

overview

- noise reduction : digital filters, Kalman filter
- control : PID, LQR, MPC
- motor controll: park, clark transformations, real time PID
- planning, decision making: image segmentation, reinforcement learning

digital filters

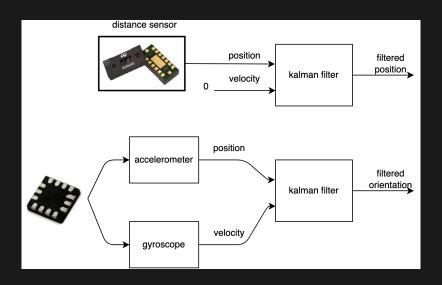
low pass filter (1st order)

$$y_n = (1 - a)y_{n-1} + ax_n$$
$$a = \frac{\Delta t}{RC + \Delta t}$$

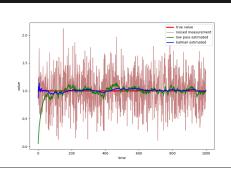
resonant filter (2nd order)

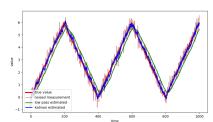
$$y_n = a_1 y_{n-1} + a_2 y_{n-2} + b_0 x_n + b_1 x_{n-1} + b_2 x_{n-2}$$

digital filters - usage



digital filters - low pass vs Kalman





1D kalman is simple

```
class KalmanFilter:
def init (self, dims = 1):
    self.x hat = torch.zeros(dims)
    self.p = 0.5*torch.ones(dims)
    z - position measurement
    dz - velocity measurement
    pz - position measurement uncertaininty
    pdz - velocity measurement uncertaininty
def step(self, z, dz, pz, pdz, dt = 1):
    self.x hat = self.x hat + dz*dt
    self.p = self.p + pdz*(dt**2)
    k = self.p/(self.p + pz)
    self.x_hat = self.x_hat + k*(z - self.x_hat)
    self.p = (1.0 - k)*self.p
    return self.x_hat
```

- 1, calibrate sensor variance (pz, pdz)
- 2, measure position z and velocity v
- 3, dynamics model

$$z_{n-1}=z_n+v_n dt$$

4, compute filter step

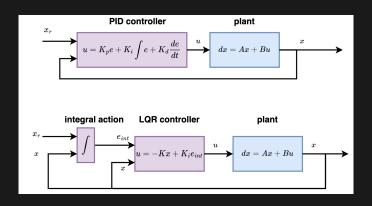
usage





- sensors de-noising
- sensors fusion (gyro + accelerometer), GPS
- navigation, error checking
- Apollo guidance computer (Apollo 8, 1968)
- Hakuto lander crash !!! (2023, April 2023)

controll: PID vs LQR



recommended sources

- book : Maxim Lapan, 2020, Deep Reinforcement Learning Hands-On second edition
- book : Enes Bilgin, 2020, Mastering Reinforcement Learning with Python
- youtuber : Yannic Kilcher, link
- youtuber : Two Minute Papers, link
- web : Paper With Code, link
- web : Intellabs, link



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