## 1&2 Your own classifier & pre-coded machine learning techniques

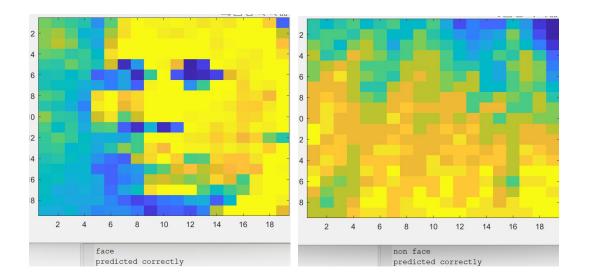
```
function classification data = class train(X, Y)
classification_data = [X;Y];
end
function Y = classify(X,CD)
train set = CD(1:size(CD,1)-1,:);
diff = zeros(1,size(train_set,2));
for i = 1:size(train_set,2)
   diff(i) = norm(X - train_set(:,i));
end
[value,position] = min(diff);
Y = CD(end, position);
end
mean err rate test 💥
1x4 double
                         3
                                   4
    0.1575
              0.1525
                         0.0433
                                   0.1575
 mean err rate train 💥
1x4 double
               2
                         3
    1
         0
              0.0115
                              0
```

1:my own method. 2:regression tree. 3:SVM 4:Nearest neighbor

The error rates of training data is 0, because I use nearest neighbor method, the classify result is just choosing the nearest data in the training data which can find exactly the same picture when we test it on the training data, so the label can not be wrong.

The error rates of test data is higher than training data. And among them, the SVM classifier(3rd) has the lowest error rate, and the Regression tree Classifier and Nearest Neighbor Classifier have almost the same error rates.

My own NN classifier has the same error rate with the NN method by built in function.

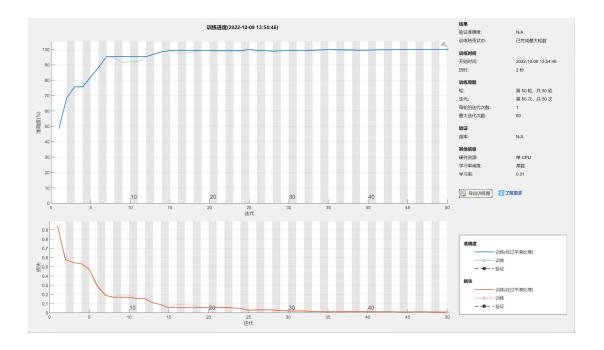


## 3 Testing a simple CNN model

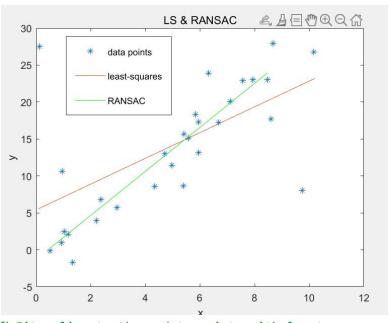
mean_err_rate_test 🗶		mean_err_rate_train 1x4 double		×	
1x4 double					
1	2		1	2	
0.0515		0	0	10000	0

This model is much more complex than the previous models, this is a CNN model contains 3 convolutional layers, 2 max pooling layers and a fully connected layer. And it takes over 5 minutes to run this model, it's a long time compared to the previous methods.

The error rate in test data is lower than regression tree and NN and my own classifier, and a bit higher than the SVM classifier.



4 Line fit



```
% Fit a line to these data points with least squares
train_xm = [xm,ones(N,1)];
train ym = ym;
p_ls = train_xm\train_ym;
plot(x_fine, p_ls(1) * x_fine + p_ls(2))
% Fit a line to these data points using RANSAC.
bestins = 0;
theshold = 1;
for i=1:100
   idx = randperm(N,2);
   k = (ym(idx(1),1)-ym(idx(2),1))/(xm(idx(1),1)-xm(idx(2),1));
   m = ym(idx(1),1) - k*xm(idx(1),1);
   line = [k -1 m];
   distance = abs(line*[xm';ym';ones(1,N)]);
   ins = sum(distance<theshold);</pre>
   if ins > bestins
       bestins = ins;
       trueline = line;
   end
end
pa(1) = trueline(1);
pa(2) = trueline(3);
plot(x_fine,pa(1)*x_fine+pa(2),'k--')
```

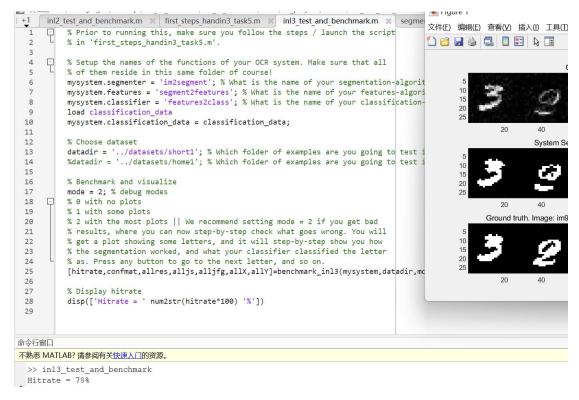
Need 2 points to sample to estimate the line model in RANSAC.

Least squares method can be affected by noisy points (outliers), and can not fit the line well, and RANSAC approach can refit a line to the inliers, and this line is robust to the outliers that RANSAC identified and ignored.

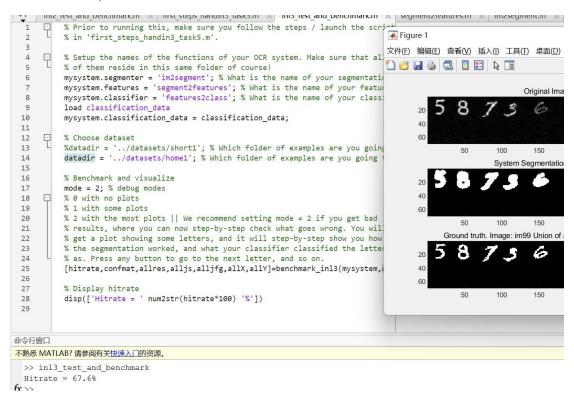
```
% WRITE CODE BELOW TO COMPUTE THE 4 ERRORS
LS_squared_vertical_errors = 0;
LS_squared_orthogonal_errors = 0;
RANSAC squared vertical errors = 0;
RANSAC_squared_orthogonal_errors = 0;
for i =1:N
   LS_squared_vertical_errors = (p_ls(1) * xm(i) + p_ls(2) - ym(i))^2 +
LS squared vertical errors;
   LS_squared_orthogonal_errors =
(abs(p_ls(1)*xm(i)+p_ls(2)-ym(i))/sqrt(p_ls(1)^2+1))^2 +
LS_squared_orthogonal_errors;
   RANSAC_squared_vertical_errors = (modelInliers(1) * xm(i) + modelInliers(2)
- ym(i))^2 + RANSAC squared vertical errors;
   RANSAC_squared_orthogonal_errors = (abs(modelInliers(1)*xm(i) +
modelInliers(2)-ym(i))/sqrt(modelInliers(1)^2+1))^2 +
RANSAC_squared_orthogonal_errors;
end
LS squared vertical errors
LS_squared_orthogonal_errors
RANSAC_squared_vertical_errors
RANSAC_squared_orthogonal_errors
LS_squared_vertical_errors =
  1.4839e+03
LS_squared_orthogonal_errors =
  367.7324
RANSAC_squared_vertical_errors =
  1.9028e+03
RANSAC_squared_orthogonal_errors =
  192.2856
Among the four error, squared_orthogonal_error (TLS) for RANSAC is the lowest.
The two LS error are much higher than the two TLS error.
So RANSAC is the better approach than LS, because its TLS is lower.
```

## 5 OCR system construction and system testing

For the short1, Hit rate = 78%



## For the home1, Hit rate =67.6 %



When I test the short1 data, my segment function works well and give the 78% hit rate. But when I test the home1 data, there are more noise in the image, so the segment function always recognizes some noise point as a digit, so I have to adjust the gaussian filter's standard deviation from 0.5 to 1.25, and then it can work.

We do not demand a complete report for these assignments, but more text and structure would have gotten you much further.

- 1) Comment your own classifier specifically, for example why we get exactly 0 for training error.
- 3) Comment the method, ex model complexity, running times etc.
- 4) "How is the line model estimated for the minimal dataset in the RANSAC approach?" I.e. how would you calculate the line if you do not use polyfit?

"Atleast2pointstosampletoestimatethelinemodelinRANSAC" No, exactly 2 points.

Please address my comments and resubmit.