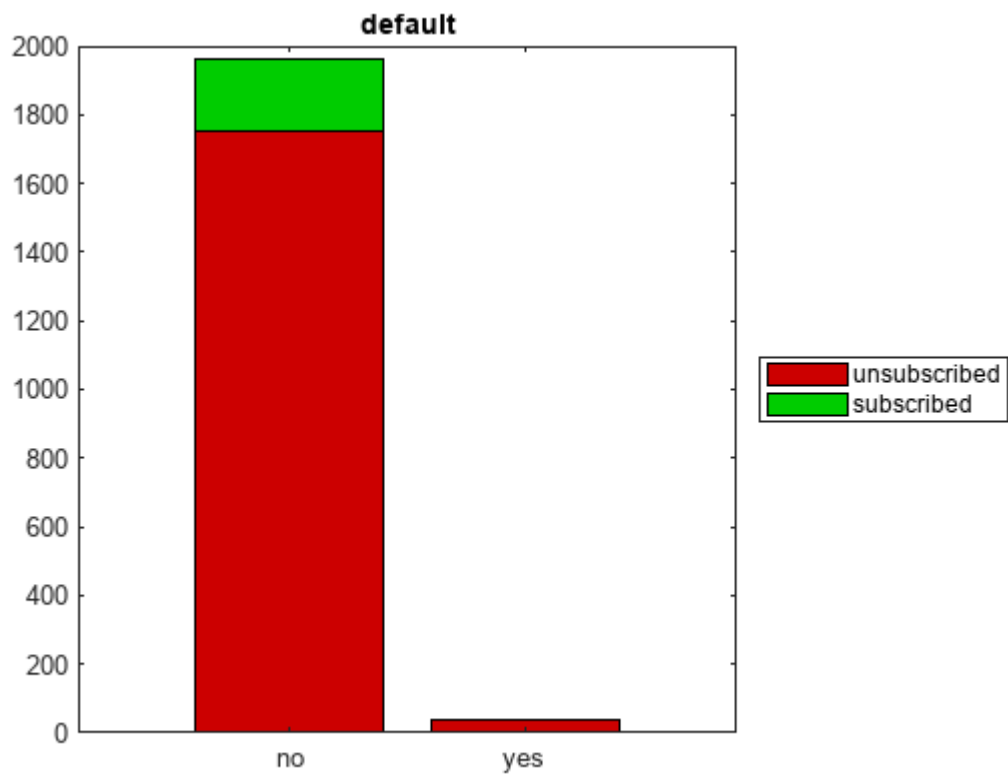
    ba = bar(plot_edges, X', 'stacked', 'FaceColor','flat');
    ba(1).CData = [0.8 0 0];
    ba(2).CData = [0 0.8 0];
end

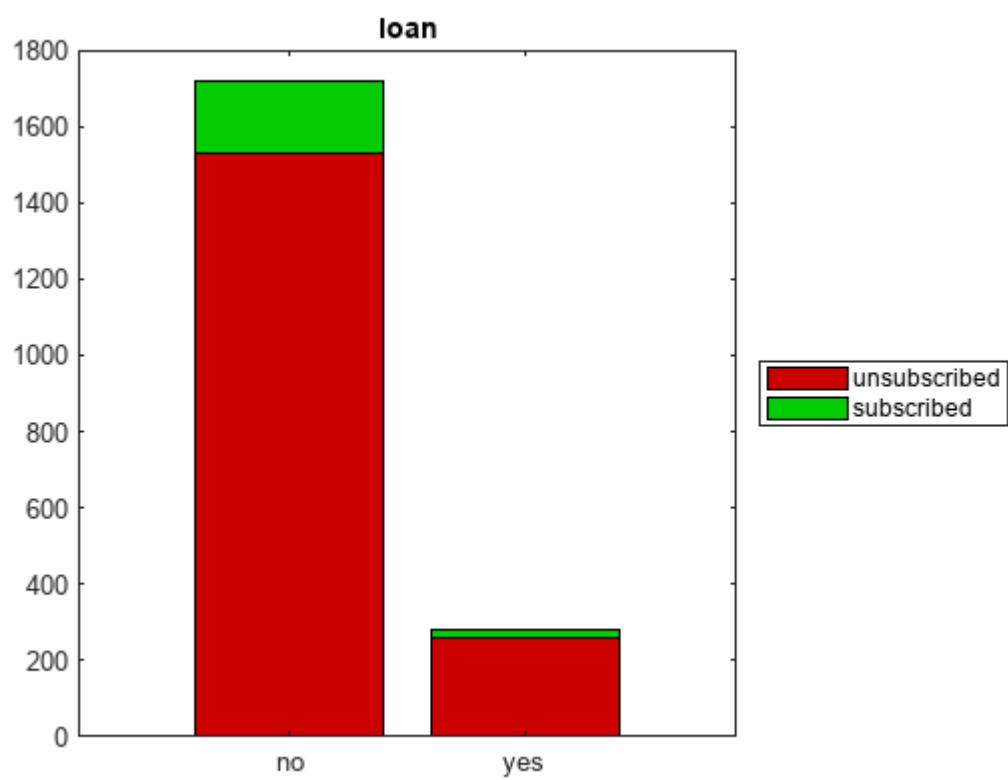
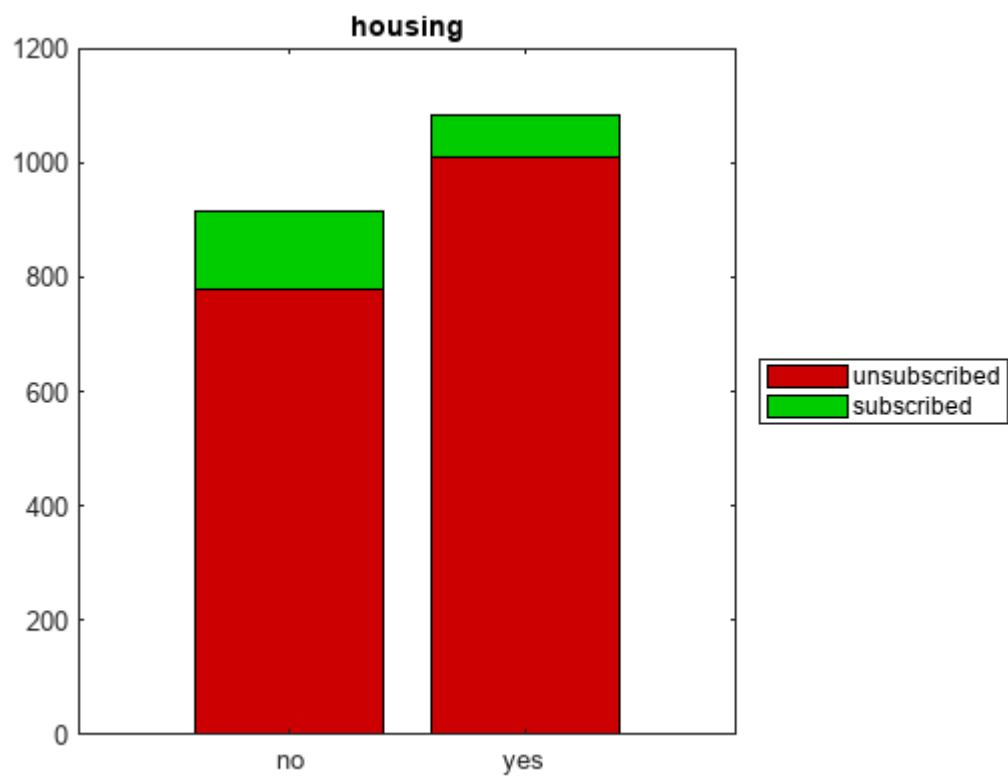
title(training_set.Properties.VariableNames{col})
if col == 5
    title(' default ')
end
legend('unsubscribed','subscribed','Location','eastoutside')
% saveas(gcf, strcat('figures/Hist_training_', test_set.Properties.VariableNames{col}, '.png'))
end

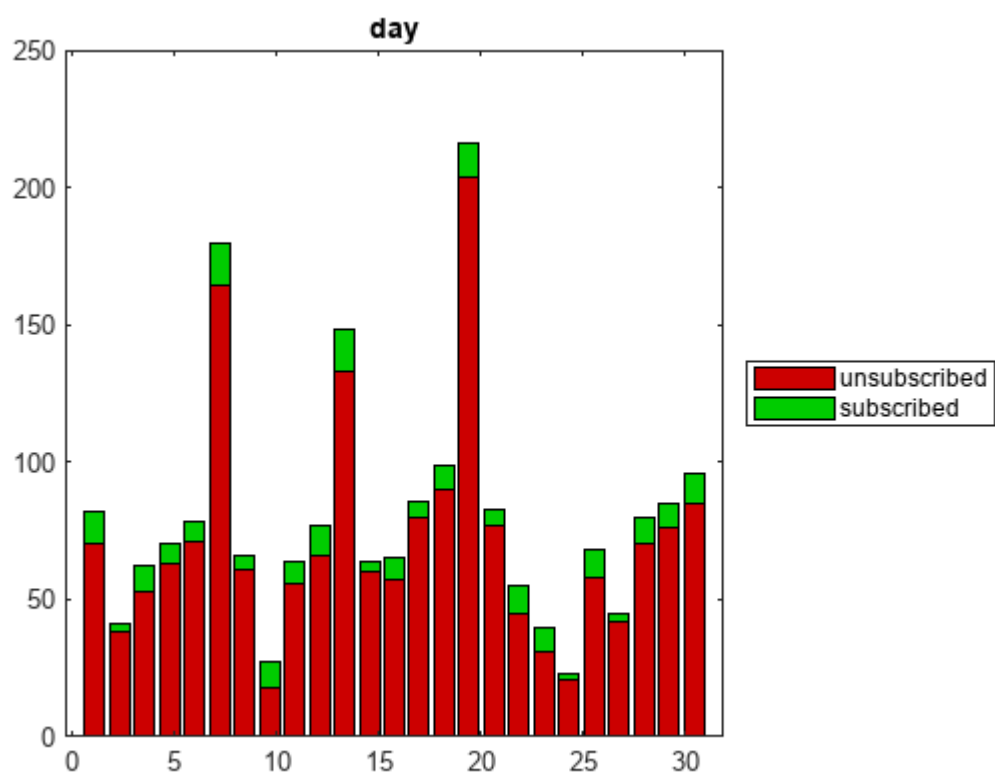
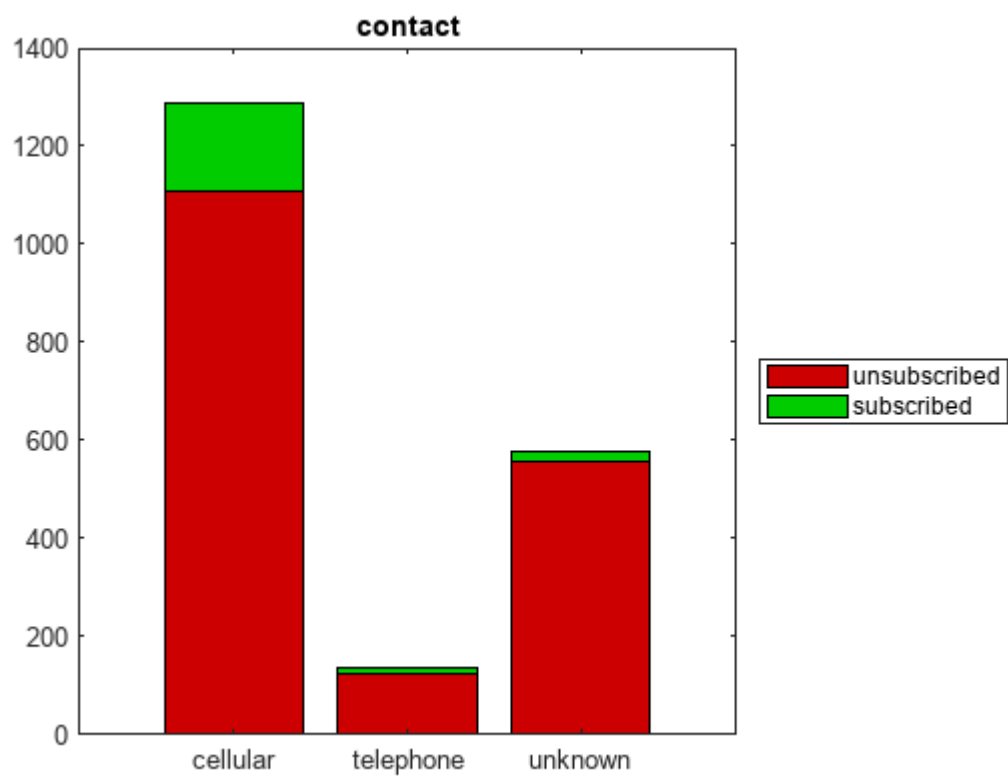
```



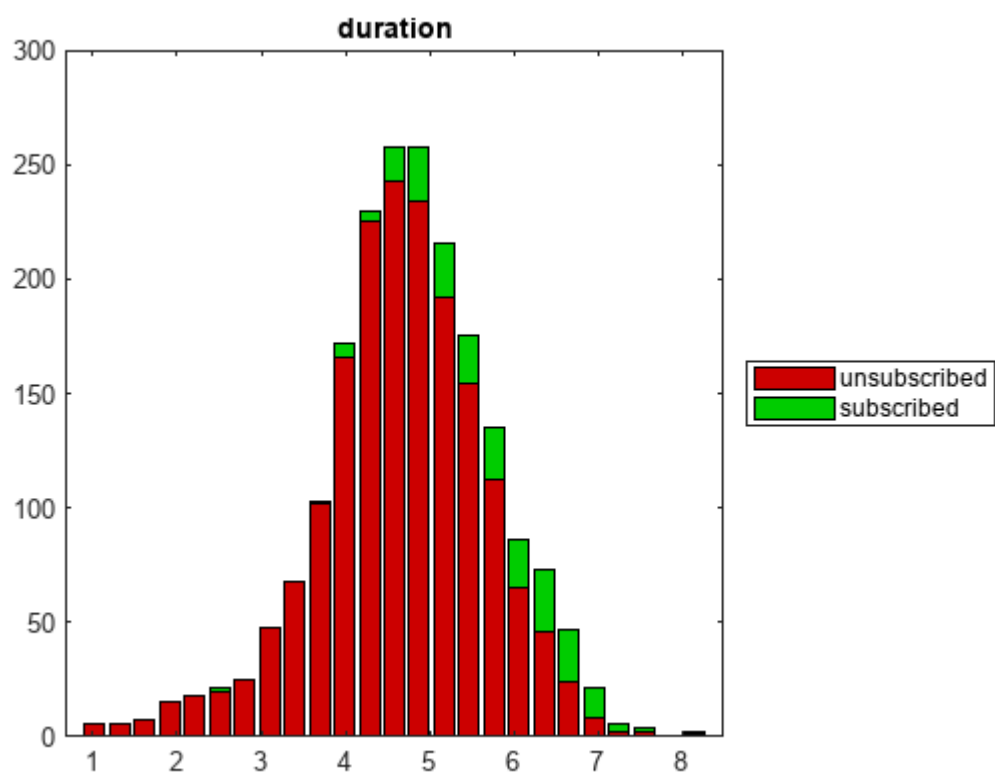
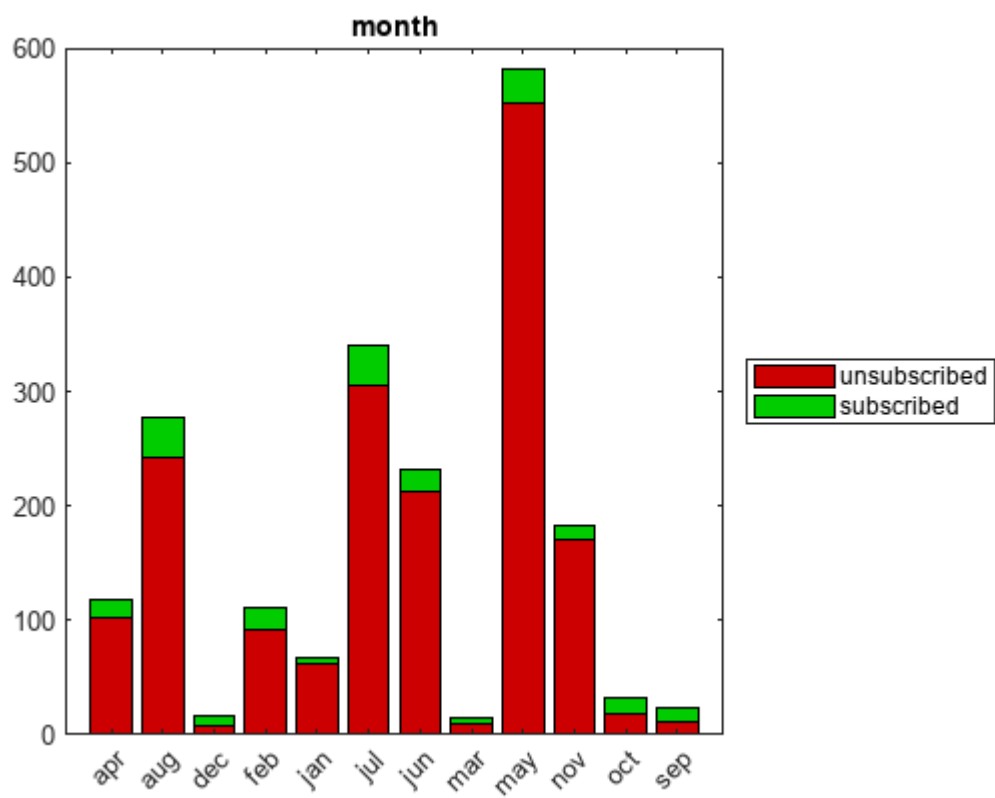


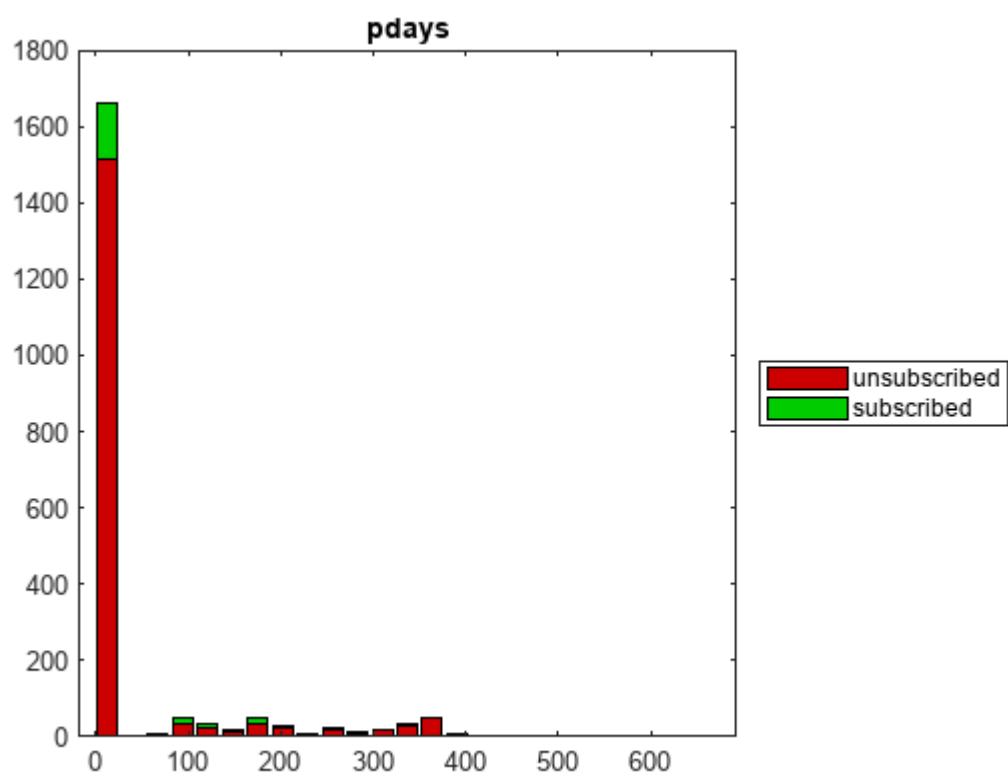
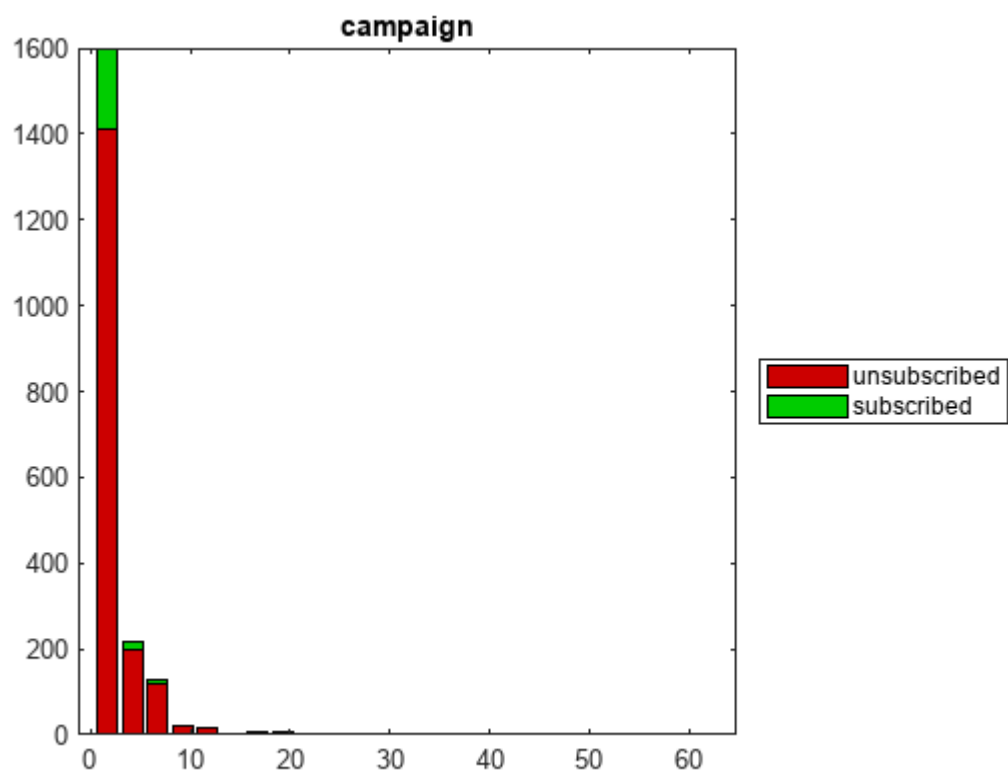


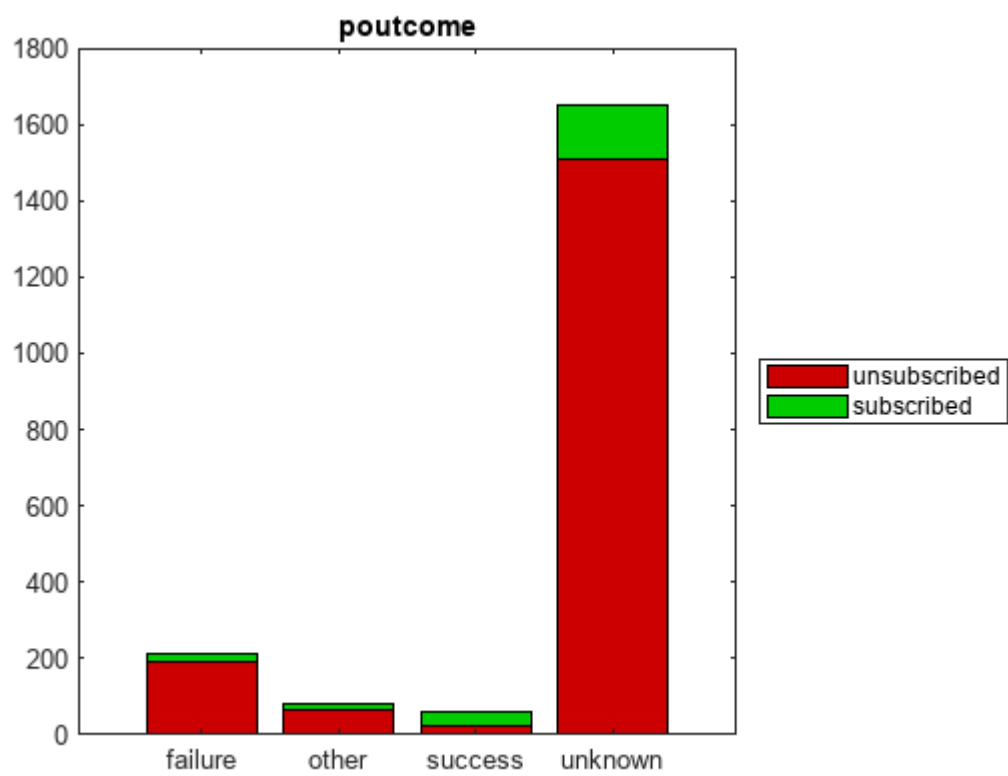
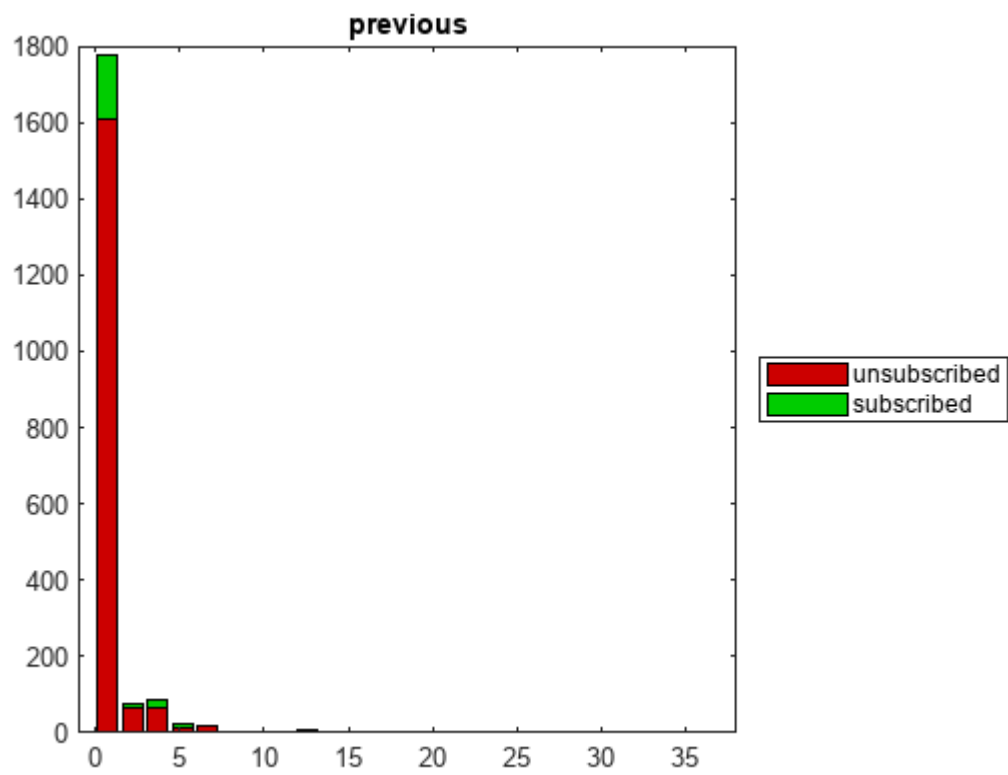


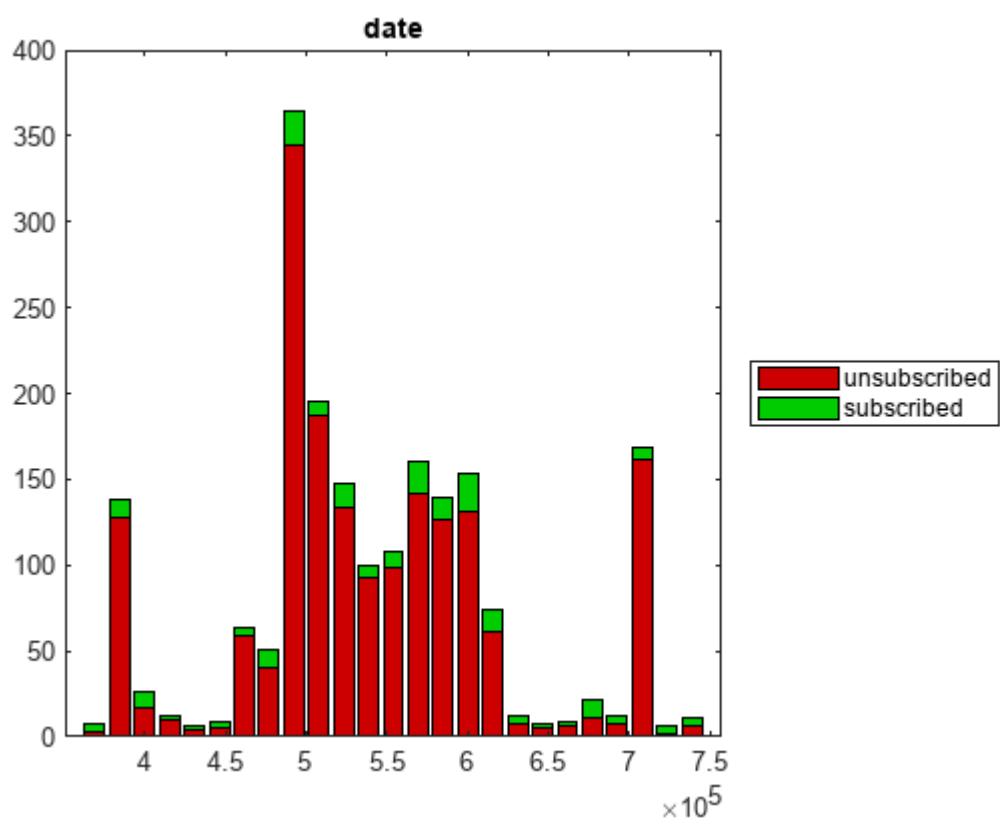
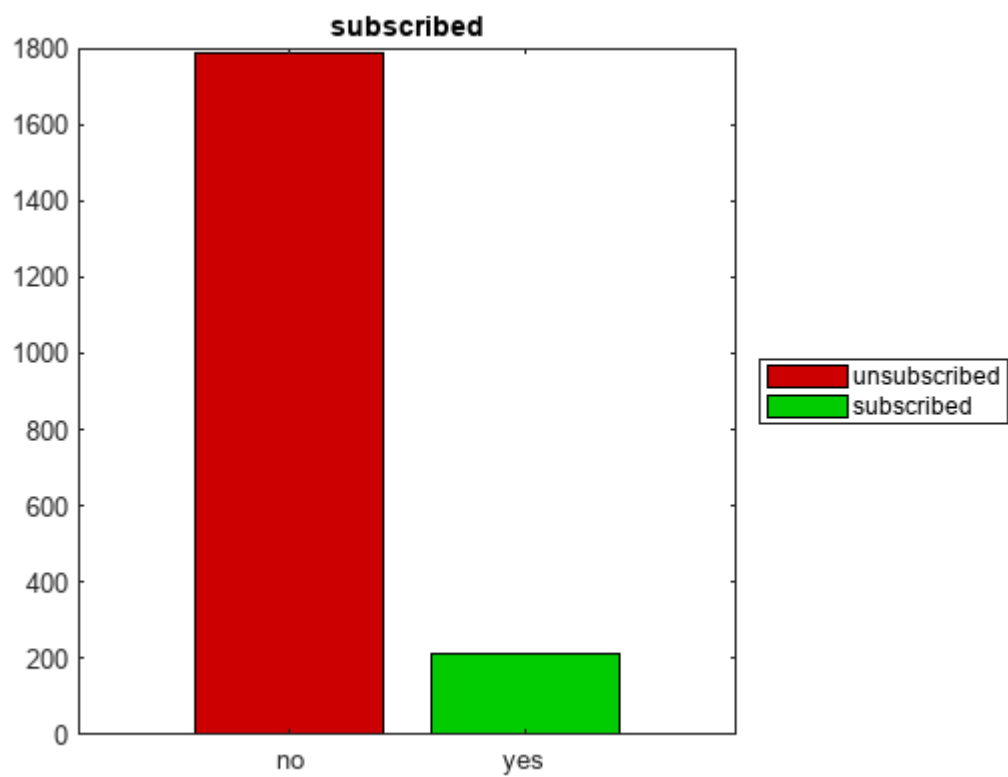












### % of Subscribed

nbins = 25;

```

for col = 1: width(training_set)
    figure

    if iscategorical(table2array(training_set(:,col)))
        Y(1,:) = countcats( table2array(unsubscribed(:,col)) ) ./ ...
            countcats( table2array(training_set(:,col)) );
        Y(2,:) = countcats( table2array( subscribed(:,col)) ) ./ ...
            countcats( table2array(training_set(:,col)) );

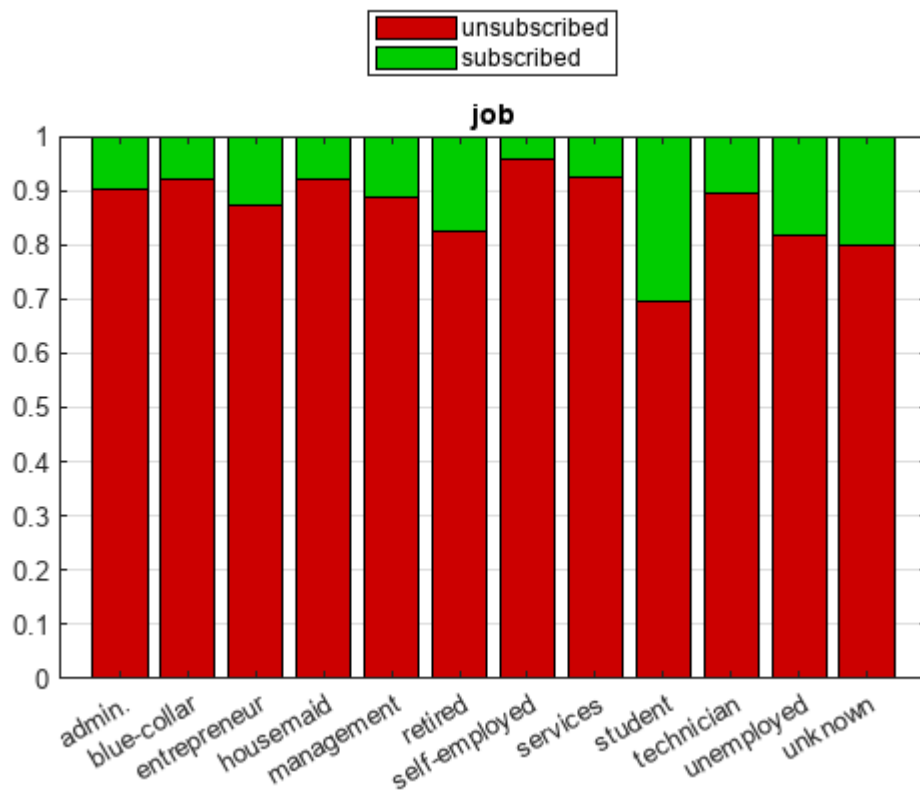
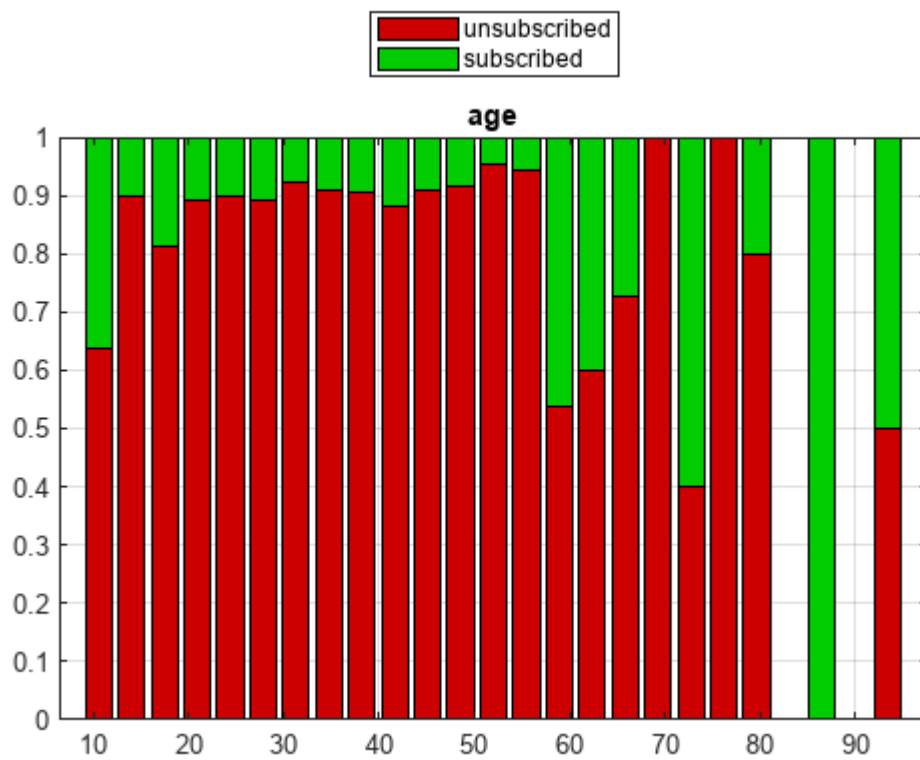
        ba = bar(categorical(categories(table2array(training_set(:,col)))), ...
            Y', 'stacked', 'FaceColor','flat');
        ba(1).CData = [0.8 0 0];
        ba(2).CData = [0 0.8 0];
        clear Y
    else
        hist_edges = linspace(min(table2array(training_set(:,col))), ...
            max(table2array(training_set(:,col))), nbins+1);
        plot_edges = linspace(hist_edges(2)/2, (hist_edges(end-1)+hist_edges(end))/2, nbins);

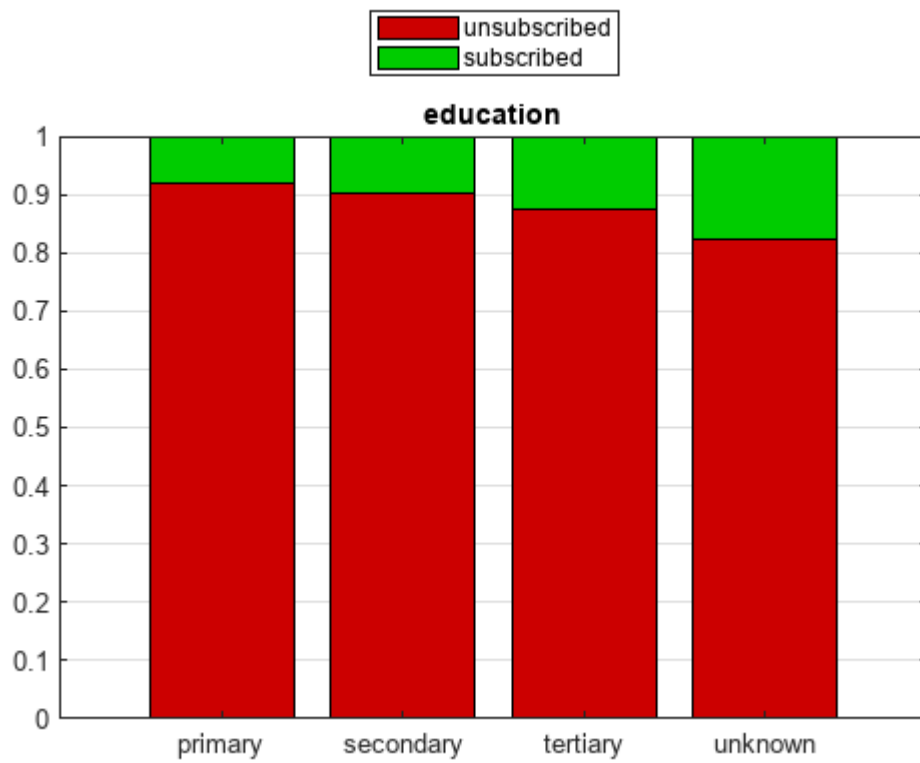
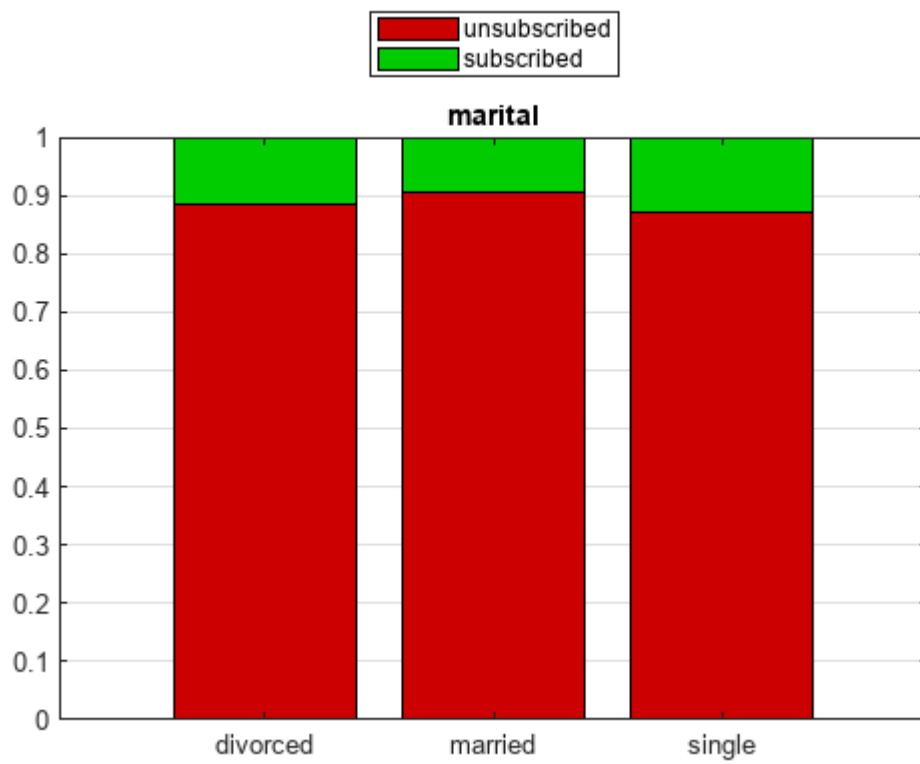
        X(1,:) = histcounts( table2array(unsubscribed(:,col)), hist_edges) ./ ...
            histcounts( table2array(training_set(:,col)), hist_edges);
        X(2,:) = histcounts( table2array( subscribed(:,col)), hist_edges) ./ ...
            histcounts( table2array(training_set(:,col)), hist_edges);

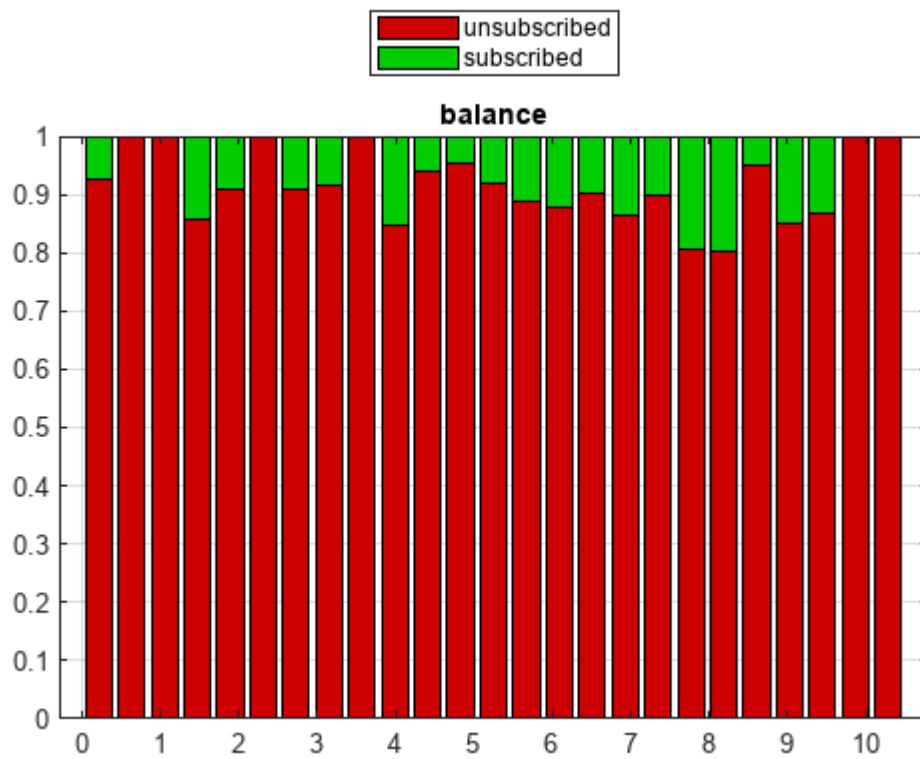
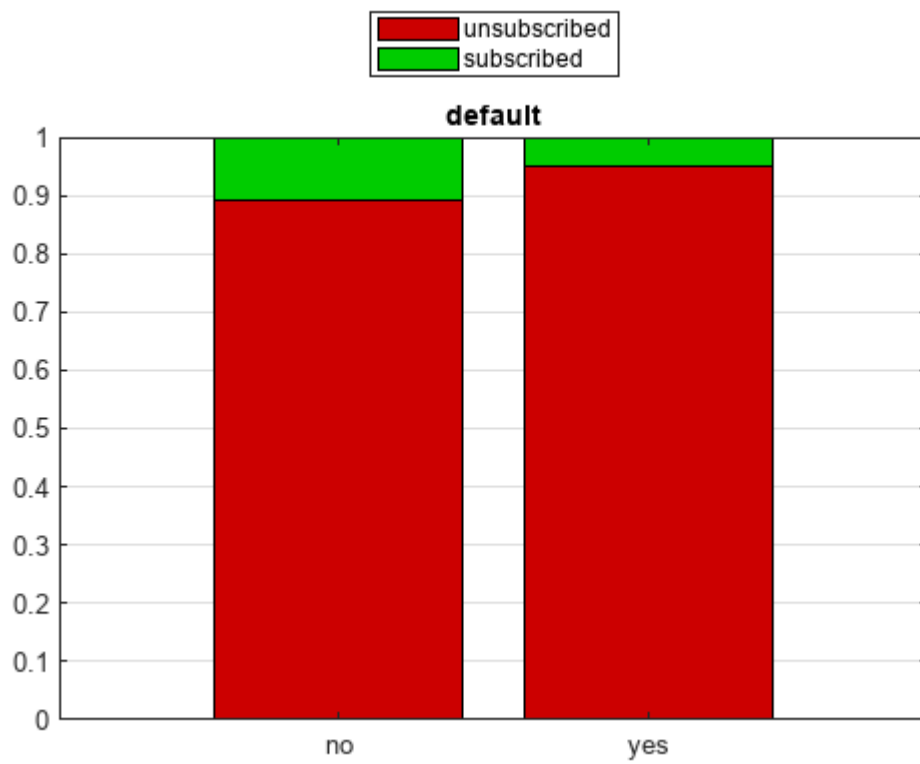
        ba = bar(plot_edges, X', 'stacked', 'FaceColor','flat');
        ba(1).CData = [0.8 0 0];
        ba(2).CData = [0 0.8 0];
    end

    title(training_set.Properties.VariableNames{col})
    if col == 5
        title(' default ')
    end
    legend('unsubscribed','subscribed','Location','northoutside')
    ylim([0 1])
    grid()
    % saveas(gcf, strcat('figures/Proportion_', test_set.Properties.VariableNames{col}, '.png'))
end

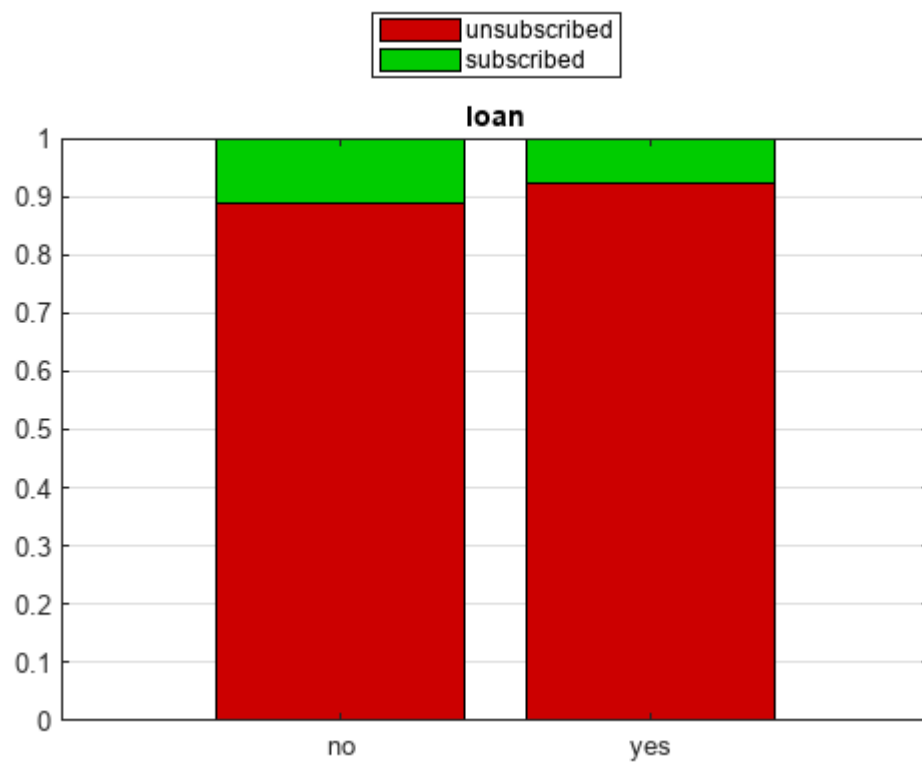
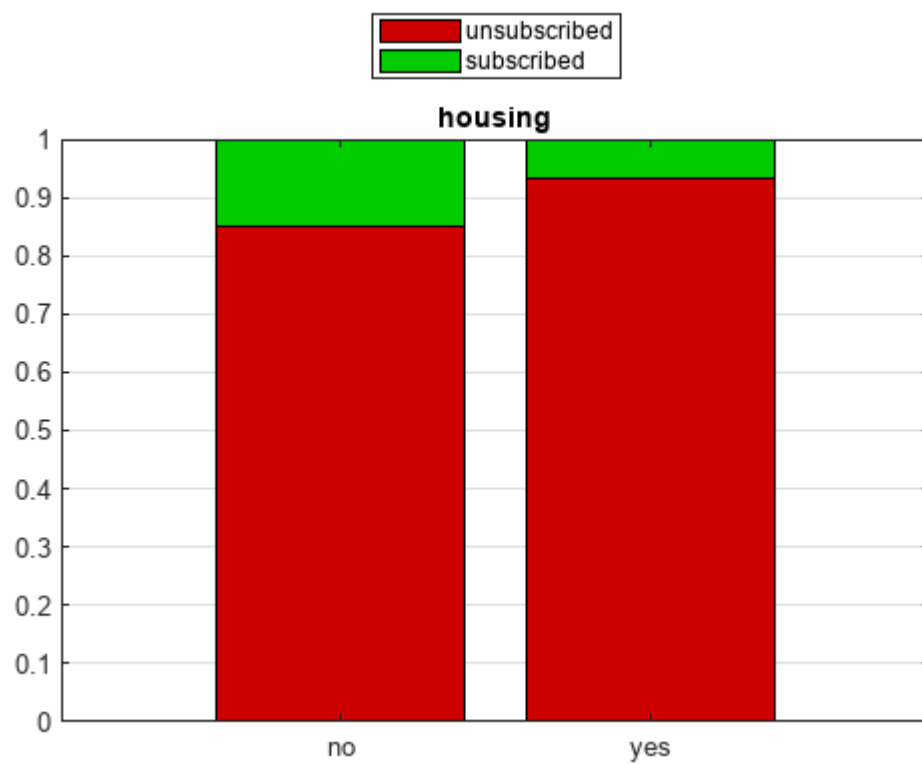
```

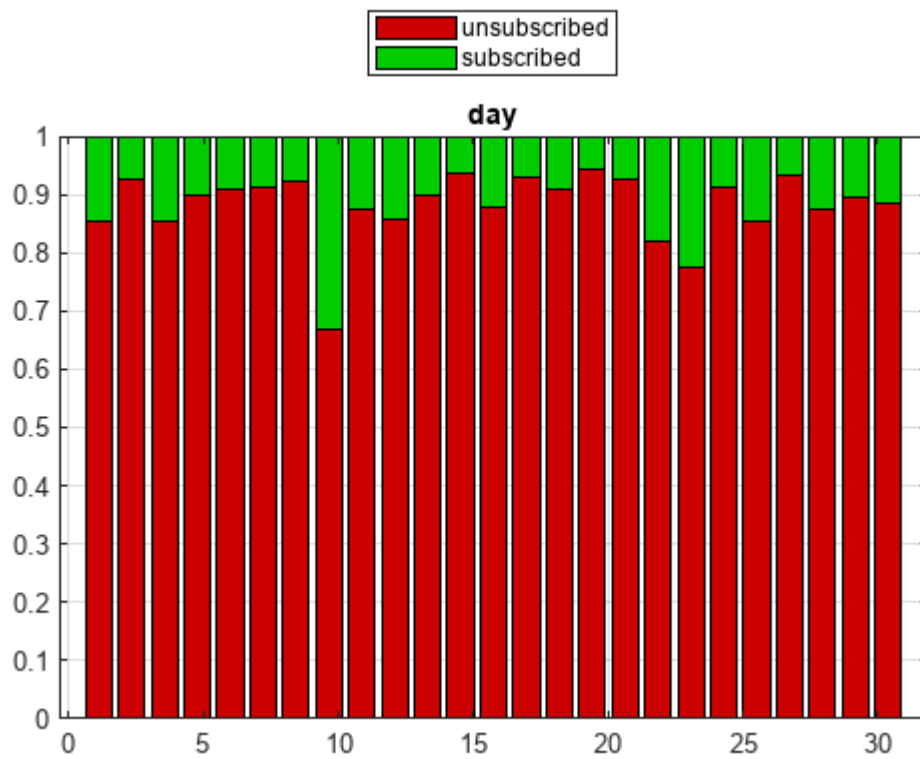
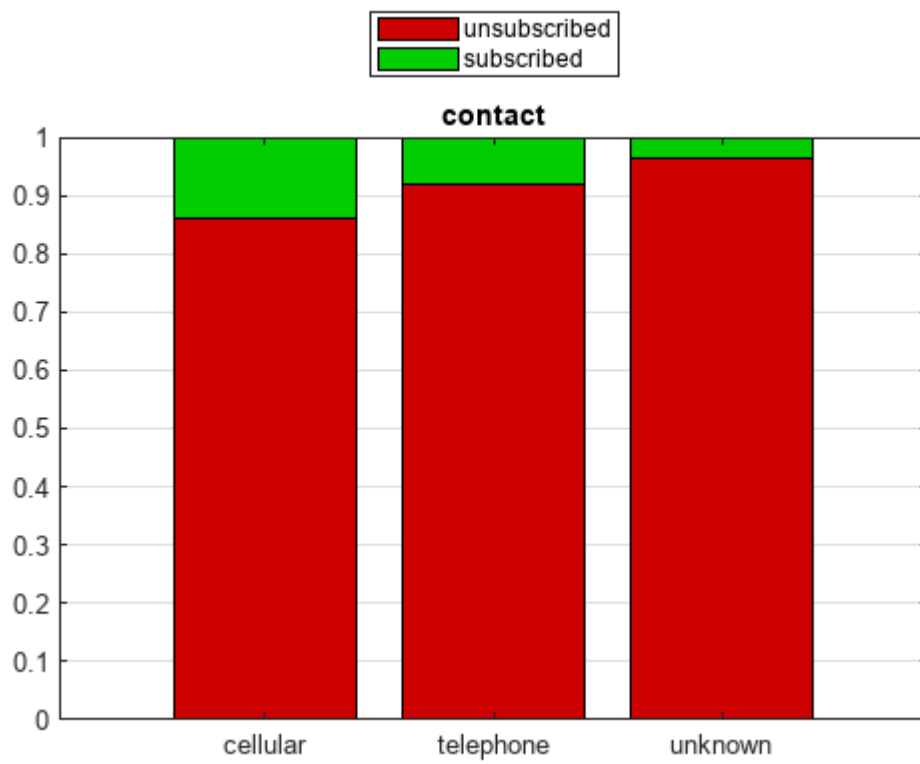


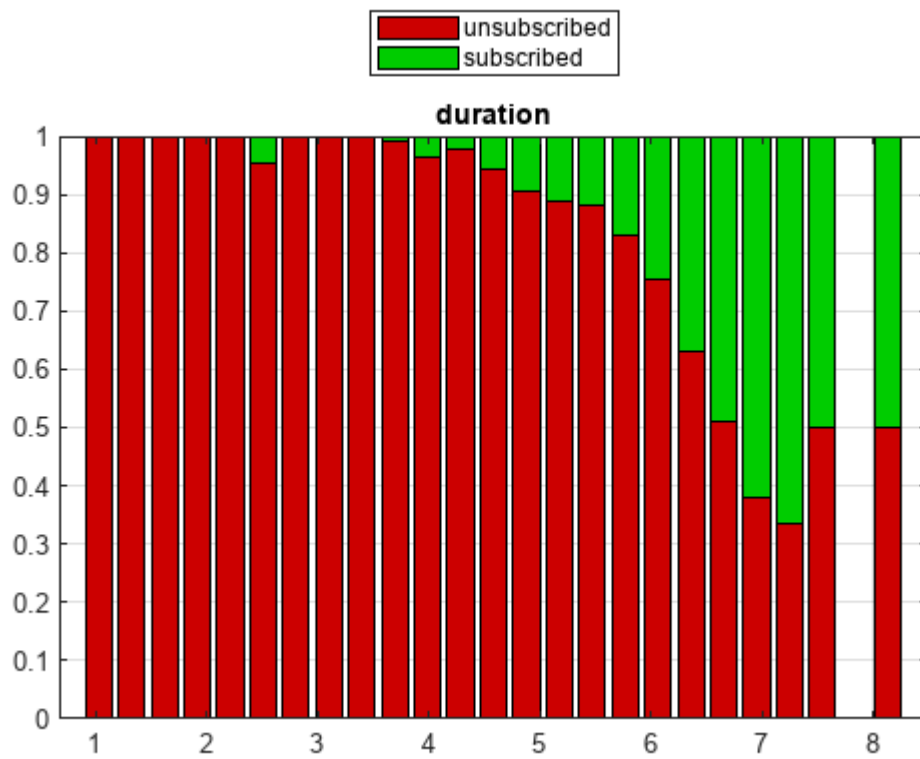
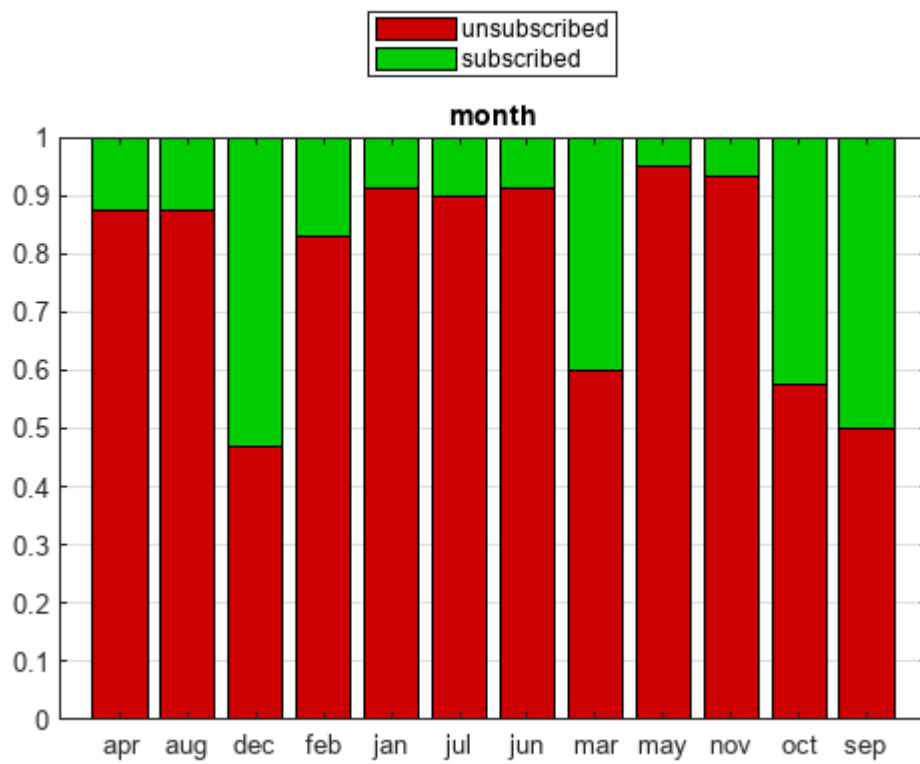


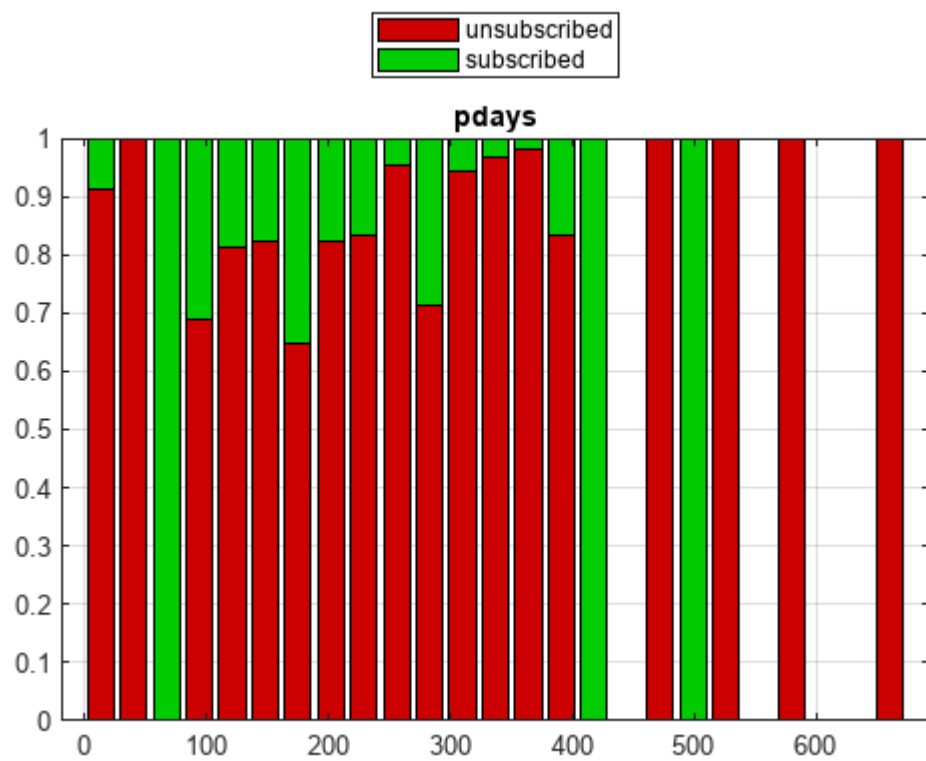
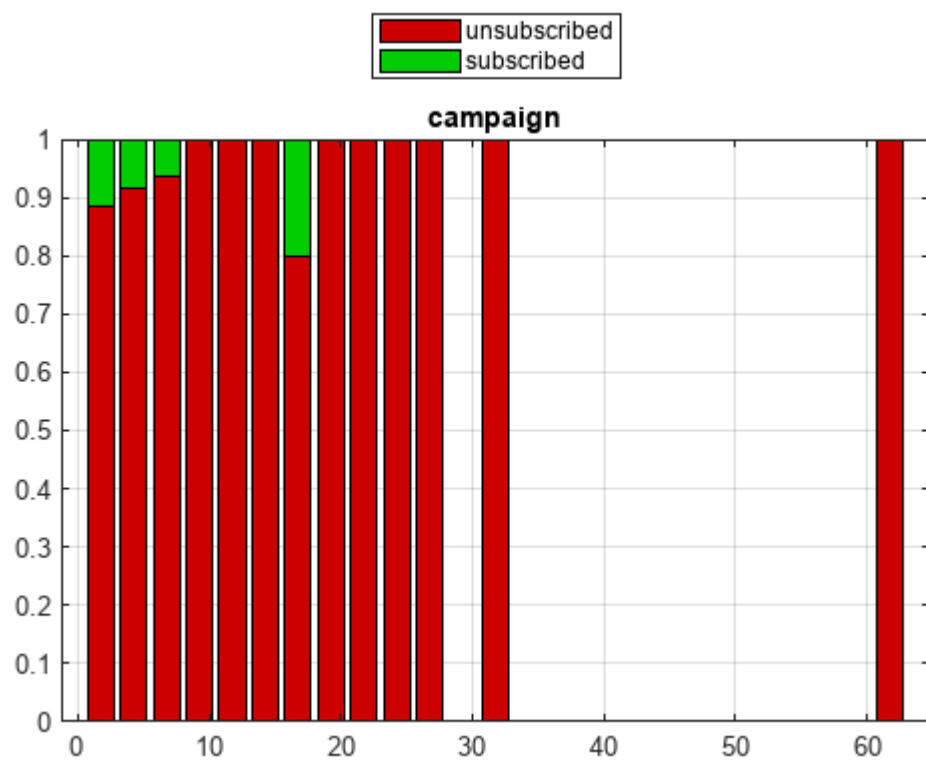


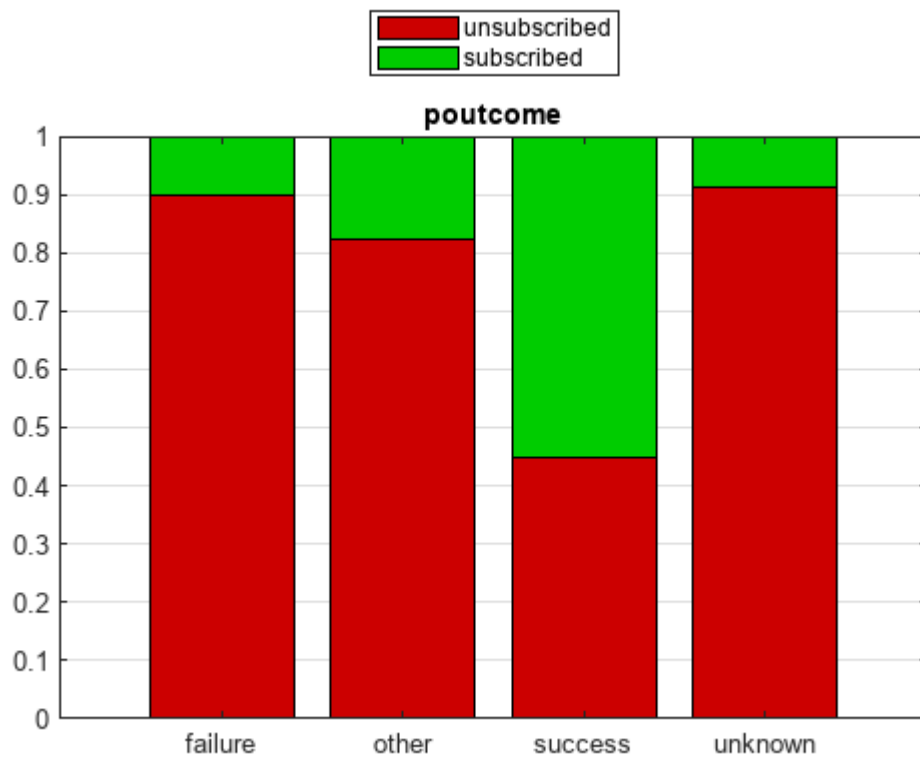
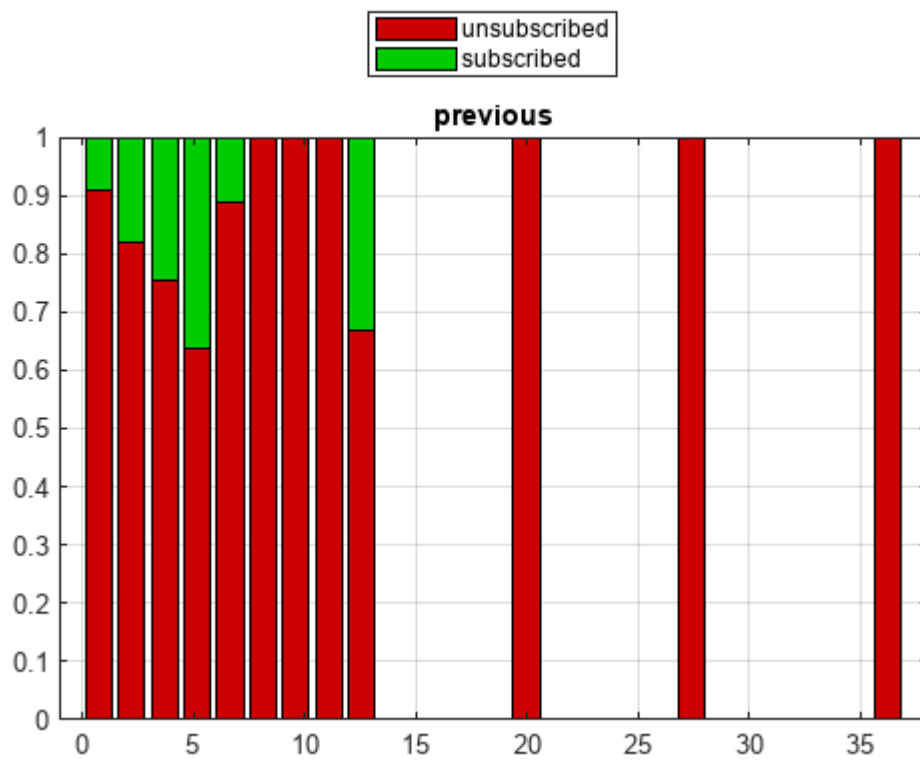


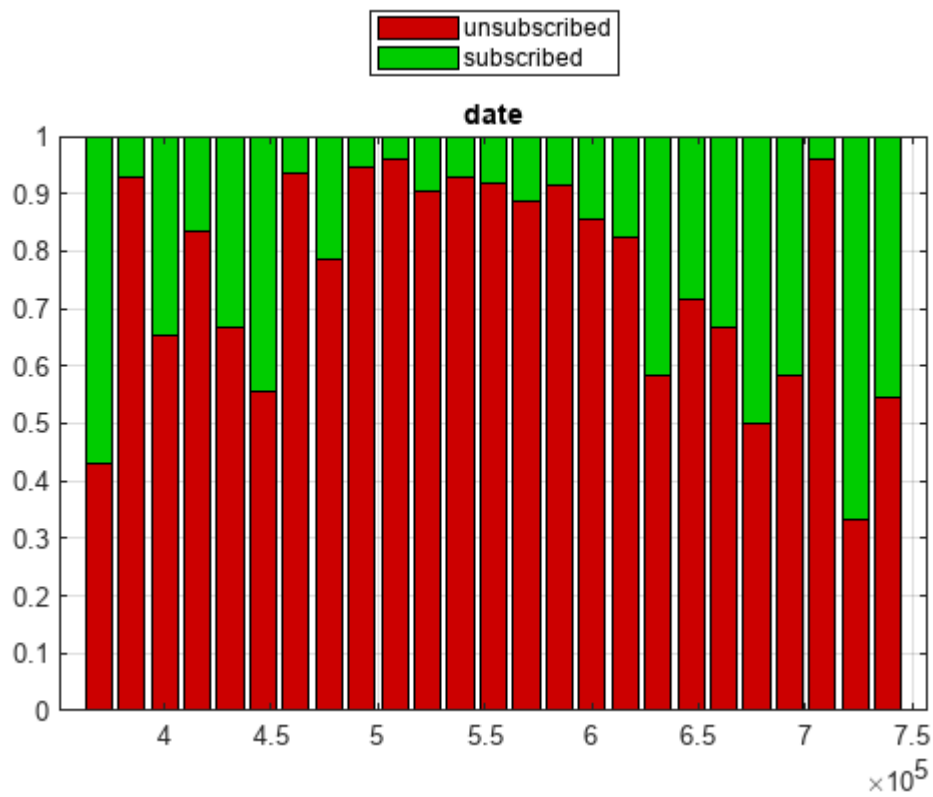
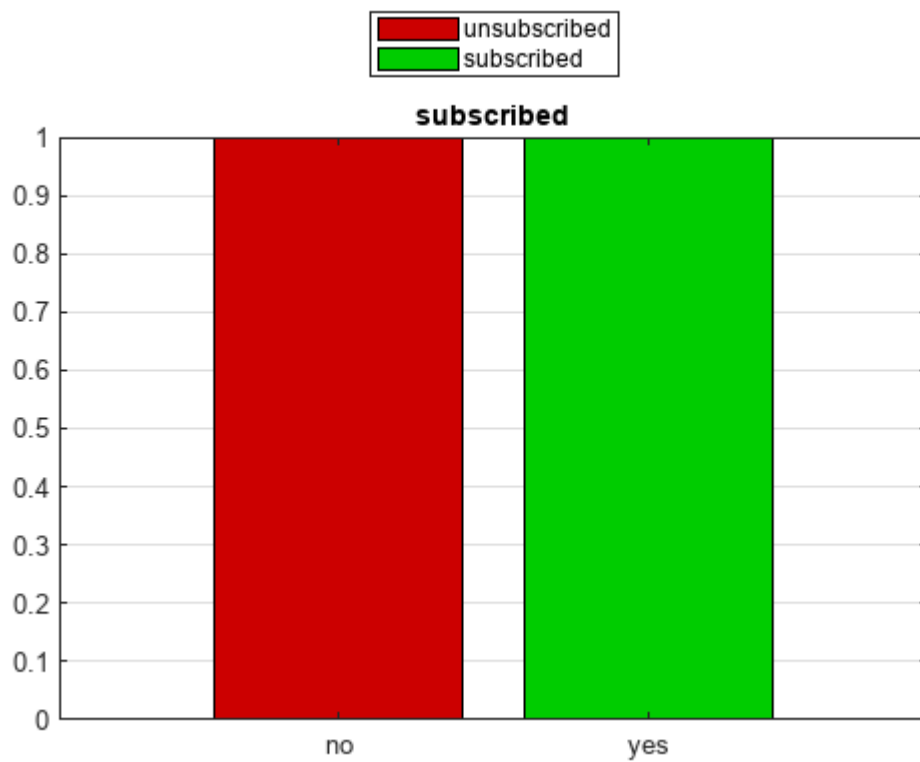












## Histograms of test set

```
nbins = 25;
```

```

for col = 1: width(test_set)
    figure

    if iscategorical(table2array(test_set(:,col)))
        Y = countcats( table2array(test_set(:,col)) );

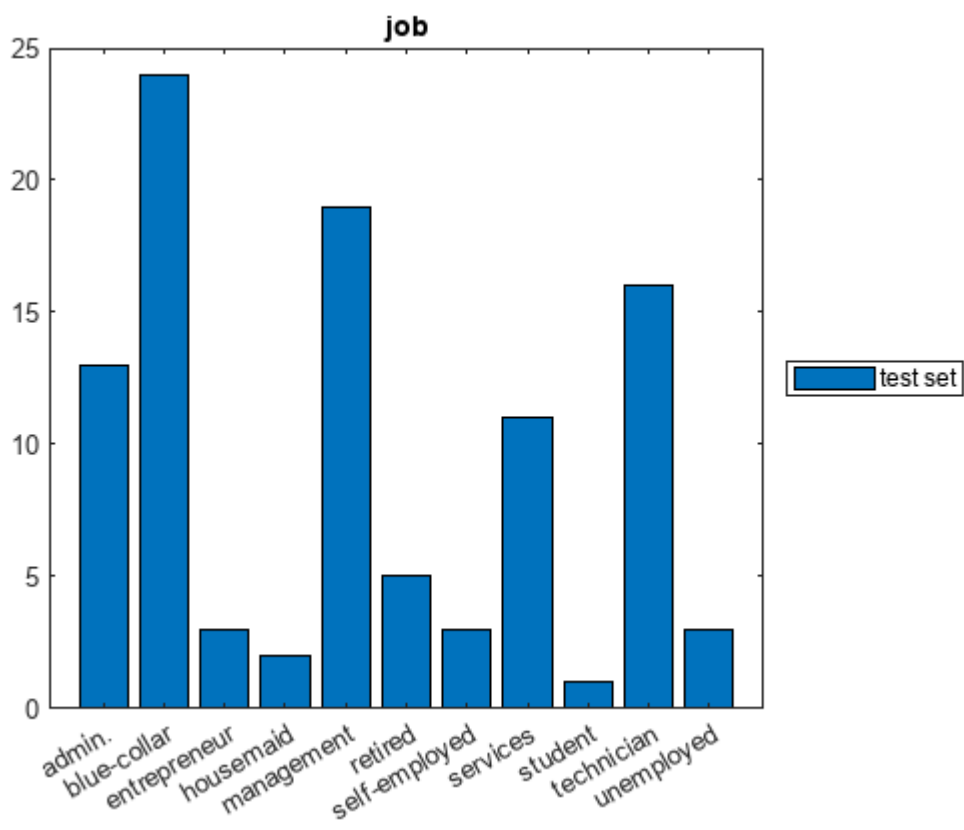
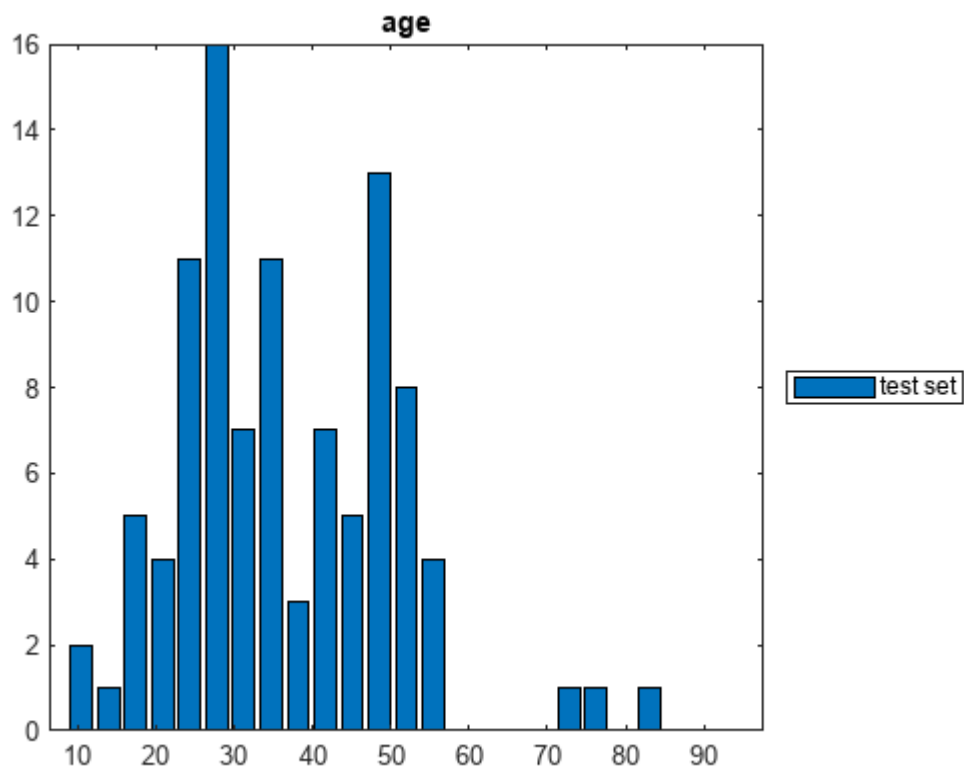
        ba = bar(categorical(categories(table2array(test_set(:,col)))), ...
            Y', 'FaceColor','flat');
        clear Y
    else
        hist_edges = linspace(min(table2array(training_set(:,col))), ...
            max(table2array(training_set(:,col))), nbins+1);
        plot_edges = linspace(hist_edges(2)/2, (hist_edges(end-1)+hist_edges(end))/2, nbins);

        X = histcounts( table2array(test_set(:,col)), hist_edges);

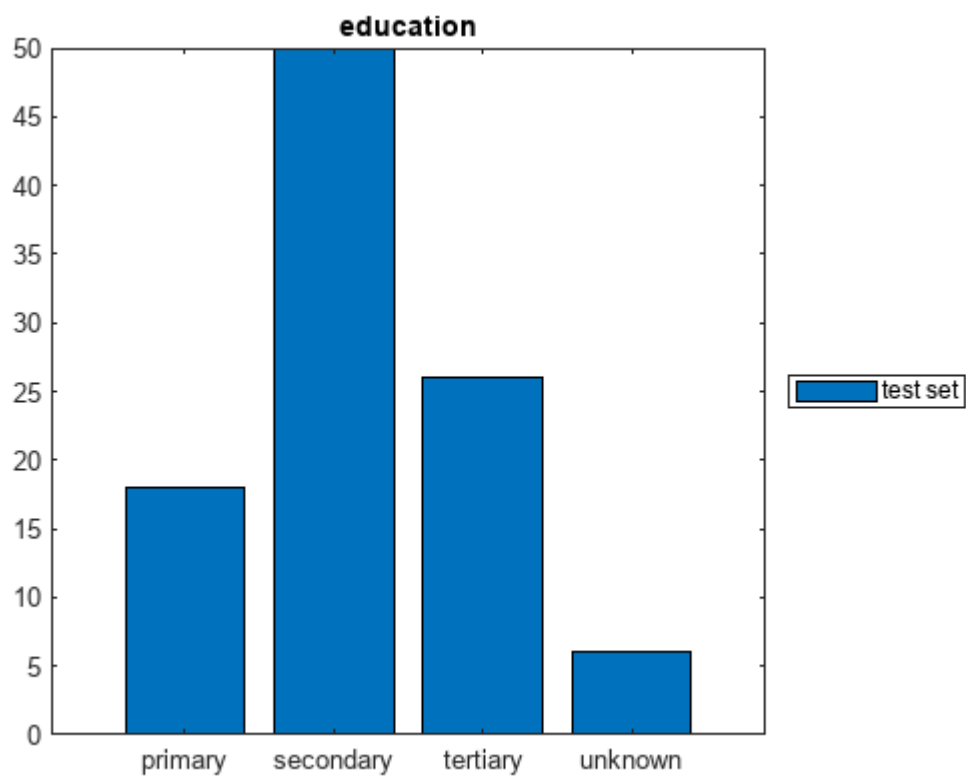
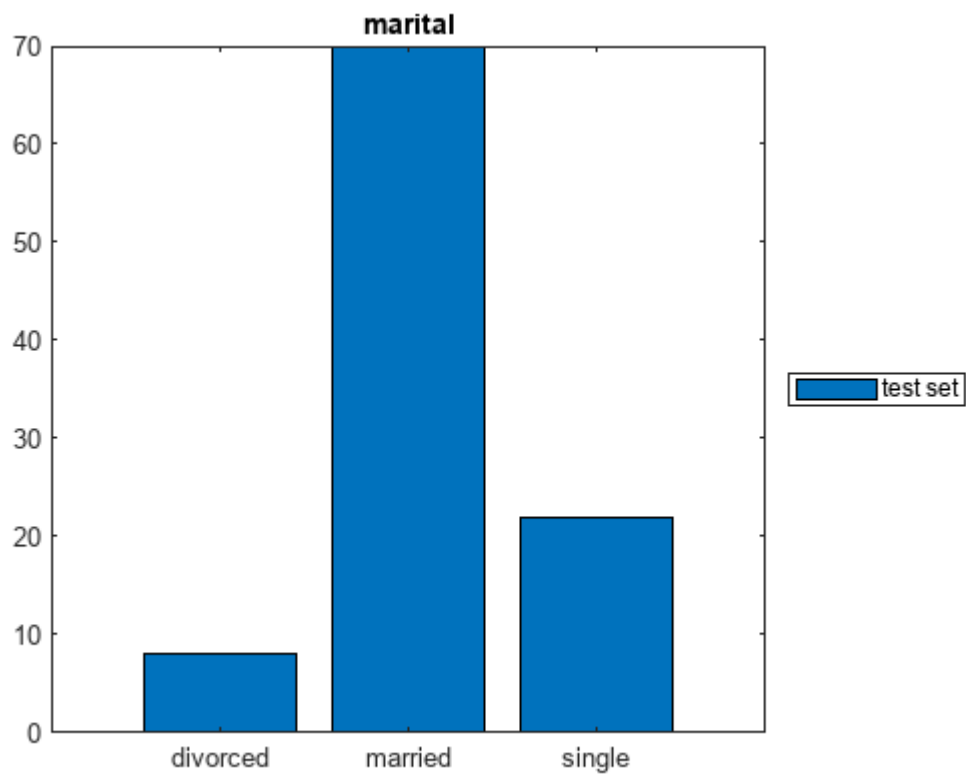
        ba = bar(plot_edges, X', 'FaceColor','flat');
    end

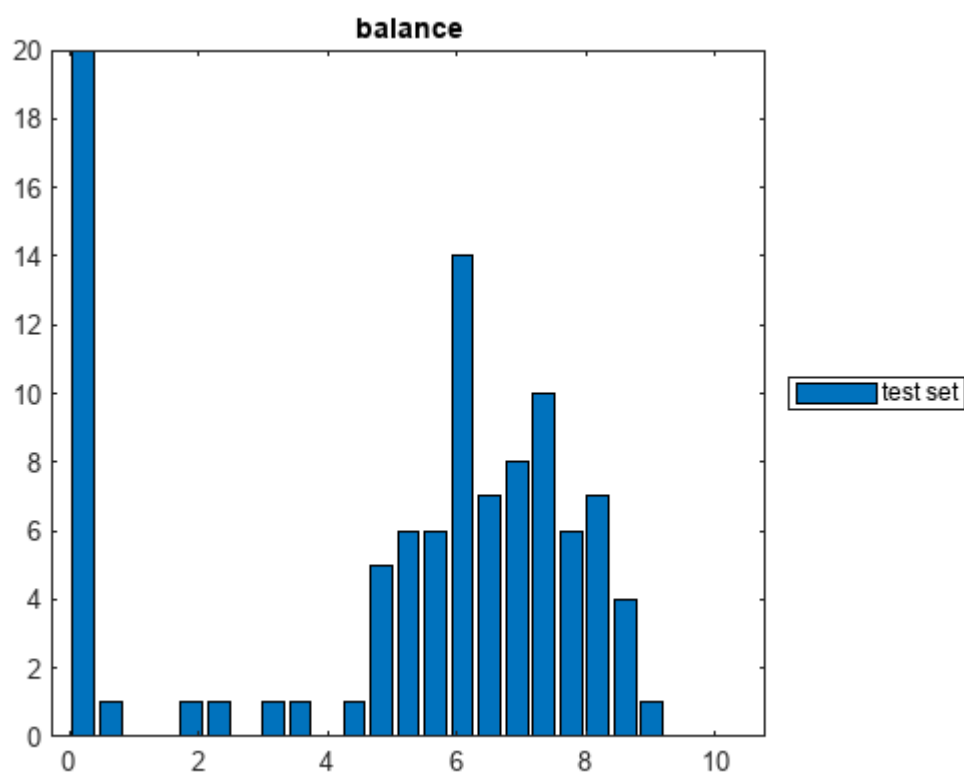
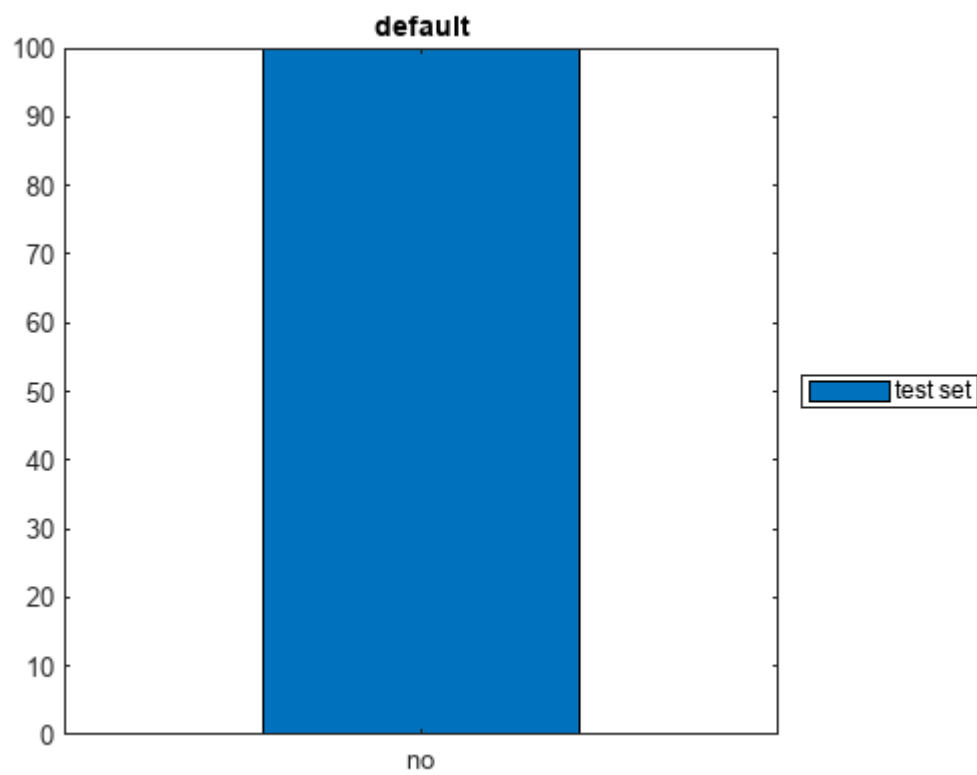
    title(test_set.Properties.VariableNames{col})
    if col == 5
        title(' default ')
    end
    legend('test set','Location','eastoutside')
%     saveas(gcf, strcat('figures/Hist_test_', test_set.Properties.VariableNames{col}, '.png'))
end

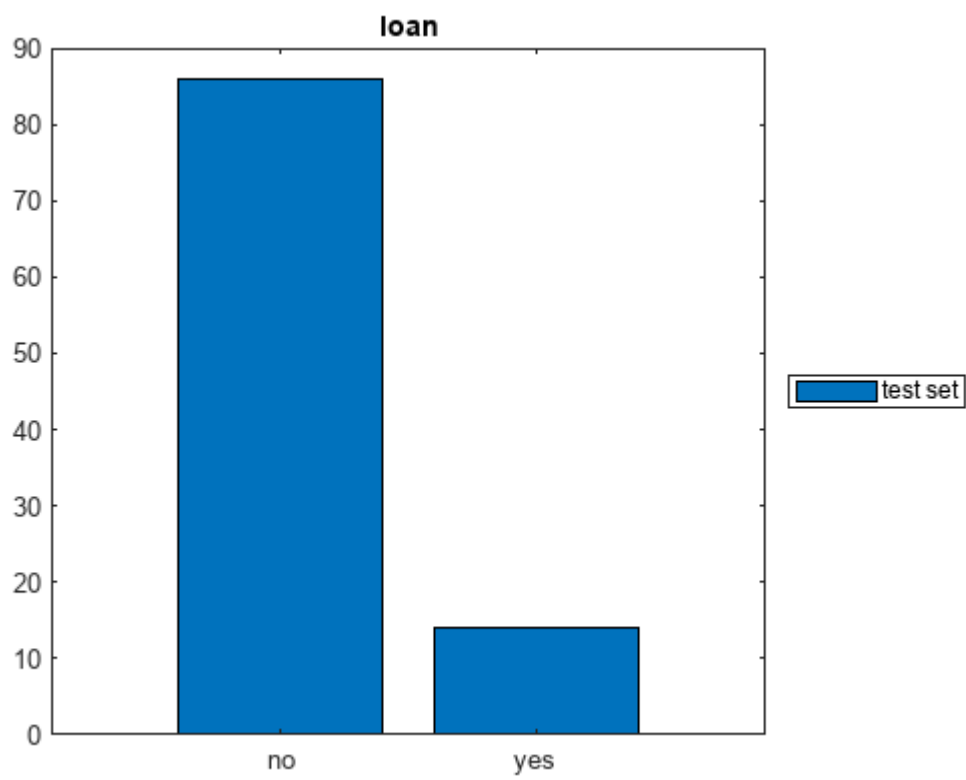
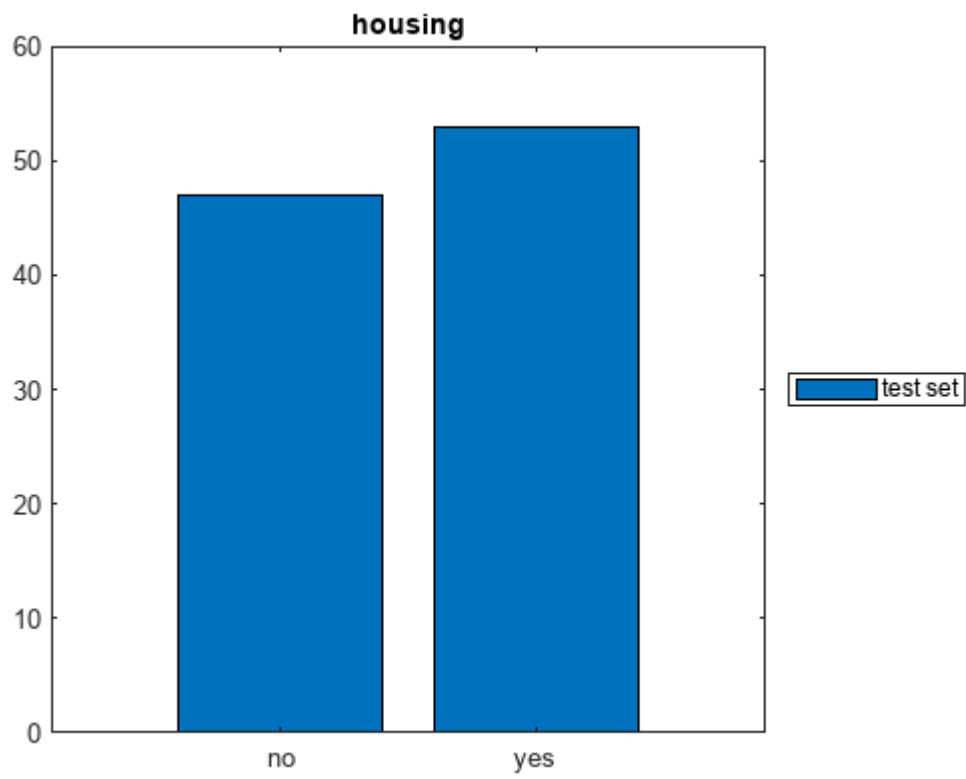
```

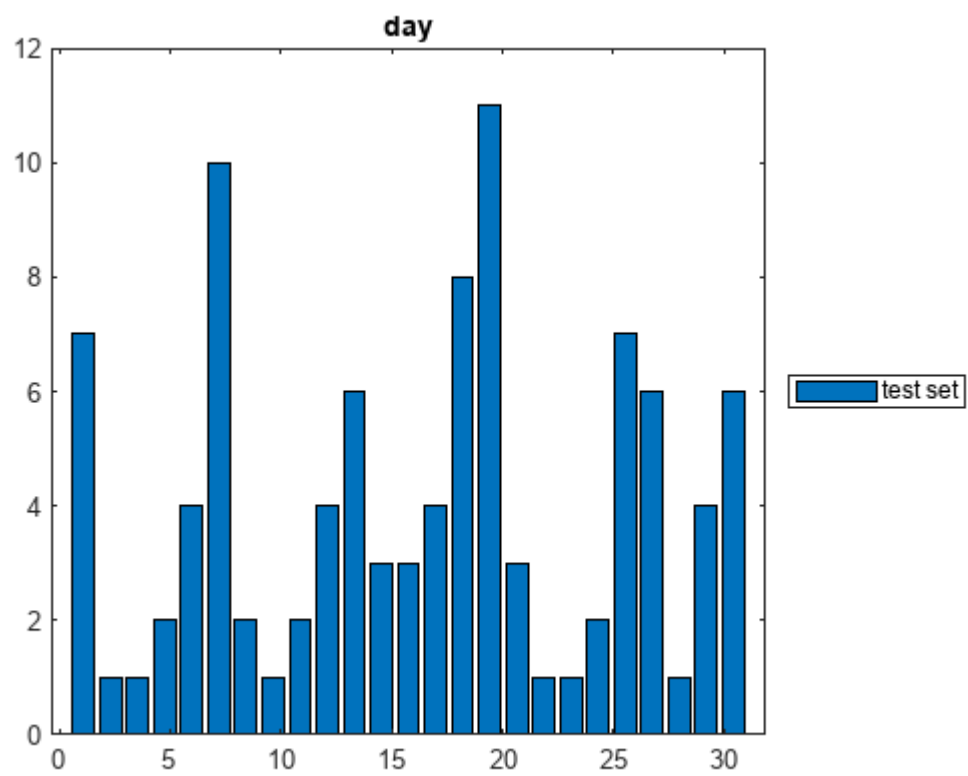
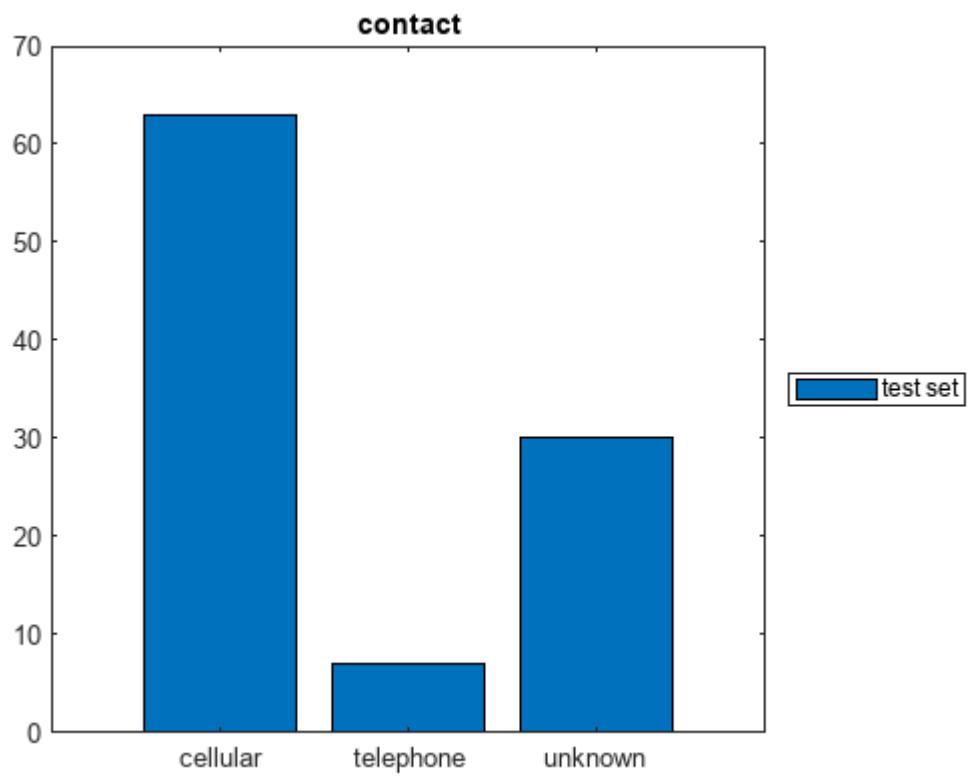


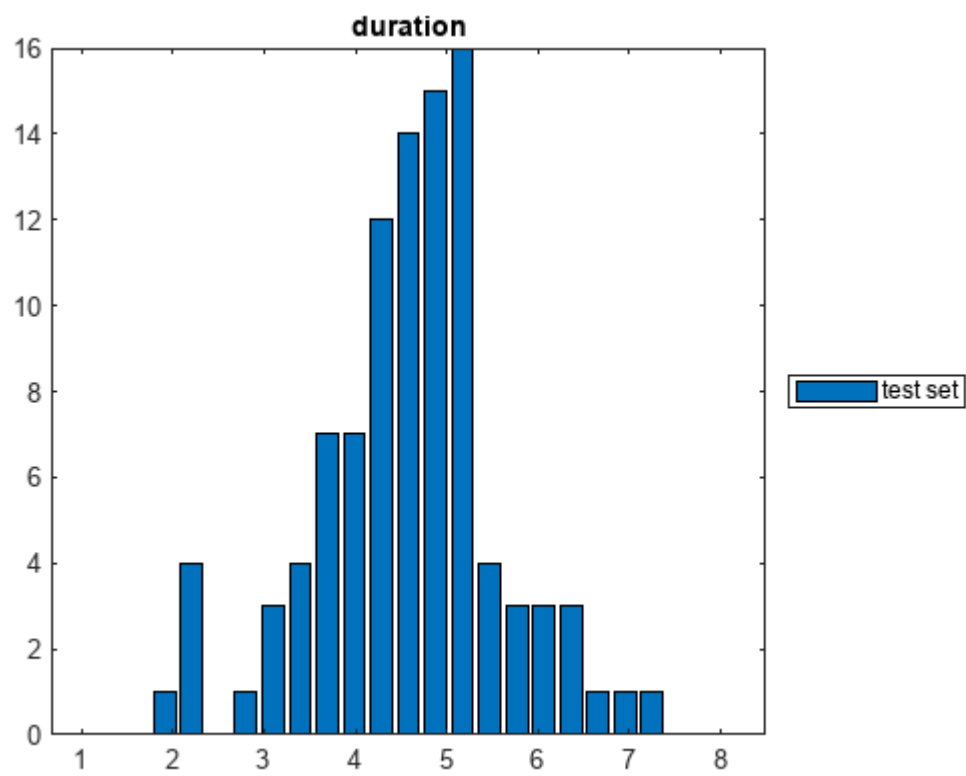
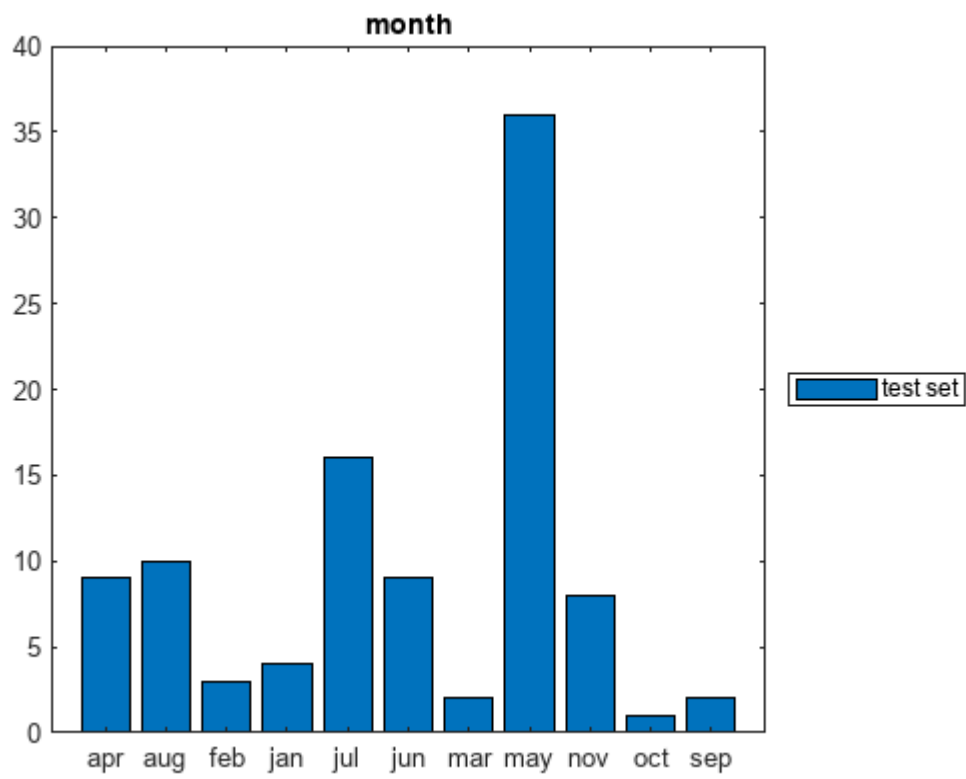


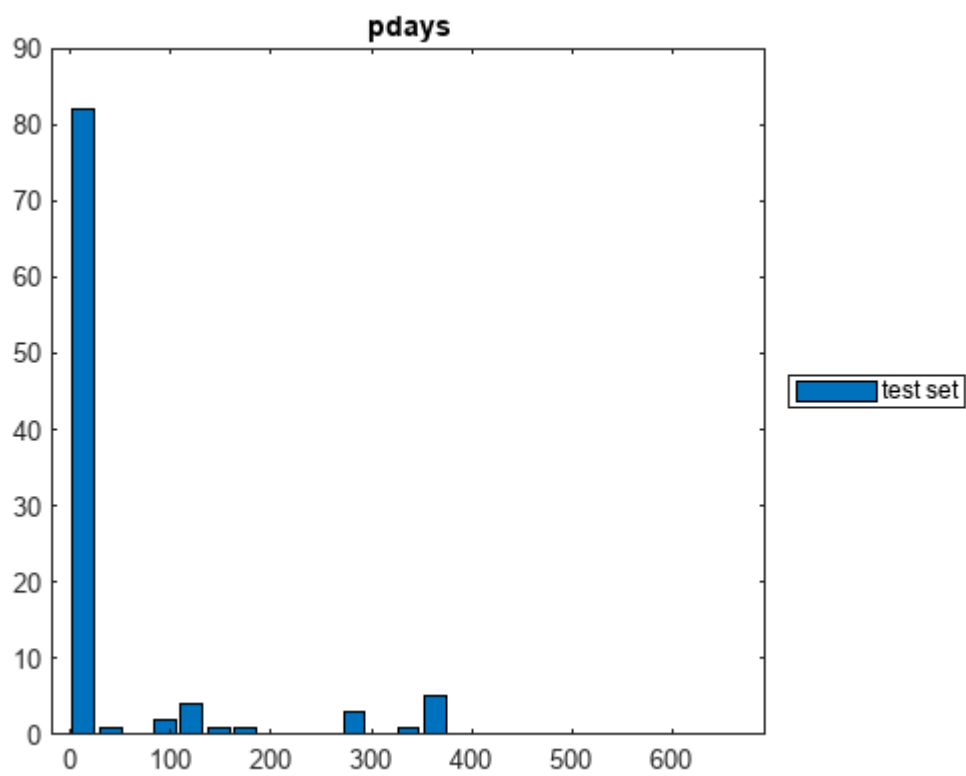
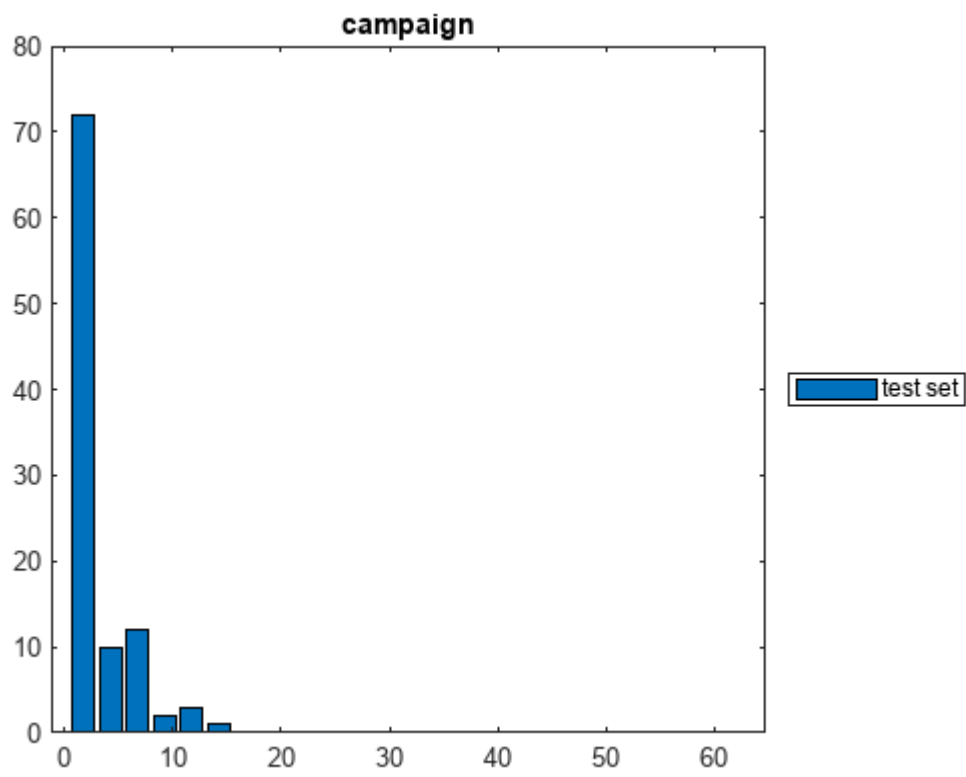


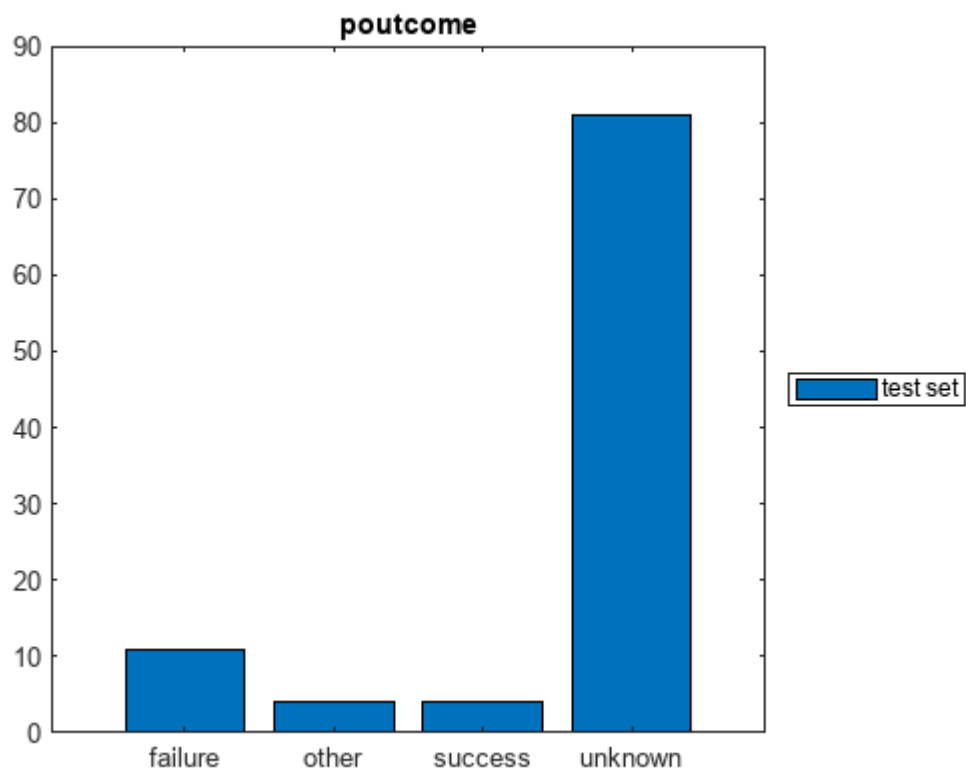
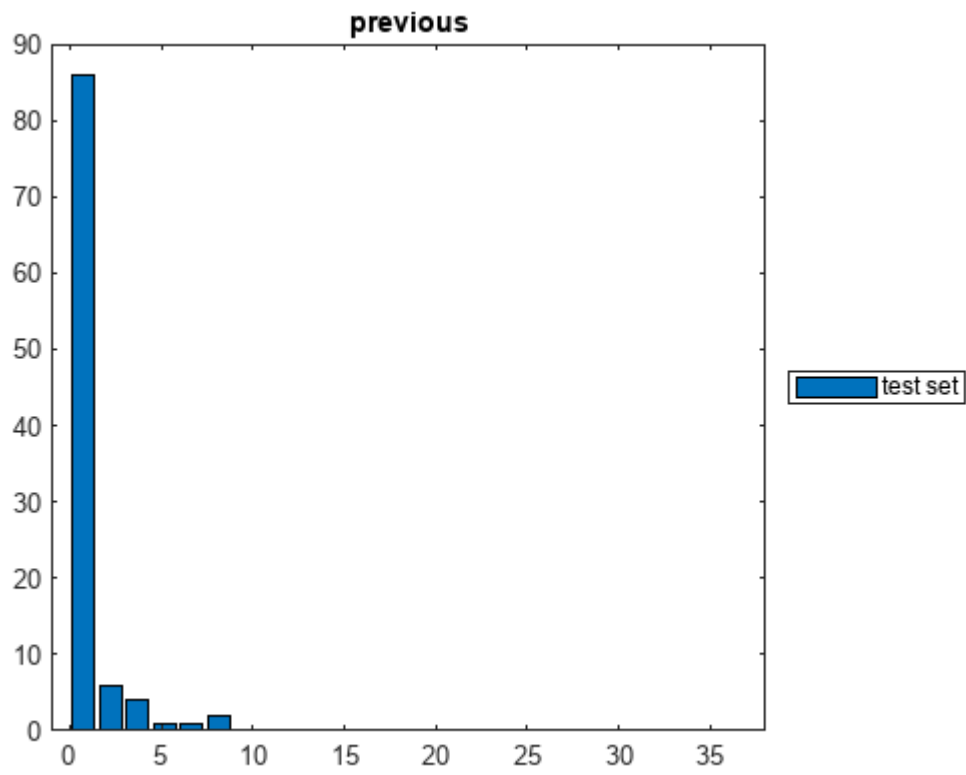


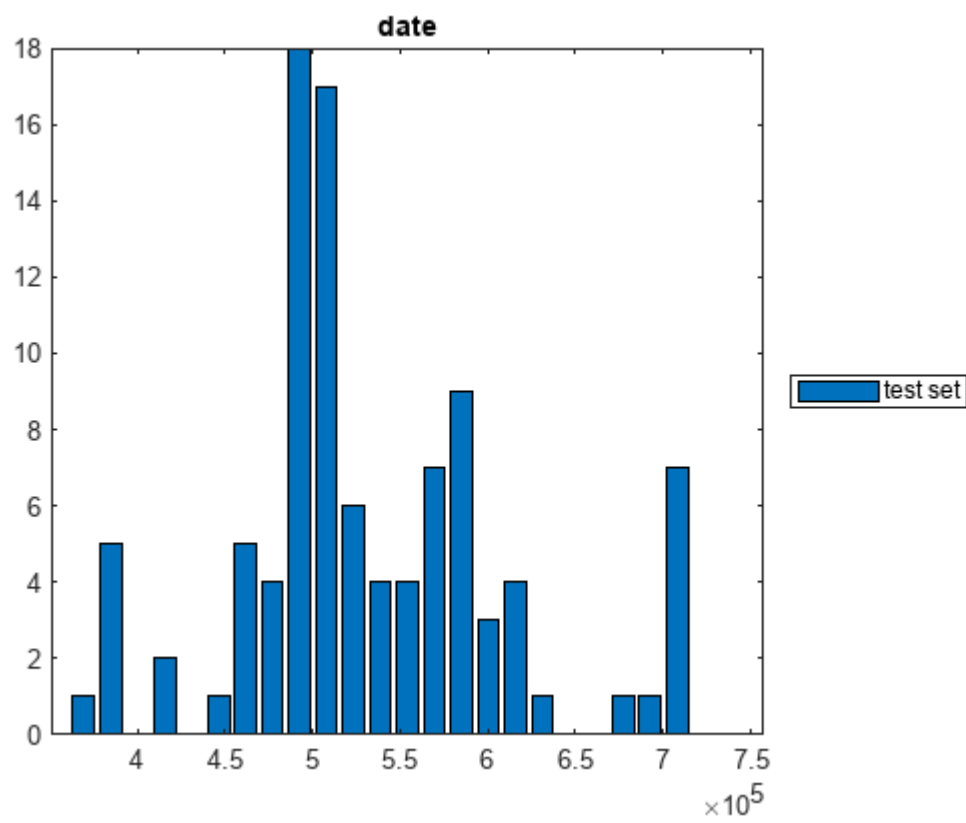
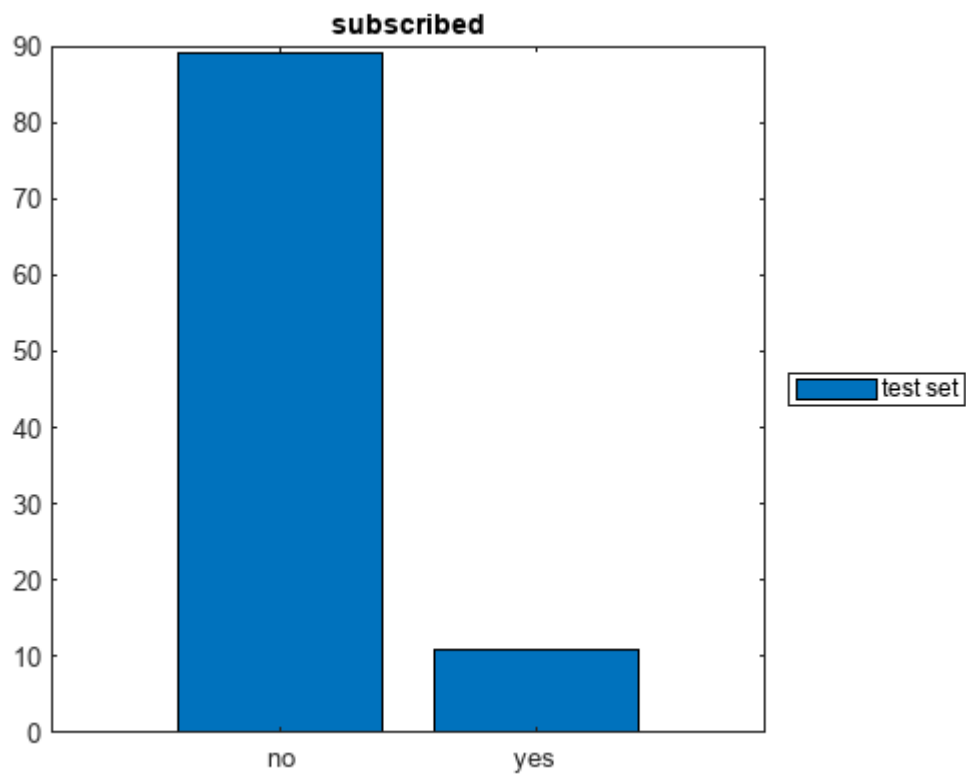












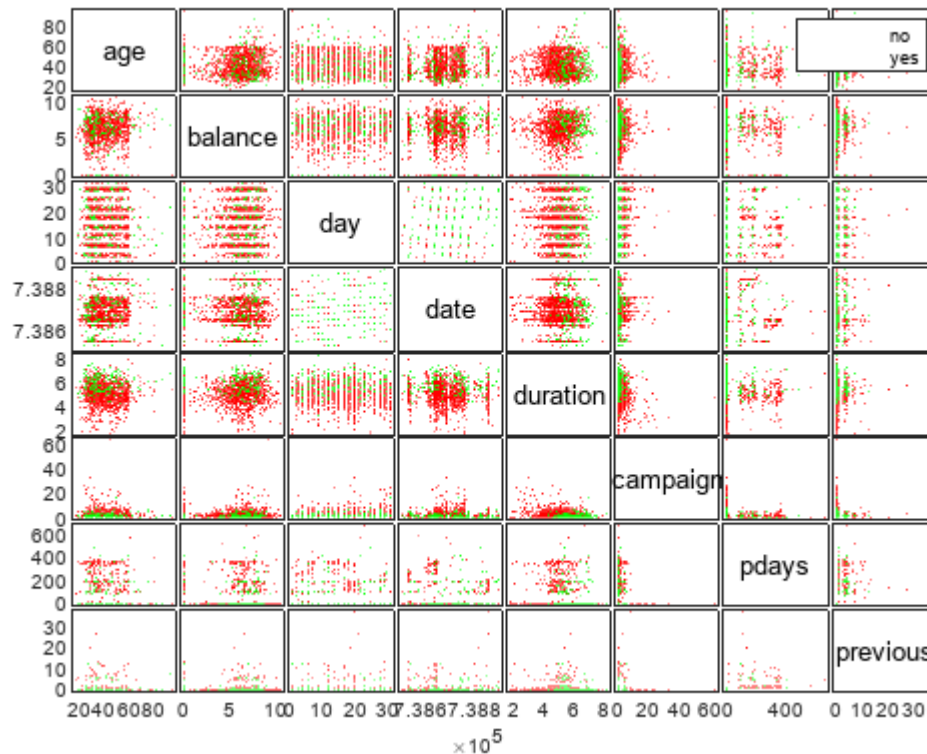


## Gplot of non-categorical variables

```
X_ = [training_set.age training_set.balance training_set.day training_set.date, ...
      training_set.duration training_set.campaign training_set.pdays training_set.previous];

xnames = {'age', 'balance', 'day', 'date', 'duration', 'campaign', 'pdays', 'previous'};

figure
[h, ax] = gplotmatrix( X_,[],training_set.subscribed,['r','g'],[],[],[],'variable',xnames);
```



```
% saveas(gcf, strcat('figures/gplot_training.png'))
```

## Classification model: decision tree with all variables

Train a decision tree with the provided dataset, capable of describing the subscription status of the client based on the other variables

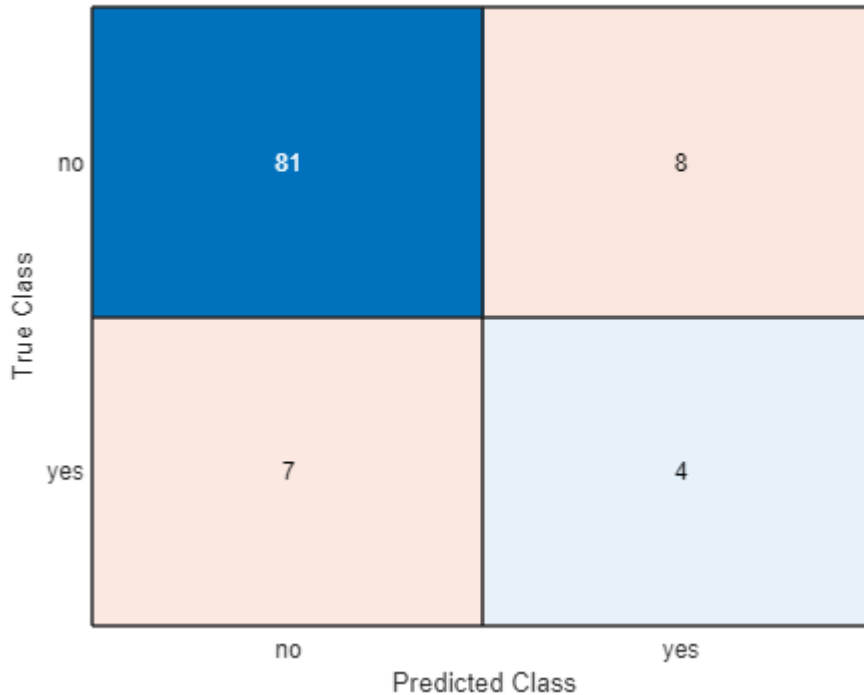
```
tree = fitctree(training_set, 'subscribed')
```

```
tree =
  ClassificationTree
    PredictorNames: {'age' 'job' 'marital' 'education' 'default' 'balance' 'housing' 'loan' 'contact'
    ResponseName: 'subscribed'
    CategoricalPredictors: [2 3 4 5 7 8 9 11 16]
    ClassNames: [no yes]
```

ScoreTransform: 'none'  
NumObservations: 2000

Properties, Methods

```
[predictions_Tfull, scores_Tfull] = predict(tree,test_set);  
confusionchart(test_set.subscribed,predictions_Tfull)
```



```
% saveas(gcf,strcat('figures/Confussion_Tfull.png')) )  
CrossVal_Tfull = kfoldLoss(crossval(tree))
```

CrossVal\_Tfull = 0.1280

```
Accuracy_Tfull = accuracy(test_set.subscribed, predictions_Tfull, 'yes', 'no')
```

Accuracy\_Tfull = 0.8500

```
Recall_Tfull = recall(test_set.subscribed, predictions_Tfull, 'yes', 'no')
```

Recall\_Tfull = 0.3636

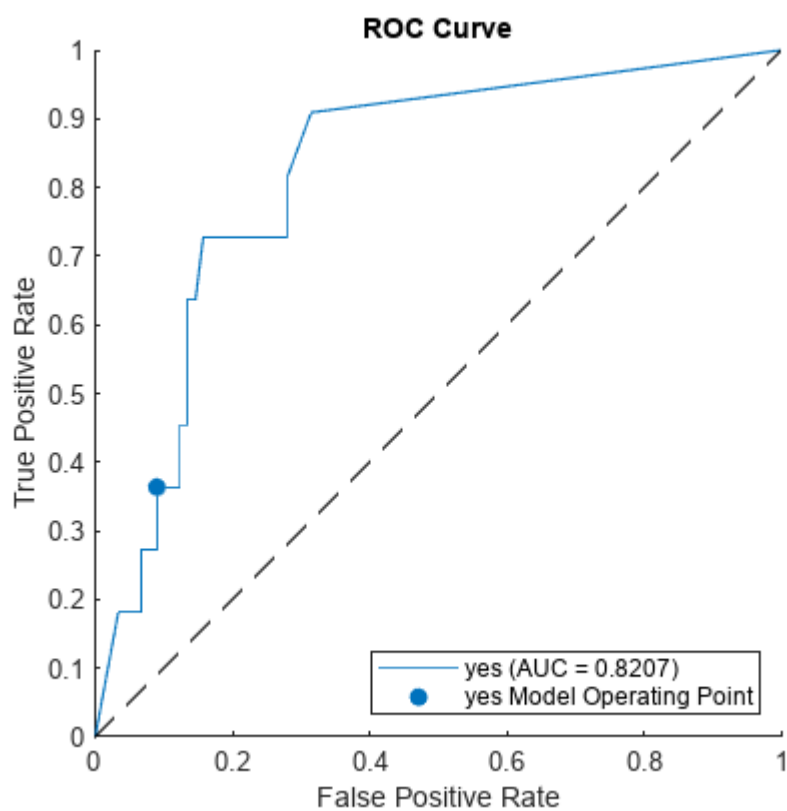
```
Precision_Tfull = precision(test_set.subscribed, predictions_Tfull, 'yes', 'no')
```

Precision\_Tfull = 0.3333

```
rocObj = rocmetrics(test_set.subscribed, scores_Tfull, tree.ClassNames);  
AUC_Tfull = rocObj.AUC(end)
```

AUC\_Tfull = 0.8207

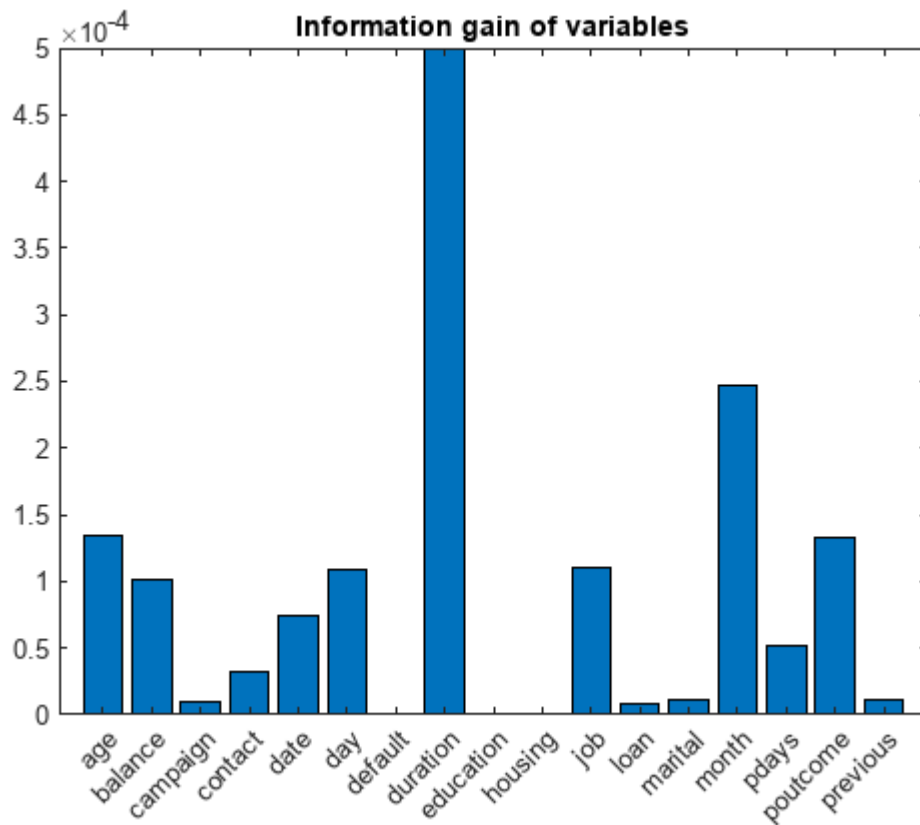
```
plot(rocObj,ClassNames='yes')
xlim([0,1])
ylim([0,1])
```



```
% saveas(gcf, strcat('figures/ROC_Tfull.png'))
```

## Reduce number of variables

```
bar(categorical([test_set.Properties.VariableNames(1:end-2),test_set.Properties.VariableNames(end)],
    predictorImportance(tree))
title('Information gain of variables')
```



```
% saveas(gcf, strcat('figures/Vars_Tfull.png')) )
```

```
tree_red = fitctree(removevars(training_set, {'default', 'education', 'housing', 'loan', 'marital',
```

```
tree_red =
  ClassificationTree
    PredictorNames: {'age' 'job' 'balance' 'contact' 'day' 'month' 'duration' 'pdays' 'poutcome'}
    ResponseName: 'subscribed'
    CategoricalPredictors: [2 4 6 9]
    ClassNames: [no yes]
    ScoreTransform: 'none'
    NumObservations: 2000
```

Properties, Methods

```
[predictions_Tred, scores_Tred] = predict(tree_red, test_set);
confusionchart(test_set.subscribed, predictions_Tred)
```

True Class	no	82	7
	yes	7	4
		no	yes
		Predicted Class	

```
% saveas(gcf, strcat('figures/Confussion_Tred.png') )
CrossVal_Tred = kfoldLoss(crossval(tree_red))
```

```
CrossVal_Tred = 0.1270
```

```
Accuracy_Tred = accuracy(test_set.subscribed, predictions_Tred, 'yes', 'no')
```

```
Accuracy_Tred = 0.8600
```

```
Recall_Tred = recall(test_set.subscribed, predictions_Tred, 'yes', 'no')
```

```
Recall_Tred = 0.3636
```

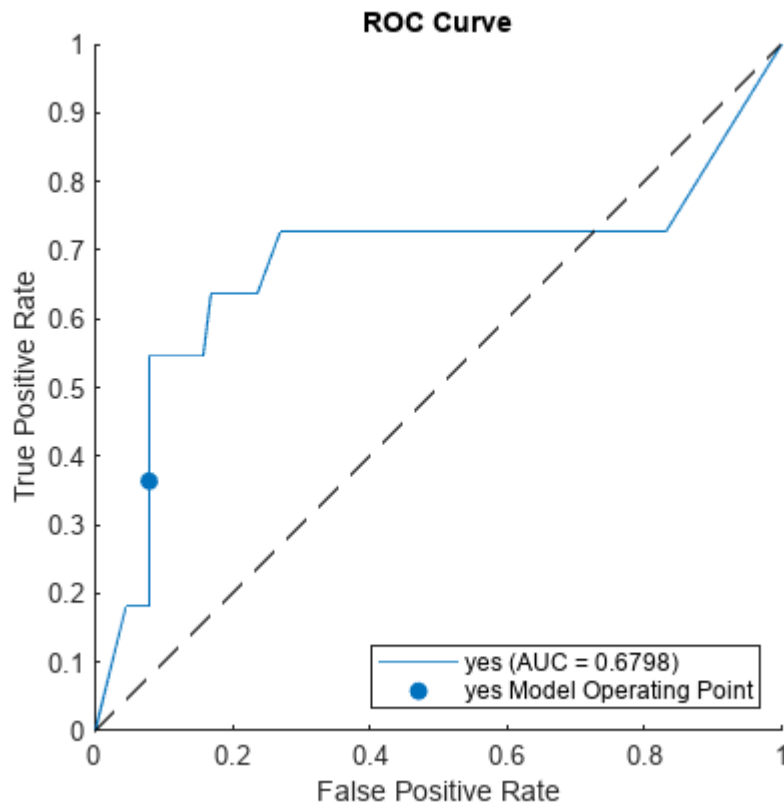
```
Precision_Tred = precision(test_set.subscribed, predictions_Tred, 'yes', 'no')
```

```
Precision_Tred = 0.3636
```

```
rocObj = rocmetrics(test_set.subscribed, scores_Tred, tree_red.ClassNames);
AUC_Tred = rocObj.AUC(end)
```

```
AUC_Tred = 0.6798
```

```
plot(rocObj, ClassNames='yes')
xlim([0,1])
ylim([0,1])
```



```
% saveas(gcf, strcat('figures/ROC_Tred.png'))
```

## Clasiffication model: Naive Bayes

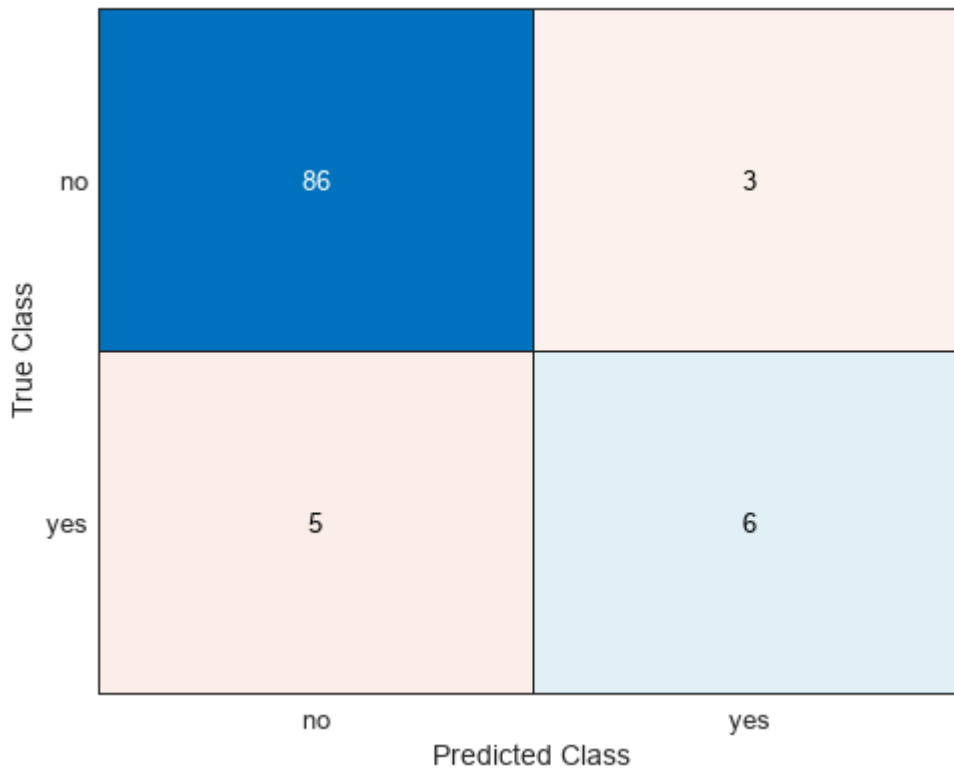
Train a decision tree with the provided dataset, capable of describing the subscription status of the client based on the other variables

```
NaiveBayes = fitcnb(training_set, 'subscribed')
```

```
NaiveBayes =
  ClassificationNaiveBayes
    PredictorNames: {'age' 'job' 'marital' 'education' 'default' 'balance' 'housing' 'loan' 'contact' ...
    ResponseName: 'subscribed'
    CategoricalPredictors: [2 3 4 5 7 8 9 11 16]
    ClassNames: [no yes]
    ScoreTransform: 'none'
    NumObservations: 2000
    DistributionNames: {'normal' 'mvnm' 'mvnm' 'mvnm' 'mvnm' 'normal' 'mvnm' 'mvnm' 'mvnm' 'normal' ...
    DistributionParameters: {2x17 cell}
    CategoricalLevels: {[ ] [12x1 double] [3x1 double] [4x1 double] [2x1 double] [ ] [2x1 double] [2x1 double] ...
```

Properties, Methods

```
[predictions_NB, scores_NB] = predict(NaiveBayes, test_set);
confusionchart(test_set.subscribed, predictions_NB)
```



```
% saveas(gcf, strcat('figures/Confussion_NB.png') )
CrossVal_NB = kfoldLoss(crossval(NaiveBayes))
```

```
CrossVal_NB = 0.1250
```

```
Accuracy_NB = accuracy(test_set.subscribed, predictions_NB, 'yes', 'no')
```

```
Accuracy_NB = 0.9200
```

```
Recall_NB = recall(test_set.subscribed, predictions_NB, 'yes', 'no')
```

```
Recall_NB = 0.5455
```

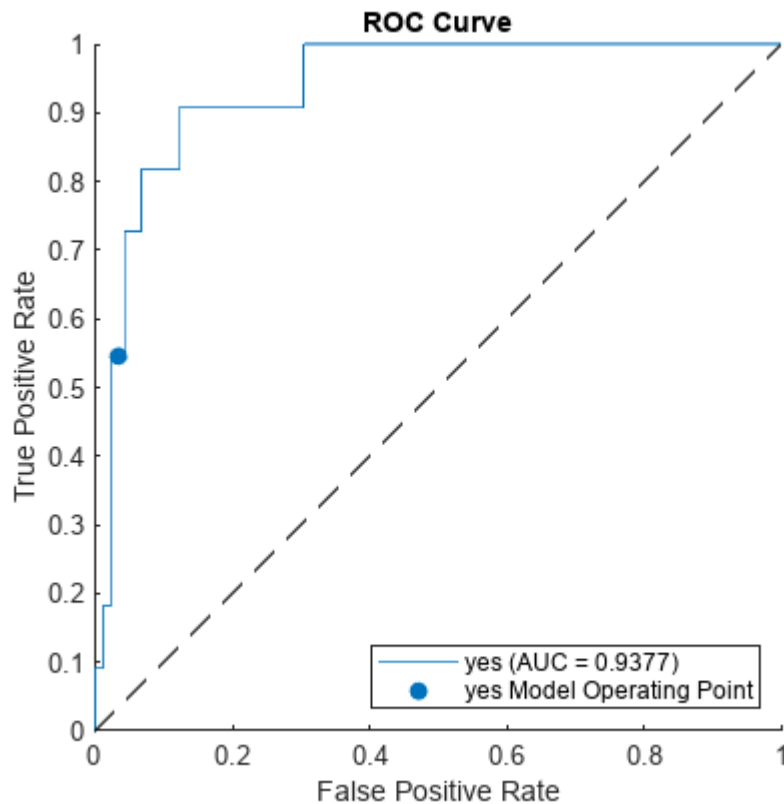
```
Precision_NB = precision(test_set.subscribed, predictions_NB, 'yes', 'no')
```

```
Precision_NB = 0.6667
```

```
rocObj = rocmetrics(test_set.subscribed, scores_NB, NaiveBayes.ClassNames);
AUC_NB = rocObj.AUC(end)
```

```
AUC_NB = 0.9377
```

```
plot(rocObj, ClassNames='yes')
xlim([0,1])
ylim([0,1])
```



```
% saveas(gcf, strcat('figures/ROC_NB.png'))
```

## Include prior probabilities

```
prob_subscribed = sum(training_set.subscribed == 'yes')/numel(training_set);
% classNames = ['yes', 'no'];
prior = [1-prob_subscribed prob_subscribed];
```

```
NB_prior = fitcnb(training_set, 'subscribed', 'Prior', prior)
```

```
NB_prior =
```

```
ClassificationNaiveBayes
```

```
    PredictorNames: {'age' 'job' 'marital' 'education' 'default' 'balance' 'housing' 'loan' 'contact'
```

```
    ResponseName: 'subscribed'
```

```
    CategoricalPredictors: [2 3 4 5 7 8 9 11 16]
```

```
    ClassNames: [no yes]
```

```
    ScoreTransform: 'none'
```

```
    NumObservations: 2000
```

```
    DistributionNames: {'normal' 'mvnm' 'mvnm' 'mvnm' 'mvnm' 'normal' 'mvnm' 'mvnm' 'mvnm' 'normal'
```

```
    DistributionParameters: {2x17 cell}
```

```
    CategoricalLevels: {[ ] [12x1 double] [3x1 double] [4x1 double] [2x1 double] [ ] [2x1 double] [2x1 double]
```

Properties, Methods

```
% NB_prior = fitcnb(training_set, 'subscribed', 'ClassNames', classNames, 'Prior', prior)
```

```
[predictions_NBprior, scores_NBprior] = predict(NB_prior, test_set);
confusionchart(test_set.subscribed, predictions_NBprior)
```



True Class	no	88	1
	yes	9	2
		no	yes
		Predicted Class	

```
% saveas(gcf, strcat('figures/Confussion_NBprior.png')) )
CrossVal_NBprior = kfoldLoss(crossval(NB_prior))
```

```
CrossVal_NBprior = 0.0193
```

```
Accuracy_NBprior = accuracy(test_set.subscribed, predictions_NBprior, 'yes', 'no')
```

```
Accuracy_NBprior = 0.9000
```

```
Recall_NBprior = recall(test_set.subscribed, predictions_NBprior, 'yes', 'no')
```

```
Recall_NBprior = 0.1818
```

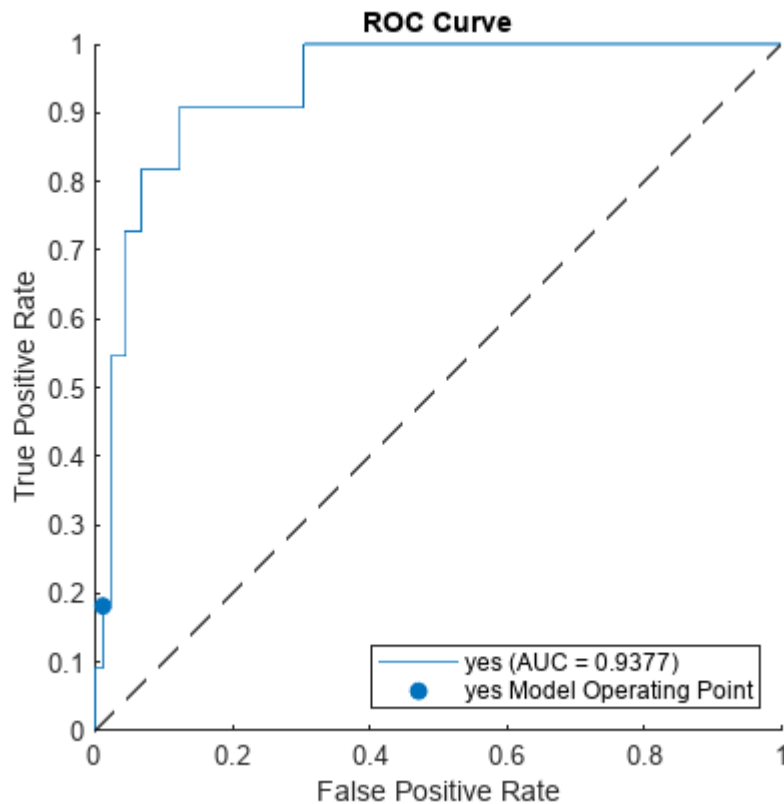
```
Precision_NBprior = precision(test_set.subscribed, predictions_NBprior, 'yes', 'no')
```

```
Precision_NBprior = 0.6667
```

```
rocObj = rocmetrics(test_set.subscribed, scores_NBprior, NB_prior.ClassNames);
AUC_NBprior = rocObj.AUC(end)
```

```
AUC_NBprior = 0.9377
```

```
plot(rocObj, ClassNames='yes')
xlim([0,1])
ylim([0,1])
```



```
% saveas(gcf, strcat('figures/ROC_NBprior.png'))
```

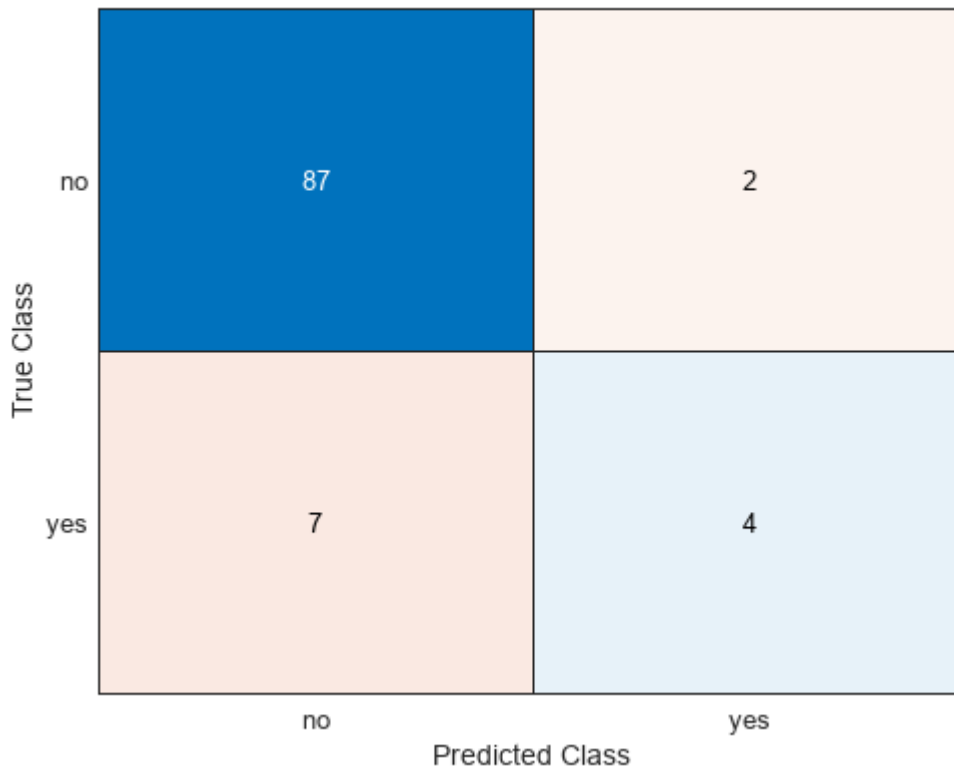
## Reduce number of variables

```
NB_red = fitcnb(removevars(training_set, {'default', 'education', 'housing', 'loan', 'marital', 'prev
```

```
NB_red =
  ClassificationNaiveBayes
    PredictorNames: {'age' 'job' 'balance' 'contact' 'day' 'month' 'duration' 'pdays' 'poutcome'}
    ResponseName: 'subscribed'
    CategoricalPredictors: [2 4 6 9]
    ClassNames: [no yes]
    ScoreTransform: 'none'
    NumObservations: 2000
    DistributionNames: {'normal' 'mvnm' 'normal' 'mvnm' 'normal' 'mvnm' 'normal' 'normal' 'mvnm'}
    DistributionParameters: {2x9 cell}
    CategoricalLevels: {[ ] [12x1 double] [ ] [3x1 double] [ ] [12x1 double] [ ] [ ] [4x1 double]}
```

Properties, Methods

```
[predictions_NBred, scores_NBred] = predict(NB_red, test_set);
confusionchart(test_set.subscribed, predictions_NBred)
```



```
% saveas(gcf, strcat('figures/Confussion_NBred.png')) )
CrossVal_NBred = kfoldLoss(crossval(NB_red))
```

```
CrossVal_NBred = 0.1075
```

```
Accuracy_NBred = accuracy(test_set.subscribed, predictions_NBred, 'yes', 'no')
```

```
Accuracy_NBred = 0.9100
```

```
Recall_NBred = recall(test_set.subscribed, predictions_NBred, 'yes', 'no')
```

```
Recall_NBred = 0.3636
```

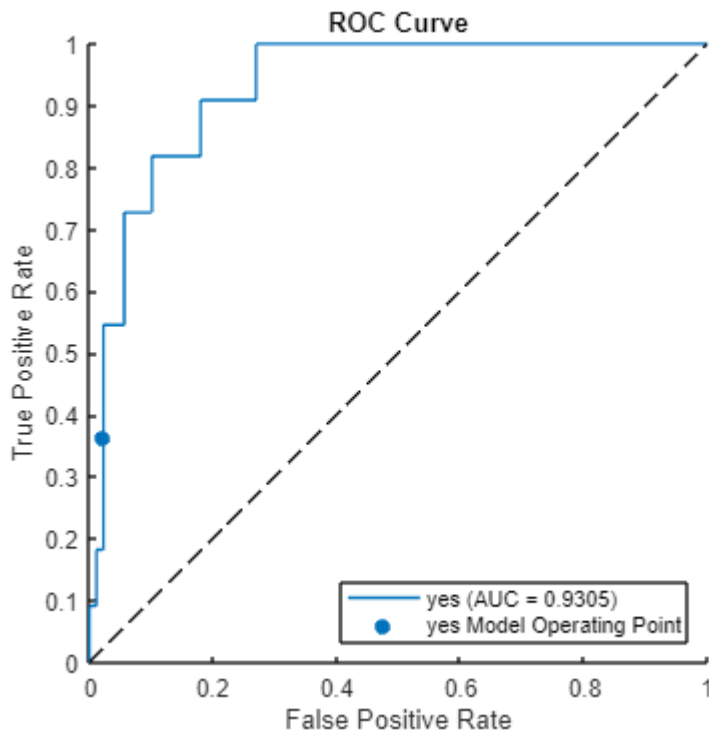
```
Precision_NBred = precision(test_set.subscribed, predictions_NBred, 'yes', 'no')
```

```
Precision_NBred = 0.6667
```

```
rocObj = rocmetrics(test_set.subscribed, scores_NBred, NB_red.ClassNames);
AUC_NBred = rocObj.AUC(end)
```

```
AUC_NBred = 0.9305
```

```
plot(rocObj, ClassNames='yes')
xlim([0,1])
ylim([0,1])
```



```
% saveas(gcf, strcat('figures/ROC_NBred.png'))
```

## Analysis of False Positives and False Negatives

```
% Indexes of false positives for different methods
i_FP = [test_set.subscribed == 'no' & predictions_Tfull == 'yes', ...
        test_set.subscribed == 'no' & predictions_Tred == 'yes', ...
        test_set.subscribed == 'no' & predictions_NB == 'yes', ...
        test_set.subscribed == 'no' & predictions_NBprior == 'yes', ...
        test_set.subscribed == 'no' & predictions_NBred == 'yes'];

i_FN = [test_set.subscribed == 'yes' & predictions_Tfull == 'no', ...
        test_set.subscribed == 'yes' & predictions_Tred == 'no', ...
        test_set.subscribed == 'yes' & predictions_NB == 'no', ...
        test_set.subscribed == 'yes' & predictions_NBprior == 'no', ...
        test_set.subscribed == 'yes' & predictions_NBred == 'no'];

test_set.FP = sum(i_FP, 2);
test_set.FN = sum(i_FN, 2);

FP_set_any = test_set(test_set.FP>0 & test_set.FP<5,:);
FP_set_all = test_set(test_set.FP==5,:);
FN_set_any = test_set(test_set.FN>0 & test_set.FN<5,:);
FN_set_all = test_set(test_set.FN==5,:);
```

## Make some graphs

```
nbins = 25;

for col = 1: width(training_set)
    figure

    if iscategorical(table2array(test_set(:,col)))
        Y(1,:) = countcats( table2array(FP_set_any(:,col)) ) ./ ...
            countcats( table2array(test_set(:,col)) );
        Y(2,:) = countcats( table2array(FP_set_all(:,col)) )' ./ ...
            countcats( table2array(test_set(:,col)) );
        Y(3,:) = countcats( table2array(FN_set_any(:,col)) ) ./ ...
            countcats( table2array(test_set(:,col)) );
        Y(4,:) = countcats( table2array(FN_set_all(:,col)) ) ./ ...
            countcats( table2array(test_set(:,col)) );

        ba = bar(categorical(categories(table2array(test_set(:,col)))), ...
            Y', 'stacked', 'FaceColor','flat');
        ba(1).CData = [1.0 0 0];
        ba(2).CData = [0.6 0 0];
        ba(3).CData = [0 1.0 0];
        ba(4).CData = [0 0.6 0];
        clear Y
    else
        hist_edges = linspace(min(table2array(training_set(:,col))), ...
            max(table2array(training_set(:,col))), nbins+1);
        plot_edges = linspace(hist_edges(2)/2, (hist_edges(end-1)+hist_edges(end))/2, nbins);

        X(1,:) = histcounts( table2array(FP_set_any(:,col)), hist_edges) ./ ...
            histcounts( table2array(test_set(:,col)), hist_edges);
        X(2,:) = histcounts( table2array(FP_set_all(:,col)), hist_edges) ./ ...
            histcounts( table2array(test_set(:,col)), hist_edges);
        X(3,:) = histcounts( table2array(FN_set_any(:,col)), hist_edges) ./ ...
            histcounts( table2array(test_set(:,col)), hist_edges);
        X(4,:) = histcounts( table2array(FN_set_all(:,col)), hist_edges) ./ ...
            histcounts( table2array(test_set(:,col)), hist_edges);

        ba = bar(plot_edges, X', 'stacked', 'FaceColor','flat');
        ba(1).CData = [1.0 0 0];
        ba(2).CData = [0.6 0 0];
        ba(3).CData = [0 1.0 0];
        ba(4).CData = [0 0.6 0];
    end

    title(training_set.Properties.VariableNames{col})
    if col == 5
        title(' default ')
    end
    legend('any FP','all FP','any FN','all FN','Location','eastoutside')
```

```
% saveas(gcf, strcat('figures/Hist_FPFN_', test_set.Properties.VariableNames{col}, '.png'))
end
```

