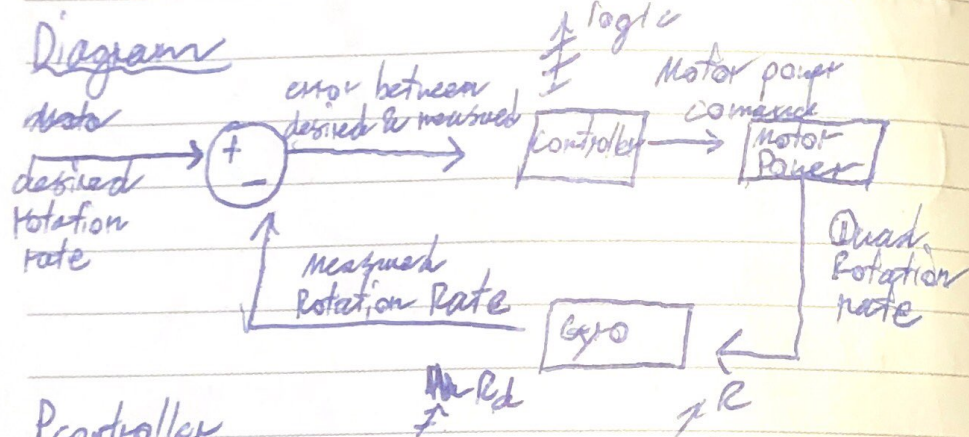


Proportional-Integral-Derivative

PID Control Loop Working



P controller

$$\text{Input}_{\text{motor}} = P \cdot (\text{Desired Rate} - \text{Rate})$$

$\rightarrow 1000 - 2000 \mu\text{s}$ $\{-75\% \rightarrow 75\% / 3\}$

$$\text{Error}(k) = R_d(n) - R(n)$$

$$\text{Input}_{\text{motor}}(k) = P \cdot (\text{Error}(n)) [\%]$$

$\rightarrow 250 \text{Hz}$

I controller

$$\text{Input}_{\text{motor}} = P \cdot \text{Error}(k) + I \int_0^{nT_s} \text{Error}(t) dt$$

$T_s = 0.001 \text{ s}$

Derivative

Discretize the integral

\rightarrow Past error \rightarrow current I average

$$I_{\text{term}}(k-1) + I \cdot (\text{Error}(k) + \text{Error}(k-1)) \cdot T_s$$

$$I_{\text{motor}}(n) = P \cdot \text{Error}(n) + I_{\text{term}}(k-1) + I \cdot I_{\text{term}}(k)$$

P shifts up \rightarrow shifts to desired

I removes oscillations

D Diminishes overshoot

$$\frac{F(x+h) + F(x)}{h}$$

D controller

$$Input_{motor} = P_{term}(k) + I_{term}(k) + \boxed{D_{term}(k)}$$

$$\hookrightarrow D \cdot \frac{d}{dt} Error(t) \rightarrow D \cdot \frac{(Error(n) - Error(n-1))}{T_s}$$

PID Controller

$$\cdot T_s = 950Hz \text{ or } 0.001S$$

$$\cancel{In_m} = \cancel{P(DesiredRate - Rate)} \rightarrow Error(n)$$

$$In_m = P \cdot (R_d(n) - R(n)) + I_{term}(n-1) + \frac{I \cdot (Error(n) + Error(n-1)) \cdot T_s}{2} + D \cdot \frac{(Error(n) - Error(n-1))}{T_s}$$

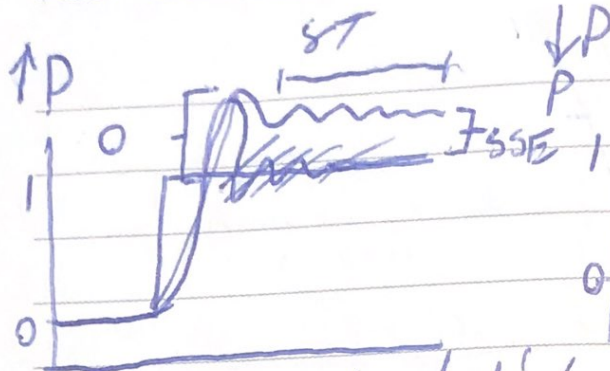
In Arduino...

$$Inp(n) = P \cdot Error_{prev} + P_{prev} + P_{prev} \cdot I_{term} + \frac{I \cdot (Error + PrevError) \cdot T_s}{2} + D \cdot (Error - PrevError)$$

T_s

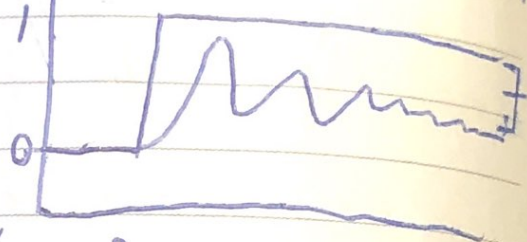
• Need this for Pitch Roll & Yaw

PID GRAPH



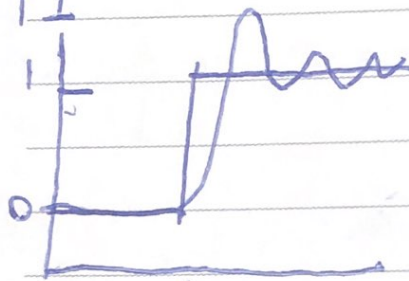
W - overshoot

ST = steady state time
SSSE = steady state error



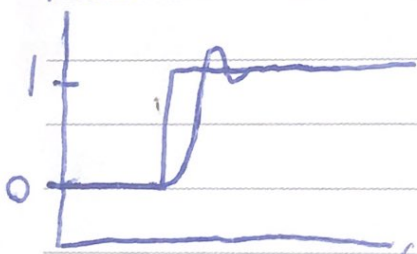
• P is overshoot, higher P means less settling time but more overshoot

PI



• I term removes steady state error

PID



• D less settling time

Kalman Filter

- Combines previous noisy data & current noisy data to make a state estimate of a system
- For example, we have a car going down the road @ a relatively consistent speed represented by:

$$\boxed{X_{t+1} = [A]X_t + U, \quad U \sim N(0, P)}$$

- where X_t is the cars position
- where $[A]$ is a numerical matrix
- where U is the noise factor dependent on a known covariance matrix P

- But say we have a GPS or device that gives us the cars position as well represented by:

$$\boxed{Y_t = [H]X_t + V, \quad V \sim N(0, Q)}$$

- where X_t is the cars position
- where H is a numerical matrix
- where V is the noise factor dependent on a known covariance matrix Q

• Kalman is a two step process

- Prediction step:

$$x_t \rightarrow (A) \rightarrow \bar{x}_{t+1} \quad \begin{matrix} \circ \text{Idealized} \\ \text{conversion} \end{matrix}$$

$$\Sigma_t \rightarrow (A) \rightarrow \Sigma_{t+1} = P \quad \begin{matrix} \circ \text{Noise as well} \\ \downarrow \\ P + A \Sigma_t + A^T \end{matrix}$$

- Update step

• combine noise & x_t estimation with y_t observation

• In order to do this, we need to determine which measurement we prefer/trust more by comparing covariance matrix P & Q between the two represented by:

$$K_x = \Sigma_{xx}$$

$$\Sigma_{xx} + Q_{xx}$$

- Where K_x is the Kalman value

- Σ_{xx} is the uncertainty matrix

represented by Σ_{t+1} in this example

- Q_{xx} is y_t covariance matrix

So we combine these two equations & determine our surprise factor

$$\begin{matrix} \bar{x}_{t+1} \\ \bar{\Sigma}_{t+1} \end{matrix} \rightarrow \begin{matrix} \downarrow \Delta x \\ \text{update} \end{matrix} \rightarrow x_{t+1}^+ = \bar{x}_{t+1} + K_x \Delta x$$

- Where Δx is our surprise factor
(Difference between two measurements)

• But now we want to determine the car's speed from its position?

• Using the covariance matrix:

$$\Sigma = \begin{bmatrix} \Sigma_{xx} & \Sigma_{x\dot{x}} \\ \Sigma_{x\dot{x}} & \Sigma_{\dot{x}\dot{x}} \end{bmatrix} \begin{matrix} \text{Mean Uncertainty} \\ \text{about state of system} \end{matrix}$$

- Σ_{xx} Uncertainty about position of car
- $\Sigma_{\dot{x}\dot{x}}$ Uncertainty about velocity of car
- $\Sigma_{x\dot{x}}$ Represent correlation between noise in measurement to the speed & position of the car

So to determine \hat{x} all we have to do is change the Kalman coefficient accordingly:

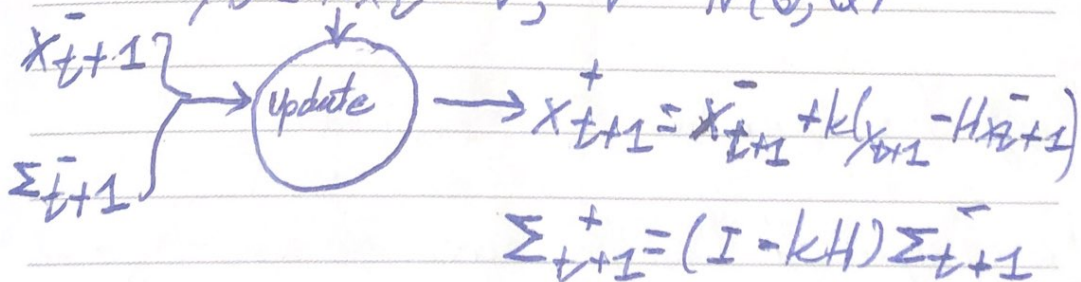
$$\hat{x}_{t+1}^+ = \hat{x}_{t+1}^- + k_t^i A x$$

Where $k_t^i = \frac{\Sigma x x^i}{\Sigma x x + Q x x}$

More generally:

$$x_{t+1} = A x_t + U, \quad U \sim N(0, P)$$

$$y_t = H x_t + V, \quad V \sim N(0, Q)$$



Measuring Angles with MPU-6050

- This is for more accurate measurement of yaw acceleration & height.
- This is because right now the flight controller responds to angles on a per second ratio, wanna change this to 10° pitch = 10° on joystick

Math \rightarrow degree/s

$$\text{Angle Pitch} = \int_0^{kT_s} \text{Ratepitch} \cdot dt$$

\rightarrow $\text{Anglepitch} = \text{degree} \ln(kT_s) - \text{degree} \ln(0)$

- Can't represent this so need to use logic

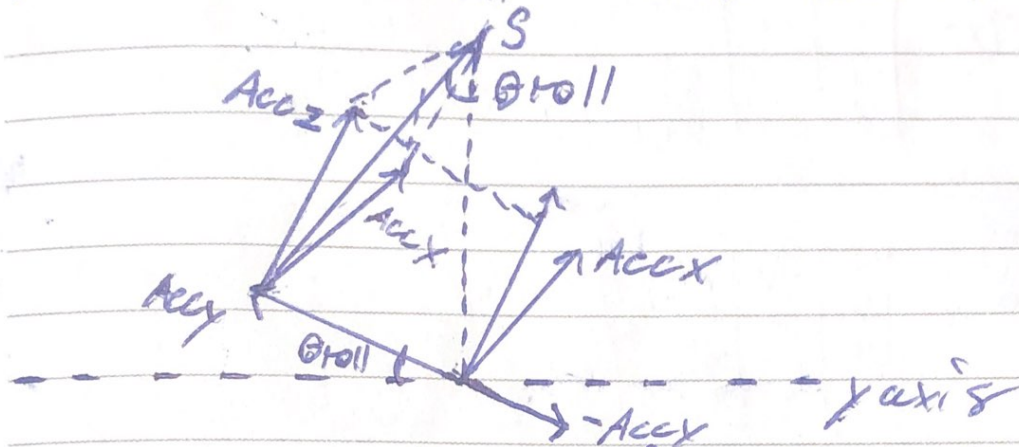
$\rightarrow \text{AnglePitch}(k) = \text{AnglePitch}(k-1) + \text{Ratepitch}(k) \cdot T_s$

\rightarrow Previous + current

- However this will keep dragging measurement errors

- ~~With new accels~~ Also doesn't allow us to account for Pitch & Roll changes with Yaw @ same time

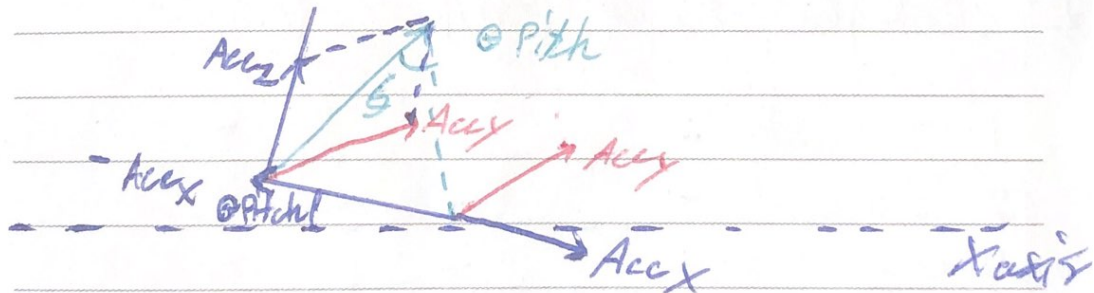
- Will use accelerometer to counter this
- More Math - Roll around the x axis



• So $\tan(\theta_{roll}) = \frac{Accy}{S}$, $S = \sqrt{Accx^2 + Accz^2}$

↳ $\tan(\theta_{roll}) = \frac{Accy}{\sqrt{Accx^2 + Accz^2}} \rightarrow \theta_{roll} = \tan^{-1}\left(\frac{Accy}{\sqrt{Accx^2 + Accz^2}}\right)$

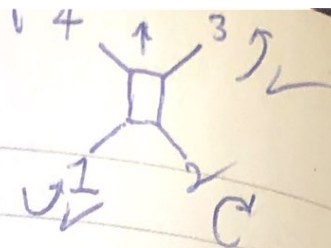
- More Math - Pitch Around the Y Axis



• So $\tan(\theta_{pitch}) = \frac{-Accx}{S}$

↳ $\theta_{pitch} = \tan^{-1}\left(\frac{-Accx}{\sqrt{Accx^2 + Accz^2}}\right)$

Steering DRONE



Throttle

$$TU = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix} \rightarrow \begin{bmatrix} 75\% \\ 75\% \\ 75\% \\ 75\% \end{bmatrix}$$

Right

$$ROLL = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix} \rightarrow \begin{bmatrix} 85\% \\ 75\% \\ 75\% \\ 25\% \end{bmatrix}$$

Pitch
~~DOWN~~

$$= \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix} \rightarrow \begin{bmatrix} 75\% \\ 75\% \\ 25\% \\ 25\% \end{bmatrix}$$

Yaw
CCW

$$= \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix} \rightarrow \begin{bmatrix} 25\% \\ 75\% \\ 25\% \\ 75\% \end{bmatrix}$$

LINEAR COMBINATION

1 = Throttle	-	-	+	-
2 = Throttle	+	-	+	+
3 = Throttle	+	-	-	-
4 = Throttle	-	-	-	+

Updated Flight Controller

- Current one is a degree per second response for pitch Roll & Yaw
- i.e., if we hold pitch @ 45° it will continue to pitch $45^\circ/s$
so @ $t=1$, 45°
 $t=2$, 90°
 $t=$ $135^\circ \dots$

- But we want it to be, hold stick @ 45° quadcopter stays @ 45°

Updated PID Loop

