## Design of kalman filter

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In [ ]:
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```
import numpy as np
import matplotlib.pyplot as plt
```

```
In [4]:
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```
# data
x0 = 4000 \# m
v0 = 280 \# m/s
#obserations(values from sensor)
x0 \text{ obs} = 4000 \# m
v0_obs = 280 \# m/s
x1_{obs} = 4260 \# m
v1 \text{ obs} = 282 \# m/s
x2 obs = 4550 # m
v2 obs = 285 # m/s
x3 \text{ obs} = 4860 \# m
v3_{obs} = 286 \# m/s
x4 obs = 5110 # m
v4 \text{ obs} = 290 \# m/s
#initial conditions
ax = 2 # m/sec^2
v0 = 280 # m/s
del t = 1 # sce
del x = 25 \# m
#process error in process(Covariance Matrix)
del p x = 20 \# m
del_p_vx = 5 \# m/sec
#obseravtion error
del x = 25 \# m
del vx = 6 \# m/sec
#noise and distrubance
wk = 0
Qk = 0
zk = 0
# Data For graph
predicted_value_position = [x0]
predicted_value_velocity = [v0]
measurement_value_position = [x0_obs,x1_obs,x2_obs,x3_obs,x4_obs]
measurement_value_velocity = [v0_obs,v1_obs,v2_obs,v3_obs,v4_obs]
kalman_value_position = [x0]
kalman value velocity = [v0]
```

## In [5]:

```
x_k1 = np.array([[x0],[v0]])

# The initial process covariance Matrix
# p_k = [(del_p_x)^2 del_p_x*del_p_vx; del_p_x*del_p_vx (del_p_vx)^2]
p_k = np.array([[(del_p_x * del_p_x), (del_p_x * del_p_vx)],[(del_p_x * del_p_vx)],
p_k[0,1] = 0
p_k[0,1] = 0

for t in range(1,5):

# The predicted state
```

```
\# x \ kp = A^*x \ kl + B^*u \ k + wk (u k = ax)
    x \text{ kp} = \text{np.dot}((\text{np.array}([[1, \text{del t}], [0,1]])), x \text{ k1}) + (\text{np.array}([[(\text{del t} * \text{del t})/2], [\text{del t}]])) *
ax) + wk
    predicted_value_position.append(x_kp[0])
    predicted value velocity.append(x kp[1])
# The predicted process covariance Matrix
\# p_k p = A * p_k * A' + Qk
    A = np.array([[1,del_t],[0,1]])
    p_kp = np.dot(A, (np.dot(p_k, A.transpose()))) + Qk
    p kp[0,1] = 0
    p kp[1,0] = 0
# Calcuate the kalman gain
\# k = (p kp*H')/(H*p kp*H' + R)
    H = np.array([[1,0],[0,1]])
    R = np.array([[25*25,0],[0,6*6]])
    \label{eq:kapha} $$k = \text{np.divide}((\text{np.dot}(p_kp, \text{H.transpose}())), (\text{np.dot}(H, \text{np.dot}(p_kp, \text{H.transpose}())) + R))$$
    k[0,1] = 0
    k[1,0] = 0
# The new observation
# yk = C*y_lm + zk
    c = np.array([[1,0],[0,1]])
    if t == 1:
        y_lm = np.array([[x1_obs],[v1_obs]])
    elif t == 2:
        y lm = np.array([[x2 obs],[v2 obs]])
    elif t == 3:
        y_lm = np.array([[x3_obs],[v3_obs]])
    elif t == 4:
        y_lm = np.array([[x4_obs],[v4_obs]])
    yk = np.dot(c, y lm) + zk
# Calculate the current state
\# xk = x kp + k(yk - H*x kp)
    xk = x_kp + np.dot(k, (yk - np.dot(H,x_kp)))
    x \ k1 = xk # xk will add with the first stage (1) the predicted state
    kalman_value_position.append(xk[0])
    kalman value velocity.append(xk[1])
# Update the process covariance matrix
# pk = (I - k*H)p kp
    I = np.array([[1,0],[0,1]])
    pk = np.dot((I - np.dot(k, H)), p kp)
    p k = pk
/srv/conda/envs/notebook/lib/python3.6/site-packages/ipykernel launcher.py:30: RuntimeWarning:
invalid value encountered in true divide
In [6]:
# position
print (predicted value position)
print(measurement_value_position)
print(kalman value position)
```

```
In [7]:
```

[4000, 4260, 4550, 4860, 5110]

```
# velocity
print(predicted_value_velocity)
print(measurement_value_velocity)
print(kalman_value_velocity)
```

[4000, array([4272.5]), array([4553.85054707]), array([4844.15764332]), array([5127.05898493])]

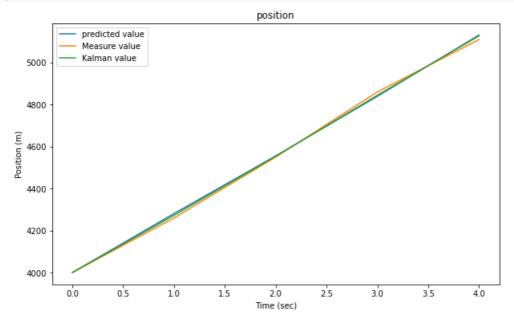
[4000, array([4281.]), array([4555.5]), array([4839.14124475]), array([5131.38286854])]

```
[280, array([282.]), array([284.]), array([286.29069767]), array([288.22522523])]
[280, 282, 285, 286, 290]
[280, array([282.]), array([284.29069767]), array([286.22522523]), array([288.55147059])]
```

## In [10]:

```
t = range(0,5)
# position

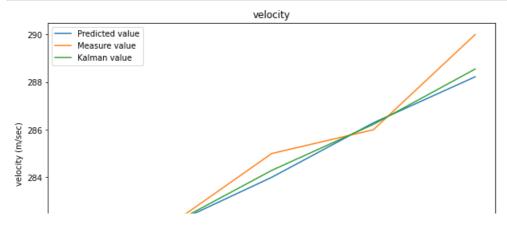
plt.plot(t, predicted_value_position, label = 'predicted value')
plt.plot(t, measurement_value_position, label = 'Measure value')
plt.plot(t, kalman_value_position, label = 'Kalman value')
plt.xlabel("Time (sec)")
plt.ylabel("Position (m)")
plt.title("position")
plt.legend()
plt.rcParams['figure.figsize'] = (10,6)
plt.show()
```

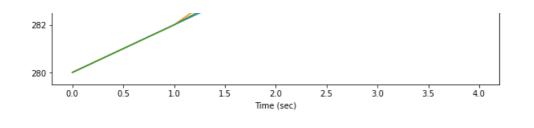


## In [9]:

```
# velocity

plt.plot(t,predicted_value_velocity, label = 'Predicted value')
plt.plot(t,measurement_value_velocity, label = 'Measure value')
plt.plot(t,kalman_value_velocity, label = 'Kalman value')
plt.xlabel("Time (sec)")
plt.ylabel("velocity (m/sec)")
plt.title("velocity")
plt.title("velocity")
plt.legend()
plt.rcParams['figure.figsize'] = (10,6)
plt.show()
```





In [ ]: