## Lab7: Development of forward-backward Kalman filter in conditions of correlated state noise

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```
clc; clear; close all;
addpath('functions/');
n = 200;
sigmaA = 0.2; sigmaN = 20; sigmaXi = 1;
x1 = 5; v1 = 1; t = 1;
G = [0; 0; 1];
H = [1, 0, 0];
P = [ 10000, 0, 0; 0, 10000, 0; 0, 0, 10000; ];
lambda = 0.1;
M = 500;
ErrSum = zeros(3,n);
F = [ 1, t, (t^2)/2;
0, 1, t;
0, 0, exp(-lambda*t) ];
sigmaZeta = (sigmaA^2)*(1-exp(-2*lambda*t));
```

part I: Development of optimal Kalman filter in conditions of correlated state noise

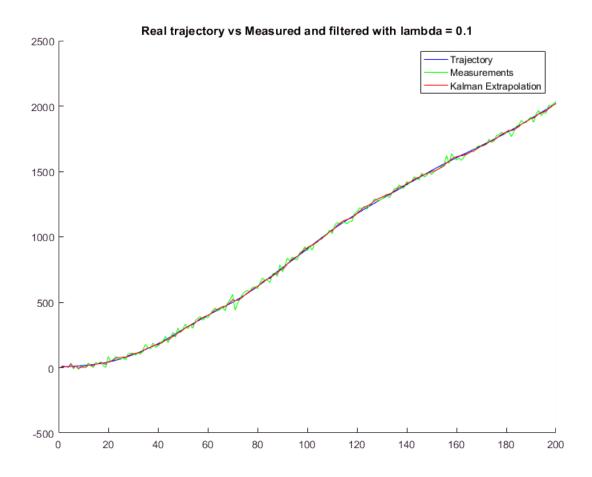
```
for j=1:M
    A = gaussMarkov( n, sigmaA, sigmaXi, lambda, t);
    Noise = normrnd(0, sigmaN, 1, n);

[ X, V, Z ] = calcTrajectory3( A, Noise, x1, v1, t);
    XVA = [ X; V; A ];

[ Xk, SigmaX ] = calcKalman3(Z, sigmaZeta, sigmaN, x1, v1, F, G, H, P, 0);
    ErrCur = ( XVA - Xk ).^2;
    ErrSum = ErrSum + ErrCur;
end

FinalError = ( ErrSum./(M-1) ).^0.5;
```

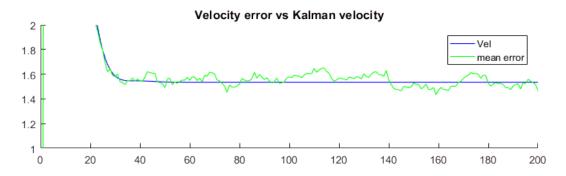
```
figure('position', [0, 0, 800, 600]); hold on
   plot(X,'blue');
   plot(Z, 'green');
   plot(Xk(1,:), 'red');
   title( ['Real trajectory vs Measured and filtered with lambda = ', num2str(lambda) ]);
   legend('Trajectory', 'Measurements', 'Kalman Extrapolation');
```



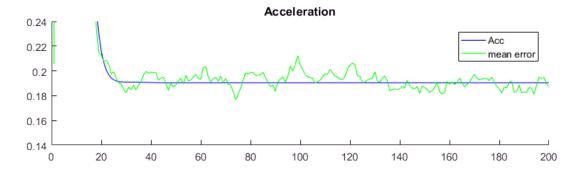
```
plotStartIndex=1;
figure('position', [0, 0, 800, 200]); hold on;
  plot(SigmaX(1, plotStartIndex:end), 'blue');
  plot(FinalError(1, plotStartIndex:end), 'green');
  axis([0 n 8 11])
  title('Trajectory error');
  legend('Trajectory', 'mean error');
```

## Trajectory error Trajectory mean error

```
figure('position', [0, 0, 800, 200]); hold on;
  plot(SigmaX(2, plotStartIndex:end), 'blue');
  plot(FinalError(2, plotStartIndex:end), 'green');
  axis([0 n 1 2])
  title('Velocity error vs Kalman velocity');
  legend('Vel', 'mean error');
```

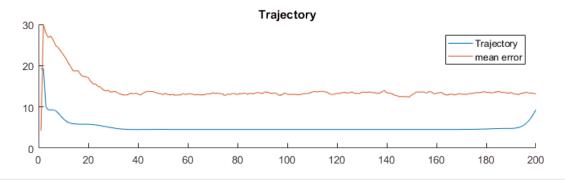


```
figure('position', [0, 0, 800, 200]); hold on;
  plot(SigmaX(3, plotStartIndex:end), 'blue');
  plot(FinalError(3, plotStartIndex:end), 'green');
  axis([0 n 0.14 0.24])
  title('Acceleration');
  legend('Acc', 'mean error');
```

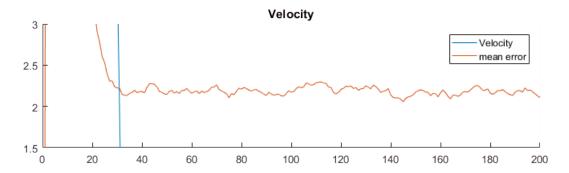


## part II: Development of optimal smoothing to increase the estimation accuracy

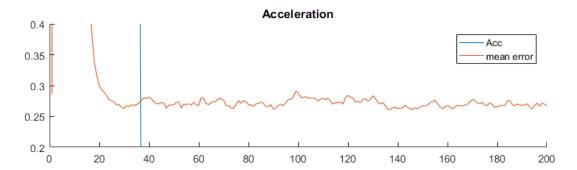
```
for j=1:M
    A = gaussMarkov( n, sigmaA, sigmaXi, lambda, t);
    Noise = normrnd(0, sigmaN, 1, n);
    [ X, V, Z ] = calcTrajectory3( A, Noise, x1, v1, t);
    XVA = [X; V; A];
    [ Xk, SigmaX, PiN] = calcKalmanSmooth(Z, sigmaZeta, sigmaN, <math>x1, v1, F, G, H, P, 0);
    ErrCur = (XVA - Xk).^2;
    ErrSum = ErrSum + ErrCur;
end
FinalError = ( ErrSum./(M-1) ).^0.5; %true estimation error
for i=1:n
PP = PiN\{i\};
PP1(i) = sqrt(PP(1,1));
PP2(i) = sqrt(PP(2,2));
PP3(i) = sqrt(PP(3,3));
end
figure ('position', [0, 0, 800, 200]); hold on
plot(PP1)
plot(FinalError(1,:))
title('Trajectory');
legend('Trajectory', 'mean error');
```



```
figure('position', [0, 0, 800, 200]); hold on
plot(PP2)
plot(FinalError(2,:))
axis([0 n 1.5 3])
title('Velocity');
legend('Velocity', 'mean error');
```



```
figure('position', [0, 0, 800, 200]); hold on
plot(PP3)
plot(FinalError(3,:))
axis([0 n 0.2 0.4])
title('Acceleration');
legend('Acc', 'mean error');
```



```
function [ Xk, SigmaX ] = calcKalman3(Z, sigmaA, sigmaN, x1, v1, F, G,
H, P, bias )
    n = length(Z);
    Xk = zeros(3, n);
    Xk(:, 1) = [2; 0; 0];
    Q = sigmaA * (G*G');
    SigmaX = zeros(3,n);
    SigmaX(1,1) = sqrt(P(1,1));
    SigmaX(2,1) = sqrt(P(2,2));
    SigmaX(3,1) = sqrt(P(3,3));
    for i=2:n
        P=F*P*F'+Q;
        K=P*H'/(H*P*H'+ sigmaN^2);
        Xk(:,i) = F*Xk(:, i-1) + G*bias;
        Xk(:,i) = Xk(:,i)+K*(Z(i)-H*Xk(:,i));
        P = (eye(3)-K*H)*P;
        SigmaX(1,i) = P(1,1)^{(1/2)};
        SigmaX(2,i) = P(2,2)^{(1/2)};
        SigmaX(3,i) = P(3,3)^{(1/2)};
    end
end
Not enough input arguments.
Error in calcKalman3 (line 3)
    n = length(Z);
```

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```
function [ Xk, SigmaX, PiN ] = calcKalmanSmooth(Z, sigmaA, sigmaN, x1,
   v1, F, G, H, P, bias )
              n=length(Z);
              Xk = zeros(3, n);
              Xk(:, 1) = [2; 0; 0];
              Q=sigmaA * (G*G');
              Ak=cell(1,200);
              PiN=cell(1,200);
              Ppredict=cell(1,200);
              Pfiltrate=cell(1,200);
                             Pfiltrate(1)={P};
                             Ppredict(1) = \{P\};
              SigmaX = zeros(3,n);
              SigmaX(1,1) = sgrt(P(1,1));
              SigmaX(2,1) = sqrt(P(2,2));
              SigmaX(3,1) = sqrt(P(3,3));
              for i=2:n
                            P=F*P*F'+Q;
                            Ppredict(i)={P};
                            K=P*H'/(H*P*H'+ sigmaN^2);
                            Xk(:,i) = F*Xk(:, i-1) + G*bias;
                            Xk(:,i) = Xk(:,i)+K*(Z(i)-H*Xk(:,i));
                             P = (eye(3)-K*H)*P;
                            Pfiltrate(i)={P};
                             SigmaX(1,i) = sqrt(P(1,1));
                             SigmaX(2,i) = sqrt(P(2,2));
                             SigmaX(3,i) = sqrt(P(3,3));
              end
              Xks(:,n)=Xk(:,n);
              PiN{n}=Pfiltrate{n};
           for i=n-1:-1:1
                         P1=Ppredict{i};
                         P2=Pfiltrate{i};
                                        Ak(i) = \{P2*F'*(inv(P1))\};
                                           PiN(i) = \{Pfiltrate\{(i)\} + Ak\{(i)\} * (PiN\{(i+1)\} - Ak\{(i)\} + Ak\{(i)\} \} = \{Pfiltrate\{(i)\} + Ak\{(i)\} + Ak\{(i)\} \} = \{Pfiltrate\{(i)\} + Ak\{(i)\} \} = \{Pfiltrate\{
Ppredict{(i)})*(Ak{(i)}')};
                                           Xks(:,i)=Xk(:,i)+Ak\{(i)\}*(Xks(:,i+1)-F*Xk(:,i));
               end
Not enough input arguments.
Error in calcKalmanSmooth (line 3)
              n=length(Z);
```