Table of Contents

Parameter Identification	1
Fitting Linear Function	1
Finding Lambda	1
Running on Test Set	2
Fitting Polynomial Function	2
Finding Lambda	2
Finding polynomial degree order	
Running on Test Set	4
Fitting Gassian Kernel	4
Finding Lambda	
Finding Sigma	5
Running on Test Set	6
Fitting Laplace Kernel	
Finding Lambda	6
Finding Sigma	7
Running on Test Set	8
Fitting Histogram Intersection Kernel	8
Finding Lambda	8
Running on Test Set	9
Fitting Wavelet Kernel	
Finding Lambda	9
Running on Test Set	
Functions	1

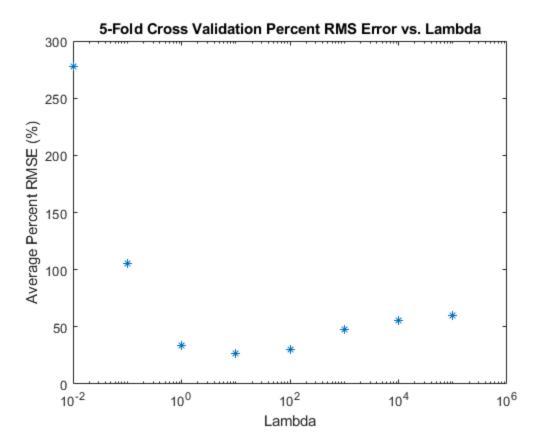
Parameter Identification

```
% Authors: James, Rahul, Tom
% EE 269
% 11/9/2021

close all;
clear;
clc;
% List of kernels to try:http://crsouza.com/2010/03/17/kernel-functions-for-machine-learning-applications/#laplacian
```

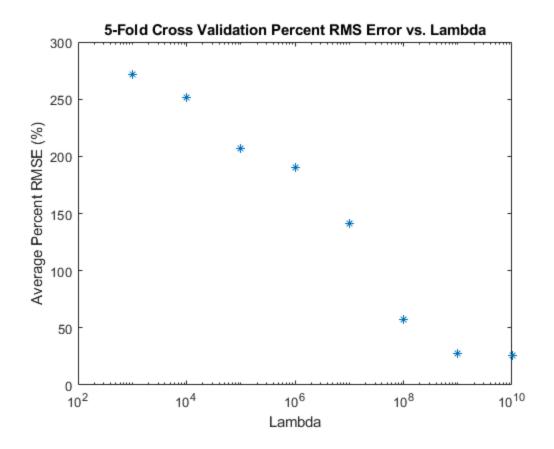
Fitting Linear Function

Finding Lambda



Fitting Polynomial Function Finding Lambda

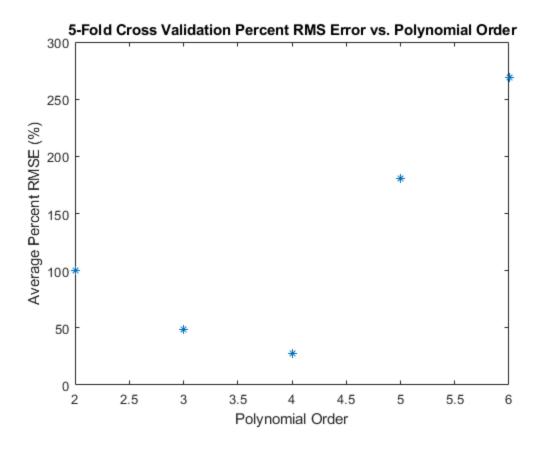
```
load("cross_validation_data.mat");
kernel = @polynomial_kernel;
lambdas = logspace(3,10,8);
order = 4;
fig_handle1 = find_lambda(X_valid1,X_valid2,X_valid3,X_valid4,X_valid5,...
Y_valid1,Y_valid2,Y_valid3,Y_valid4,Y_valid5,kernel,lambdas,order);
```



Finding polynomial degree order

```
load("cross_validation_data.mat");
kernel = @polynomial_kernel;
lambda = 10^9;
orders = 2:6;
fig_handle2 = find_hyperparam(X_valid1, X_valid2, X_valid3, X_valid4, X_valid5,...

Y_valid1, Y_valid2, Y_valid3, Y_valid4, Y_valid5, kernel, lambda, orders, 'Polynomial Order');
```



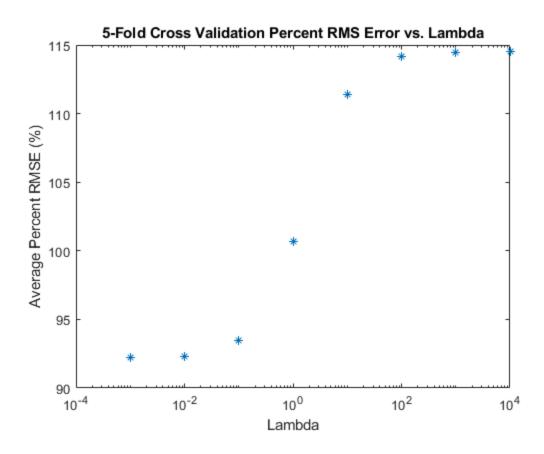
Fitting Gassian Kernel

Finding Lambda

```
load("cross_validation_data.mat");
kernel = @gaussian_kernel;
lambdas = logspace(-3,4,8);
sigma = 0.1;
fig_handle1 = find_lambda(X_valid1,X_valid2,X_valid3,X_valid4,X_valid5,...
```

Percent RMS Error (on Test Set with Polynomial Kernel) = 20.877%

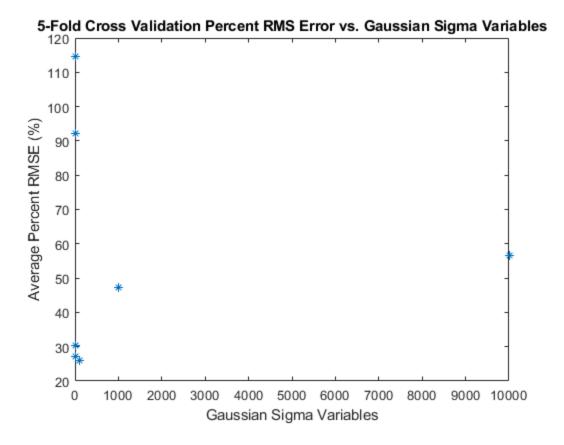
Y_valid1,Y_valid2,Y_valid3,Y_valid4,Y_valid5,kernel,lambdas,sigma);



Finding Sigma

```
load("cross_validation_data.mat");
kernel = @gaussian_kernel;
lambda = 10^-3;
sigmas = logspace(-3,4,8);
fig_handle2 = find_hyperparam(X_valid1,X_valid2,X_valid3,X_valid4,X_valid5,...

Y_valid1,Y_valid2,Y_valid3,Y_valid4,Y_valid5,kernel,lambda,sigmas,'Gaussian
Sigma Variables');
```



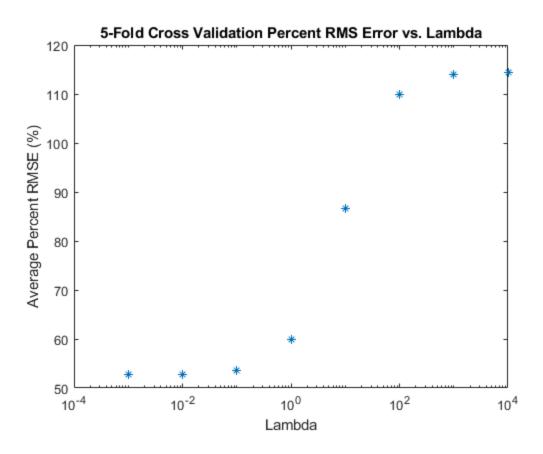
Fitting Laplace Kernel

Finding Lambda

```
load("cross_validation_data.mat");
kernel = @laplace_kernel;
lambdas = logspace(-3,4,8);
sigma = 0.1;
fig_handle1 = find_lambda(X_valid1,X_valid2,X_valid3,X_valid4,X_valid5,...
```

Percent RMS Error (on Test Set with Gaussian Kernel) = 10.0235%

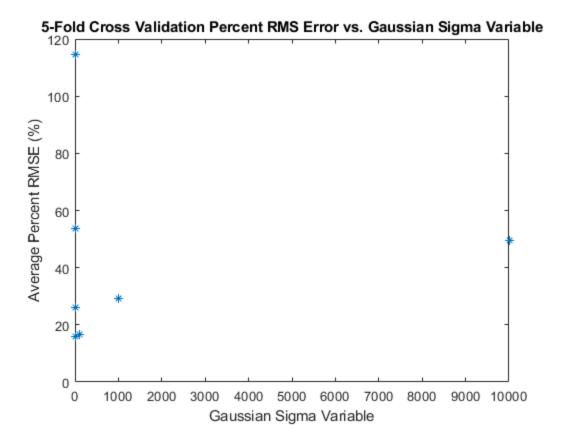
Y_valid1,Y_valid2,Y_valid3,Y_valid4,Y_valid5,kernel,lambdas,sigma);



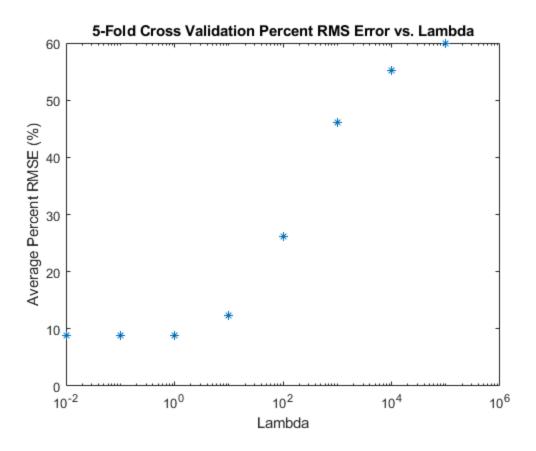
Finding Sigma

```
load("cross_validation_data.mat");
kernel = @laplace_kernel;
lambda = 10^-1;
sigmas = logspace(-3,4,8);
fig_handle2 = find_hyperparam(X_valid1,X_valid2,X_valid3,X_valid4,X_valid5,...

Y_valid1,Y_valid2,Y_valid3,Y_valid4,Y_valid5,kernel,lambda,sigmas,'Gaussian
Sigma Variable');
```



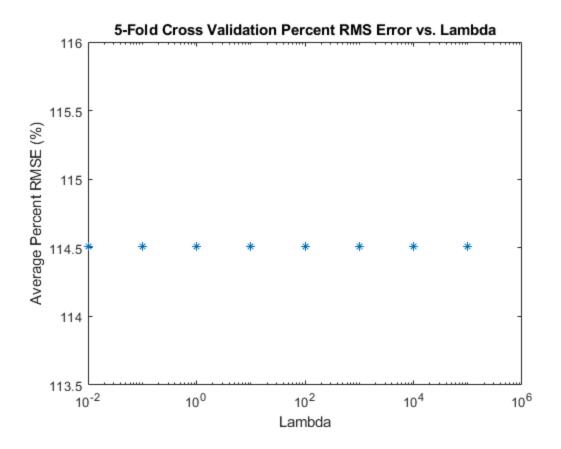
Fitting Histogram Intersection Kernel Finding Lambda



Percent RMS Error (on Test Set with Histogram Intersection Kernel) = 8.1782%

Fitting Wavelet Kernel

Finding Lambda



Functions

```
function k_xi_xj = linear_kernel(xi,xj,hyperparam)
         k_xi_xj = reshape(xi,1,[])*reshape(xj,[],1);
end

function k_xi_xj = polynomial_kernel(xi,xj,order)
         k_xi_xj = (1 + reshape(xi,1,[])*reshape(xj,[],1))^order;
end
```

Percent RMS Error (on Test Set with Wavelet Kernel) = 113.6997%

```
function k_xi_xj = gaussian_kernel(xi,xj,sigma)
    k xi xj = \exp(-(norm(reshape(xi,1,[]) - reshape(xj,1,[]),2)^2)/
(2*sigma^2));
end
function k_xi_xj = laplace_kernel(xi,xj,sigma)
    k_xi_xj = \exp(-norm(reshape(xi,1,[]) - reshape(xj,1,[]),2)/sigma);
end
function k_xi_xj = histogram_intersection_kernel(xi,xj,hyperparam)
    k_xi_xj = sum(min(reshape(xi,[],1),reshape(xj,[],1)));
end
function k_xi_xj = wavelet_kernel(xi,xj,hyperparam)
     k_xi_xj = prod(wavelet(reshape(xi,1,[])).*wavelet(reshape(xi,1,[])));
end
function y = wavelet(x)
    y = cos(1.75*x).*exp(-(x.^2)/2);
end
%rows of X are samples
function K = form_K(X_train, X_test, kernel, hyper_param)
    N = length(X train(:,1));
    M = length(X_test(:,1));
    K = zeros(M,N);
    for i = 1:1:M
        for j = 1:1:N
            K(i,j) = kernel(X_test(i,:),X_train(j,:),hyper_param);
        end
    end
end
function y_test_pred = predict(X_train,Y_train,X_test,lambda,
kernel, hyper param)
    K_train = form_K(X_train, X_train, kernel, hyper_param);
    alpha = inv((1/lambda)*K train + eye(size(K train)))*Y train;
    K_test = form_K(X_train, X_test, kernel, hyper_param);
    y_test_pred = (1/lambda)*K_test*alpha;
end
function avg prmse =
 cross_validate(X1,X2,X3,X4,X5,Y1,Y2,Y3,Y4,Y5,kernel,lambda,hyperparam)
    X = \{X1, X2, X3, X4, X5\};
    Y = \{Y1, Y2, Y3, Y4, Y5\};
    num folds = 5;
    folds = 1:1:num folds;
    prmse = 0;
    for i = folds
        idxs = folds;
        idxs(idxs == i) = [];
        [x1,x2,x3,x4] = X{idxs};
        [y1,y2,y3,y4] = Y{idxs};
        X_{train} = [x1;x2;x3;x4];
```

```
X_{test} = X\{i\};
        Y \text{ train} = [y1;y2;y3;y4];
        Y_{test} = Y{i};
        Y_test_pred = predict(X_train,Y_train,X_test,lambda,
 kernel,hyperparam);
        prmse = prmse + calc_percent_rmse(Y_test,Y_test_pred);
    end
    avg prmse = prmse/num folds;
end
function prmse =
 test_evaluation(X_test,Y_test,X1,X2,X3,X4,X5,Y1,Y2,Y3,Y4,Y5,kernel,lambda,hyperparam)
    X \text{ train} = [X1; X2; X3; X4; X5];
    Y_train = [Y1;Y2;Y3;Y4;Y5];
    Y_test_pred = predict(X_train,Y_train,X_test,lambda, kernel,hyperparam);
    prmse = calc_percent_rmse(Y_test,Y_test_pred);
end
function fig handle =
 find_lambda(X_valid1,X_valid2,X_valid3,X_valid4,X_valid5,...
 Y_valid1,Y_valid2,Y_valid3,Y_valid4,Y_valid5,kernel,lambdas,hyperparam)
    avg_prmses = zeros(length(lambdas),1);
    figure();
    for i = 1:1:length(lambdas)
        lambda = lambdas(i);
        avg_prmses(i) =
 cross_validate(X_valid1, X_valid2, X_valid3, X_valid4, X_valid5, ...
 Y valid1, Y valid2, Y valid3, Y valid4, Y valid5, kernel, lambda, hyperparam);
    semilogx(lambdas,avg_prmses,'*');
    title('5-Fold Cross Validation Percent RMS Error vs. Lambda');
    xlabel('Lambda'); ylabel('Average Percent RMSE (%)');
    fig handle = gcf;
end
function fig_handle =
 find_hyperparam(X_valid1,X_valid2,X_valid3,X_valid4,X_valid5,...
 Y_valid1,Y_valid2,Y_valid3,Y_valid4,Y_valid5,kernel,lambda,hyperparams,hyperparam_name)
    avg_prmses = zeros(length(hyperparams),1);
    figure();
    for i = 1:1:length(hyperparams)
        hyperparam = hyperparams(i);
        avg prmses(i) =
 cross_validate(X_valid1,X_valid2,X_valid3,X_valid4,X_valid5,...
 Y_valid1,Y_valid2,Y_valid3,Y_valid4,Y_valid5,kernel,lambda,hyperparam);
    plot(hyperparams,avg_prmses,'*');
    title(['5-Fold Cross Validation Percent RMS Error vs. ' hyperparam name]);
    xlabel(hyperparam_name); ylabel('Average Percent RMSE (%)');
    fig_handle = gcf;
```

end

```
function prmse = calc_percent_rmse(x_actual,x_corrupted)
   N = length(x_actual);
   prmse = sqrt((1/N)*sum((reshape(x_corrupted,[],1) - reshape(x_actual,[],1))).^2))*(100*N/sum(reshape(x_actual,[],1)));
end
```

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