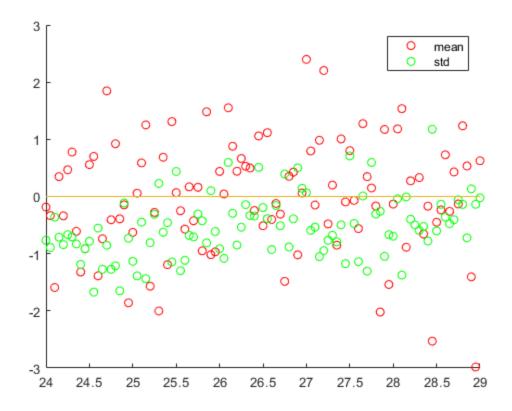
clear;clc;

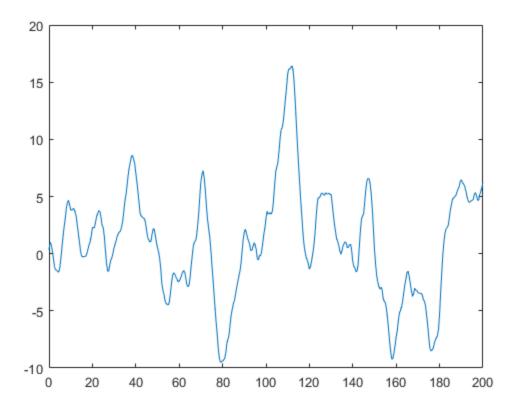
Wind Model

```
[numb,denb]=butter(2,.05);
numb=numb(1:2);
wi=filter(numb,denb,randn(500,1));
t=[0:.5:200]';
[T,\sim] = size(t);
[As,Bs,Cs,Ds]=tf2ss(numb,denb);
Wind_sys = ss(As, Bs, Cs, Ds);
x_w = zeros(2,T+1);
y_w = zeros(T,1);
v_w = randn(T,1);
u_w0 = randn(T,1);
ms\_rec = zeros(100,3);
i = 1;
for Qs = 24:0.05:29
    u_w = randn(T,1)*Qs;
    for k = 1:T
        x_w(:,k+1) = As * x_w(:,k) + Bs * u_w(k);
        y_w(k) = Cs * x_w(:,k) + Ds * v_w(k);
    ms_rec(i,:) = [Qs, mean(y_w)+0.038, std(y_w)-5];
    i = i+1;
end
figure(1);
scatter(ms_rec(:,1), ms_rec(:,2), 'r');
hold on;
scatter(ms_rec(:,1), ms_rec(:,3), 'g');
plot(ms_rec(:,1), zeros(101,1))
legend('mean', 'std');
hold off;
```



Part iv

```
Qs_det = 27;
u_w = randn(T,1)*Qs_det;
wi_Qs = filter(numb,denb,u_w);
figure(2);
plot(t,wi_Qs);
```



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clear;clc;

Wind Model

```
[numb, denb]=butter(2,.05);
t=[0:.5:200]';
[T,~] = size(t);
Qs_det=le-3;
u_w = randn(T,1)*Qs_det;
wi=filter(numb,denb,u_w); % wi is the output of wind velocity.
numb=numb(1:2); % This is the delay process, making the Ds zero.
[As,Bs,Cs,Ds]=tf2ss(numb,denb); % The wind system As,Bs,Cs,Ds.
Wind_force_ratio = 10; %This is a coefficient of relation between wind velocity and force.
```

Cart Model

```
[Ac,Bc,Cc,Dc]=ssdata(c2d(ss([0 1;0 -.2],
  [0;1],eye(2),zeros(2,1)),.5)); % This is the cart model given by the
  prof.
%Gc=[1;10];
Gc=[1e-2;10];
Motor = sin(2*pi*t/10);
```

Combined Model

[Xs;Xc] is 4x1; F=[As 0; Gc*Cs Ac] is 4x4; G=[0;Bc] is 4x1; B1=[Bs;Gc*Ds] is 4x1; H=[0 Cc] is 2x4; J=[0;0]; B2=[1;1]; X=F*X+G*u+B1*process_noise_wind; Y=H*X+J*u+B2*measurement_noise_cart;

```
F = [As/Wind_force_ratio, zeros(2); Gc*Cs/Wind_force_ratio, Ac];
%F = [As, zeros(2); Gc*Cs, Ac];
G = [zeros(2,1);Bc];
B1 = [Bs zeros(2,3);zeros(2,3) Gc*Ds]/Wind_force_ratio;
%B1 = [Bs;Gc*Ds];
%B1 = [Bs zeros(2,3);zeros(2,3) Gc*Ds];
H = [zeros(2), Cc];
```

```
J = [0;0];
B2 = eye(2);
```

Compute the real data

```
X0 = [0;0;2;1];
% The measurement noise v is a 2x1 vector, the first element
represents the
% postion and the second element represents the velocity.
% v = [v_pos; v_vel];
% To get a v.
v_pos = 0.5;
v \ vel = 0.5;
v_all = [v_pos; v_vel];
% To realize these measurement noise.
v_pos_real = randn(1,T)*sqrt(v_pos);
v_vel_real = randn(1,T)*sqrt(v_vel);
v_all_real = [v_pos_real; v_vel_real];
% The process noise is actually the noise in wind model, which is a
randn
% number.
% Qs = 27 is computed in Question 1,
%Qs=27;
Qs=1e-3;
wind_process_noise = randn(T,2)*Gc*sqrt(Qs);
wind_meas_noise = randn(2,T);
d = [wind_process_noise';zeros(1,T); wind_meas_noise];
[X_ss, Y_ss] = myss(F,G,H,J,B1,B2,Motor,X0,v_all_real,d);
```

Using Isim to compute real data

```
 v\_all\_real = v\_vel\_real'; \ Qs=27; \ wv=randn(T,1)*Gc'*sqrt(Qs); \ Y\_ss = lsim(ss(F,G,H,J),Motor,t,X0); \\ Y\_ss = Y\_ss' + [v\_pos\_real; v\_vel\_real];
```

The Kalman Predictor process

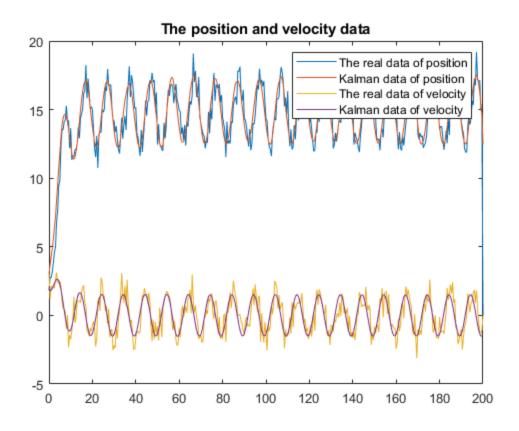
```
% We need to get Ro and Qo.
% Ro is the measurement noise matrix which is 2x2.
% Qo is the process noise matrix which is 4x4;
% In the combined model, the measurement noise only has v_pos and v_vel,
% thus the Ro should be diag([v_pos v_vel])
% Given that the process noise only has the wind noise Qs. Thus Qo should
% be diag([0,Qs,0,0]);
Ro = diag([v_pos, v_vel]);
```

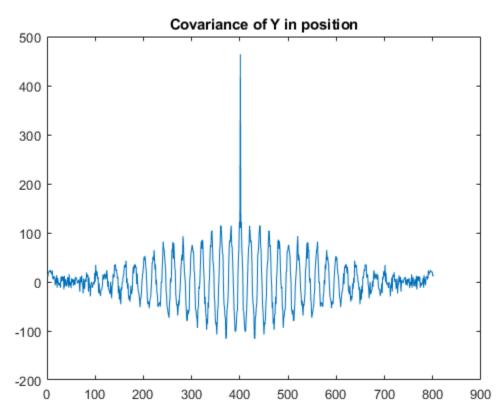
```
Qs = 1e-2;
Qo = diag([7,Qs,1e-4,1e-5]);
% Get X0 and P0
x0 = [0;0;2;1];
P0 = eye(4);
[X_hatrec, Y_hatrec, K_rec, P_rec] =
myKalman(F,G,H,J,x0,Motor,Y_ss,P0,Qo,Ro);
```

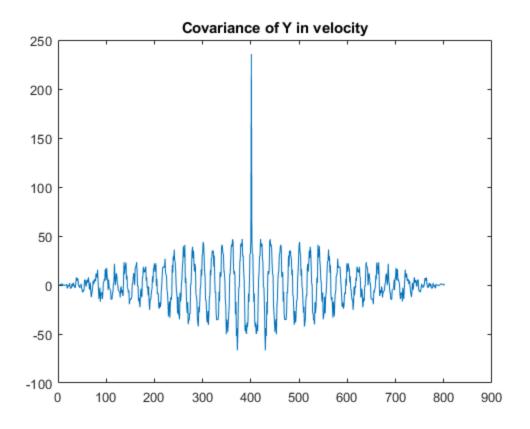
Plotting

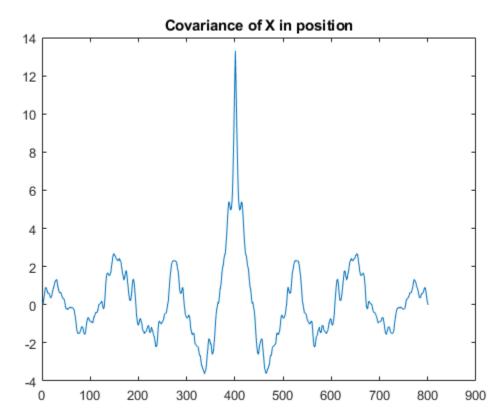
Now we have the real data y_ss and estimated Y_hatrec.

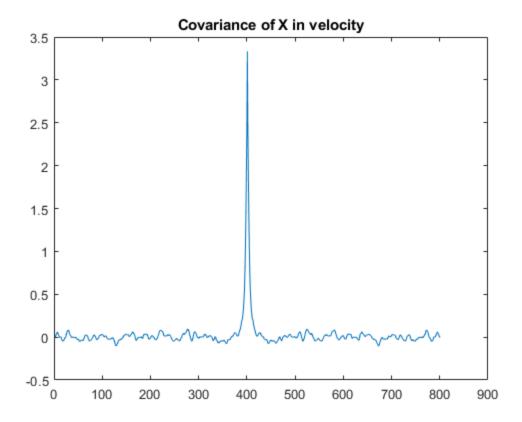
```
figure(1);
plot(t, Y_ss(1,:));
hold on
plot(t, Y_hatrec(1,:));
plot(t, Y_ss(2,:))
plot(t, Y_hatrec(2,:));
hold off;
legend('The real data of position', 'Kalman data of position', 'The real
data of velocity', 'Kalman data of velocity');
title('The position and velocity data');
% figure(2);
% plot(t, X_ss(3,:), t, X_hatrec(3,:));
% hold on;
% plot(t, X_ss(4,:), t, X_hatrec(4,:));
%Y_tilda = Y_ss - Y_hatrec;
Y_tilda = Y_hatrec - Y_ss;
%X_tilda = X_hatrec - X_ss;
%Y_tilda_2_c=xcorr(Y_tilda(2,:)',Y_tilda(2,:)');
figure(3);
plot(xcorr(Y_tilda(1,:)));
title('Covariance of Y in position');
figure(4);
plot(xcorr(Y_tilda(2,:)));
title('Covariance of Y in velocity');
X_tilda = X_hatrec - X_ss;
figure(5);
plot(xcorr(X_tilda(3,:)));
title('Covariance of X in position');
figure(6);
plot(xcorr(X_tilda(4,:)));
title('Covariance of X in velocity');
```











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clear;clc;

Wind Model

```
[numb, denb]=butter(2,.05);
t=[0:.5:200]';
[T,~] = size(t);
Qs_det=1.3;
u_w = randn(T,1)*Qs_det;
wi=filter(numb,denb,u_w); % wi is the output of wind velocity.
numb=numb(1:2); % This is the delay process, making the Ds zero.
[As,Bs,Cs,Ds]=tf2ss(numb,denb); % The wind system As,Bs,Cs,Ds.
Wind_force_ratio = 10; %This is a coefficient of relation between wind velocity and force.
```

Cart Model

```
[Ac,Bc,Cc,Dc]=ssdata(c2d(ss([0 1;0 -.2],
[0;1],eye(2),zeros(2,1)),.5)); % This is the cart model given by the
prof.
%Gc=[1;10];
Gc=[0;1];
Motor = sin(2*pi*t/10);
```

Combined Model

[Xs;Xc] is 4x1; F=[As 0; Gc*Cs Ac] is 4x4; G=[0;Bc] is 4x1; B1=[Bs;Gc*Ds] is 4x1; H=[0 Cc] is 2x4; J=[0;0]; B2=[1;1]; X=F*X+G*u+B1*process_noise_wind; Y=H*X+J*u+B2*measurement_noise_cart;

```
%F = [As/Wind_force_ratio, zeros(2); Gc*Cs, Ac];
F = [As/Wind_force_ratio, zeros(2); Gc*Cs/Wind_force_ratio, Ac];
G = [zeros(2,1);Bc];
B1 = [Bs zeros(2,3);zeros(2,3) Gc*Ds]/Wind_force_ratio;
H = [zeros(2), Cc];
J = [0;0];
B2 = eye(2);
```

Compute the real data

```
X0 = [0.5; 0.5; 2; 1];
```

```
% The measurement noise v is a 2x1 vector, the first element
represents the
% postion and the second element represents the velocity.
% v = [v_pos; v_vel];
% To get a v.
v_{pos} = 0.05;
v vel = 0.05;
v_all = [v_pos; v_vel];
% To realize these measurement noise.
v_pos_real = randn(1,T)*sqrt(v_pos);
v_vel_real = randn(1,T)*sqrt(v_vel);
v all real = [v pos real; v vel real];
% The process noise is actually the noise in wind model, which is a
randn
% number.
Qs=3;
wind process noise = randn(2,T)*Qs;
wind_meas_noise = randn(2,T);
d = [wind_process_noise; wind_meas_noise];
Ro = diag([v pos, v vel]);
Qs = 5;
Qo = diag([Qs, 0, 1e-5, 1e-5]);
[L, \sim, \sim] = dlgr(F, G, Qo, v vel);
Fnew = F - G*L;
Gnew = zeros(4,1);
Hnew = H;
Jnew = J;
[X_ss, Y_ss] = myss(Fnew,Gnew,Hnew,Jnew,B1,B2,Motor,X0,v_all_real,d);
```

The Kalman Predictor process with the feed-back control

```
% We need to get Ro and Qo.
% Ro is the measurement noise matrix which is 2x2.
% Qo is the process noise matrix which is 4x4;
% In the combined model, the measurement noise only has v_pos and v_vel,
% thus the Ro should be diag([v_pos v_vel])
% Given that the process noise only has the wind noise Qs. Thus Qo should
% be diag([0,Qs,0,0]);
```

```
% Get X0 and P0
x0 = [0.5;0.5;2;1];
P0 = eye(4);

% Compute L
% Qc = [0,0,0,0;0,0,0,0,0,0,0,0,0];
% Rc = 0.003;

% Update the matrices

% Compute K
%K = dlqr(F,H',Qo,Ro);

[X_hatrec, Y_hatrec, K_rec, P_rec] =
myKalman(Fnew,Gnew,Hnew,Jnew,x0,Motor,Y_ss,P0,Qo,Ro);
```

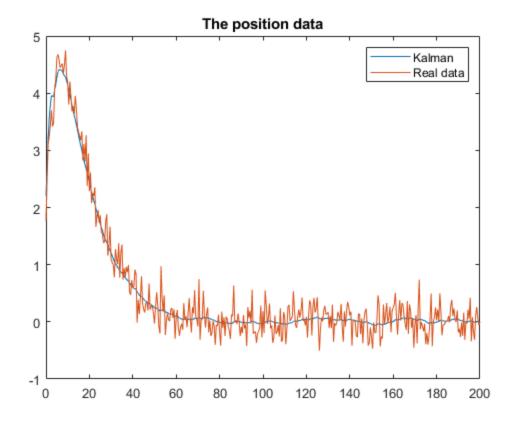
Plotting

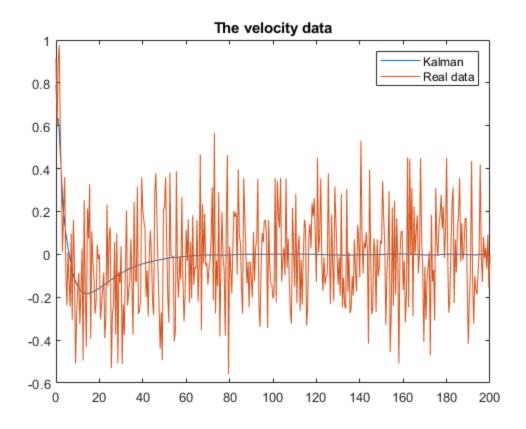
Now we have the real data y_ss and estimated Y_hatrec. figure(1); plot(t, Y_ss(1,:), t, Y_hatrec(1,:)); hold on; plot(t, Y_ss(2,:), t, Y_hatrec(2,:)); figure(2); plot(t, X_ss(3,:), t, X_hatrec(3,:)); hold on; plot(t, X_ss(4,:), t, X_hatrec(4,:));

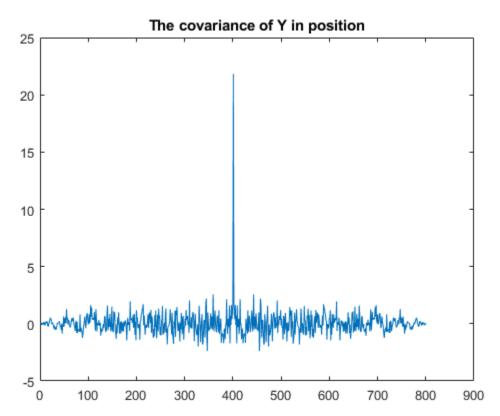
```
% figure(1);
% plot(t,X_hatrec(3,:));
% hold on;
% plot(t,Y_ss(1,:));
% hold off;
figure(1);
plot(t,Y_hatrec(1,:));
hold on;
plot(t,Y_ss(1,:));
hold off;
legend('Kalman','Real data');
title('The position data');
figure(2);
plot(t,Y_hatrec(2,:));
hold on;
plot(t,Y_ss(2,:));
hold off;
legend('Kalman','Real data');
title('The velocity data');
% figure(2);
% plot(t,X_hatrec(2,:));
% hold on;
% plot(t,Y_ss(2,:));
% hold off;
Y_tilda = Y_hatrec - Y_ss;
%Y_tilda_2_c=xcorr(Y_tilda(2,:)',Y_tilda(2,:)');
figure(3);
```

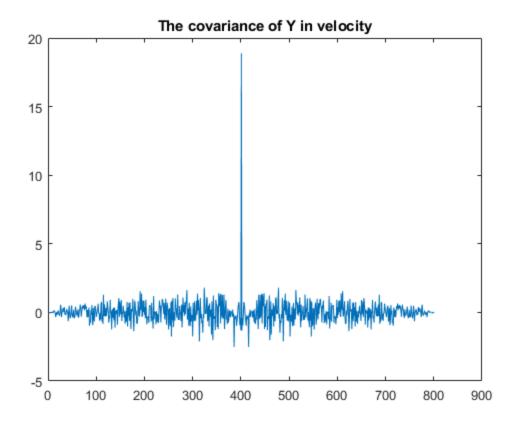
```
plot(xcorr(Y_tilda(1,:)));
title('The covariance of Y in position');
figure(4);
plot(xcorr(Y_tilda(2,:)));
title('The covariance of Y in velocity');

% X_tilda = X_hatrec - X_ss;
% %Y_tilda_2_c=xcorr(Y_tilda(2,:)',Y_tilda(2,:)');
% figure(5);
% plot(xcorr(X_tilda(3,:)));
% figure(6);
% plot(xcorr(X_tilda(4,:)));
```









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```
function [X_hatrec, Y_hatrec, K_rec, P_rec] =
myKalman(A,B,C,D,x0,u,y,P0,Qo,Ro)
Size = length(u);
X_hatrec = zeros(length(A), Size);
Y_hatrec = zeros(length(D), Size);
P_rec = zeros(Size, 2*length(A));
K_rec = zeros(Size, length(A));
X hat = x0;
Pp = P0;
for i = 1:Size
    X_{hatrec}(:,i) = X_{hat};
    P_{rec(i,:)} = [Pp(1,:), Pp(2,:)];
    K = A * Pp * C' / (C * Pp * C' + Ro);
    K_{rec(i,:)} = [K(1,:), K(2,:)];
    X_{hat} = (A - K * C) * X_{hat} + B * u(i) + K * y(:,i);
    Pp = A * Pp * A' - A * Pp * C' / (C * Pp * C' + Ro) * C * Pp * A'
 + Qo;
    Y_hatrec(:,i) = (C * X_hat);
end
end
Not enough input arguments.
Error in myKalman (line 3)
Size = length(u);
```

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Using the Lyapuna function $X = A X A^T + Q$. Xs, k+1 - Xs, k+1 = AsXs, k + Bows, k - [AsXs, k+B, ws, k] = As Xs, k + Bs Ws, k - As Xs, k - Bs Ws, k = As (Xs, k - Xs, k) + Bs LWs, k - Ws, le) E[(Xs, k+1 - Xs, k)(Xs, k+1 - Xs, k)]] =[As (Xs,k - Xs,k) + Bs (Ws,k- Ws,k)] = (Xs,k-Xs,k) As + (Ws, k - Ws,k) Bs $C_X = A_S C_X A^T + B \Omega B'$ Similarly, E([ys,k-ys,k][ys,k-ys,k])=[(s(ys,k-ys,k)+Ds(k,k-Vs,k)] = (ys,k - ys,k) Cs + (Vs,k - Vs,k) Ds T Ds=0. thus (y= Cs Cy Cs)

```
%Check observability
rank(obsv(F,H));
Ans = 4;
```

Thus this system is observable, means that it can build the initial state of wind and cart given the current condition.

Part iii

$$X_{s}, k-\tau_{t} = A_{c} X_{s}, k-\tau_{t} B_{s} W_{s}, k-\tau$$
 $X_{s}, k-\tau_{t} = A_{s} X_{s}, k-\tau_{t} + B_{s} W_{s}, k-\tau_{t}$
 $= A_{s} (A_{s} X_{s}, k-\tau_{t} + B_{s} W_{s}, k-\tau_{t}) + B_{s} W_{s}, k-\tau_{t}$
 $= A_{s}^{2} X_{s}, k-\tau_{t} + A_{s} B_{s} W_{s}, k-\tau_{t} + B_{s} W_{s}, k-\tau_{t}$
 $= A_{s}^{2} X_{s}, k-\tau_{t} + A_{s}^{2} B_{s} W_{s}, k-\tau_{t}$
 $X_{s}, k = A_{s}^{2} X_{s}, k-\tau_{t} + A_{s}^{2} B_{s} W_{s}, k-\tau_{t}$
 $E(X_{s}, k : X_{s}, k-\tau_{t}) = A_{s}^{2} E(X_{s}, k-\tau_{t} X_{s}^{2}, k-\tau_{t}) + A_{s}^{2} B_{s} E(W_{s}, k-\tau_{t}) \cdot X_{s}^{2}, k-\tau_{t}$
 $= A_{s}^{2} E(X_{s}, k-\tau_{t} X_{s}^{2}, k-\tau_{t}) + A_{s}^{2} E(X_{s}, k-\tau_{t} X_{s}^{2}, k-\tau_{t})$
 $= A_{s}^{2} E(X_{s}, k-\tau_{t} X_{s}^{2}, k-\tau_{t}) + A_{s}^{2} B_{s} W_{s}, k-\tau_{t}$
 $= A_{s}^{2} E(X_{s}, k-\tau_{t} X_{s}^{2}, k-\tau_{t}) + A_{s}^{2} B_{s} W_{s}, k-\tau_{t}$
 $= C_{s} (A_{s} X_{s}, k-\tau_{t} + C_{s} B_{s} W_{s}, k-\tau_{t}) + C_{s} A_{s}^{2} B_{s} W_{s}, k-\tau_{t}$
 $= C_{s} (A_{s}^{2} X_{s}, k-\tau_{t} + C_{s}^{2} B_{s} W_{s}, k-\tau_{t})$
 $= C_{s} (A_{s}^{2} X_{s}, k-\tau_{t} + C_{s}^{2} B_{s} W_{s}, k-\tau_{t})$
 $= C_{s} (A_{s}^{2} X_{s}, k-\tau_{t} + C_{s}^{2} B_{s} W_{s}, k-\tau_{t})$
 $= C_{s} (A_{s}^{2} X_{s}, k-\tau_{t} + C_{s}^{2} B_{s} W_{s}, k-\tau_{t})$

E(Ys, k. Ys, k-t) = (s E(Xs, k Xs, k-t) Cs

Wind:
$$X_{s,k+1} = A_s X_{s,k} + B_s W_{s,k}$$

 $Y_{s,k} = C_s X_{s,k} + D_s V_{s,k}$
Cort: $X_{c,k+1} = A_c X_{c,k} + B_c W_k + G_{rc} W_{s,k}$
 $Y_{c,k} = C_c X_{c,k} + V_{c,k}$
Combined model
$$\begin{bmatrix} X_s \\ X_c \end{bmatrix}_{k+1} = \begin{bmatrix} A_s & 0 \\ G_{rc}C_s & A_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + \begin{bmatrix} 0 \\ B_c \end{bmatrix} U_k + \begin{bmatrix} B_s & 0 \\ 0 & G_c P_s \end{bmatrix} \begin{bmatrix} W_{s,k} \\ V_{s,k} \end{bmatrix}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

$$Y_{c,k} = \begin{bmatrix} 0 & C_c \end{bmatrix} \begin{bmatrix} X_s \\ X_c \end{bmatrix}_k + V_{c,k}$$

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