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### **EE239AS HW #6**

```
% Collaborators: Vikranth, Yusi
clc
clear
close all
```

### **Problem 4**

```
load('/Users/Yusi/Documents/EE239AS/HW6/JR_2015-12-04_truncated2.mat');
% from part I of problem 1 (calculating spike counts for each direction)
n_trials = length(R);
n_electrodes = 96;
dt = 25;
```

### Part A: A Matrix

```
v_xx = 0.7;
v_yy = 0.7;
v_yx = 0;
v_xy = 0;
A = [1 0 dt 0 0;
    0 1 0 dt 0;
    0 0 v_xx v_xy 0;
    0 0 v_yx v_yy 0;
    0 0 0 0 1];
% extra column and row of zeros accounts for the bias term
fprintf('\nPart A -- Kalman Filter A Matrix: \n')
disp(A)
```

```
Part A -- Kalman Filter A Matrix:
   1.0000
           0 25.0000
       0
           1.0000
                    0
                          25.0000
       0
             0
                  0.7000
       0
               0
                       0
                          0.7000
                                         0
                        0
                                0
                                     1.0000
```

#### Part B: A Matrix

```
X = [R(1:400).cursorPos];
sample_ind = 1:25:length(X);
pos_bin = X(1:2,sample_ind);
X_{bin} = diff(pos_{bin}, 1, 2)/25;
% Euler approximation of velocities
X_k = [X_bin; ones(1, size(X_bin,2))];
A_s = X_k(:, 2:end)*pinv(X_k(:, 1:end-1));
\mbox{\ensuremath{\$}} use ML to estimate v_xx, v_yy, v_yx, v_xy
A(3:4,3:4) = A_s(1:2, 1:2);
fprintf('\nPart B -- Kalman Filter A Matrix: \n')
disp(A)
Part B -- Kalman Filter A Matrix:
    1.0000
             0 25.0000
              1.0000
                                25.0000
         0
                            0
                                                  0
                                -0.0074
         0
                   0
                        0.7798
         0
                   0
                        0.0096
                                 0.7808
                                             1.0000
         0
                           0
                                      0
```

## Part C: C Matrix

```
Y = [R(1:400).spikeRaster];

Y_reach_raster = full(Y);
Y_bin = binFunc(Y_reach_raster, dt);
Y_bin = Y_bin(:, 1:end-1);

Y_k = Y_bin;
C = zeros(n_electrodes, 5);
C_s = Y_k * pinv(X_k);

C(:,end-2:end) = C_s;
C_sum = sum(C,1);

fprintf('\nPart C -- Sum of C Across Electrodes:\n')
disp(C_sum)
```

```
Part C -- Sum of C Across Electrodes:
0 0 -4.4994 1.7053 40.1927
```

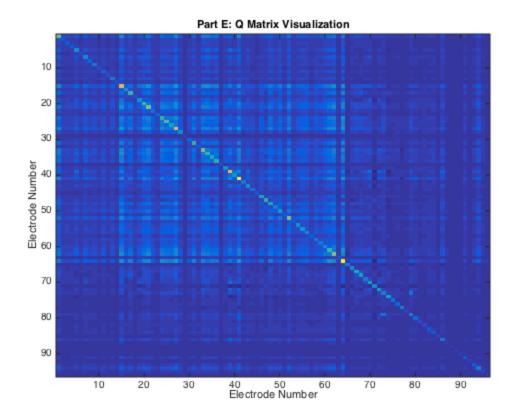
### Part D: W Matrix

```
W = zeros(size(A));
W(3:4,3:4) = W_s(1:2,1:2);
fprintf('\nPart D -- Kalman Filter W Matrix: \n')
disp(W)
Part D -- Kalman Filter W Matrix:
        0 0
           0
     0
                  0
                    -0.0008
              0.0115
     0
           0
     0
           0 -0.0008 0.0121
```

### Part E: Q Matrix

```
Q = 1/size(X_k,2)*(Y_k-C_s*X_k)*(Y_k-C_s*X_k)';

figure(1)
imagesc(Q)
title('Part E: Q Matrix Visualization')
xlabel('Electrode Number')
ylabel('Electrode Number')
% tells you how well y_k approximates Cx_k
% diagonal is for each electrode
% off diagonals are co-varied in residuals
% dark neurons are close to 0 firing rate, and C is close to 0
```



# Part F: Steady State Kalman Filter

```
S = zeros(size(W));
tol = 10^{-13};
iterativeDiff = inf;
M_1_prev = 0;
M_2_prev = 0;
iter = 0;
while any(abs(iterativeDiff)>tol)
    iter = iter+1;
    S = A*S*A' + W;
    % introduce uncertainty
    S(1:2,:) = 0;
    S(:, 1:2) = 0;
    % zero out uncertainty so it does not propagate
    % calculate S (k_k-1) using previous S (k-1_k-1) value
    K_k = S*C'*inv(C*S*C' + Q);
    % calculate Kalman gain from updated S value
    S = S - S*C'*inv(C*S*C' + Q)*C*S;
    % calculate S (k_k) using previous S (k_k-1) value
    M_1 = A-K_k*C*A;
    M_2 = K_k;
```

```
iterativeDiff = [reshape(M 1, 1, []) - reshape(M 1 prev,1,[])...
    reshape(M_2, 1, []) - reshape(M_2_prev,1,[])];
    M 1 prev = M 1;
    M_2_prev = M_2;
end
fprintf('\nPart F -- Kalman Filter M_1 Matrix:\n')
disp(M_1)
fprintf('\nPart F -- Kalman Gain (All Electrodes):\n')
disp(sum(M_2,2))
Part F -- Kalman Filter M 1 Matrix:
    1.0000
                  0 25.0000
                                                 0
              1.0000
                           0
                                 25.0000
                                                 0
         0
                   0
                       0.6776
                               -0.0254
                                            0.0123
                   0 -0.0154
                               0.6276
                                          -0.0586
         0
                                           1.0000
         0
                   0
                            0
                                      0
Part F -- Kalman Gain (All Electrodes):
         0
         0
   -0.0583
    0.1267
```

## Part G: Kalman Filter Decoding

```
X_true = [R(401:end).cursorPos];
figure(2)
scatter(X_true(1,:), X_true(2,:))
title('Part G: True Hand Positions')
xlabel('X-Position (mm)')
ylabel('Y-Position (mm)')
figure(3)
hold on
for i = 1:106
    scatter(X_decode{i}(1,:), X_decode{i}(2,:))
end
hold off
title('Part G: Kalman Filter Decoded Hand Positions')
xlabel('X-Position (mm)')
ylabel('Y-Position (mm)')
% Part G(b): Mean-Square Error
errors = cell(1,106);
mean_errors = zeros(2,106);
for i = 1:106
    pos\_vec = R(400+i).cursorPos(1:2,:);
    sample_ind_pos = 1:25:length(pos_vec);
    pos_vec = pos_vec(:, sample_ind_pos);
    errors{i} = (pos_vec - X_decode{i}(1:2,:)).^2;
    mean_errors(:,i) = mean(errors{i},2);
end
mean_error_all = sum(mean(mean_errors,2));
fprintf('\nKalman Filter Mean Square Error: %4.2f\n', mean_error_all)
Kalman Filter Mean Square Error: 3716.53
```

end

