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Spaceryde Software Design Challenge Calculations

Equations of motion

X direction:

$$D = \frac{1}{2}C_D P A(V_a)^2$$

$$F_x = \frac{-D(V_x - V_w)}{V_x}, \quad a_x = \frac{-D}{m*V_x}(v_x - V_w), \quad \frac{dv_x}{dt} = a_x$$

$$\frac{dv_{x}}{dt} + \frac{D}{m*v_{a}}(v_{x} - v_{w}) = 0$$

$$\frac{dv_{x}}{ds} + \frac{D}{m*v_{a}^{2}}(v_{x} - v_{w}) = 0 , s \leftarrow arc \ length \ of \ trajectory, \ thus \ \frac{ds}{dt} = v_{a}$$

$$\frac{dv_{x}}{dz} + \frac{D}{m*v_{a}^{2}}(v_{x} - v_{w}) = 0 , \ dz = ds * cos(\Theta), \Theta \ must \ be \ small, \ so \ dz \approx ds$$

$$\frac{dv_{x}}{dz} + \frac{C_{D}PA}{2m}(v_{x} - v_{w}) = 0$$

Using initial condition of $V_x(0) = 0$:

$$Vx = -V_w e^{\frac{C_D PA*z}{2m}} + V_w + V_{xi}$$
, $V_{xi} \leftarrow initial \ velocity \ of \ balloon$, is used when $z > 1000$

$$x = \int V_x dz = V_w e^{\frac{-C_D P A * z}{2m}} * \frac{2m}{C_D P A} + V_w * z - V_w * \frac{2m}{C_D P A} + V_{xi} * z$$

$$a_X = \frac{d}{dz}(V_X) = \frac{V_w C_D P A e^{\frac{-C_D P A * z}{2m}}}{2m}$$

Z direction:

$$F_{y} = F_{up} - mg - F_{Dz} = F_{up} - mg - \frac{1}{2}C_{D}PA(V_{z})^{2}$$

$$\frac{dV_{y}}{dt} - \frac{F_{up} - mg - \frac{1}{2}C_{D}PA(V_{z})^{2}}{m} = 0$$

Using initial condition of $V_z(0) = 0$

$$V_z = \frac{\sqrt{F_{up} - mg} * tanh(\frac{\sqrt{F_{up} - mg} * \sqrt{\frac{1}{2}C_D PA * t}}{\sqrt{\frac{1}{2}C_D PA}})}{\sqrt{\frac{1}{2}C_D PA}}$$

$$z = \int V_z dt = \frac{m*log(cosh(\frac{\sqrt{Fup-mg}*\sqrt{\frac{1}{2}C_DPA*t}}{m}))}{\frac{1}{2}C_DPA}$$

$$a_z = \frac{d}{dz}(V_z) = \frac{(F_{up} - mg) * sech^2(\frac{\sqrt{F_{up} - mg} * \sqrt{\frac{1}{2}C_D P A} * t}{m})}{m}$$

Kalman filter

 $A \leftarrow state transition matrix$

 $B \leftarrow control\ matrix$

 $U \leftarrow control\ vector$

 $Q \leftarrow process variance$

 $R \leftarrow measurement variance$

 $z \leftarrow measurement\ vector\ (raw\ GPS\ readings)$

 $H \leftarrow observation model$

 $w \leftarrow white noise vector$

 $X \leftarrow state\ vector$

 $P \leftarrow covariance error matrix$

 $K \leftarrow Kalman gain$

Matrices to use in filter:

$$Q = \begin{pmatrix} 5^2 & 0 \\ 0 & 5^2 \end{pmatrix} R = \begin{pmatrix} 60^2 & 0 \\ 0 & 60^2 \end{pmatrix} z = \begin{pmatrix} p \\ v \end{pmatrix} H = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} P_i = \begin{pmatrix} 1000 & 0 \\ 0 & 1000 \end{pmatrix}$$

Predict stage:

$$\frac{1}{x_f} = x_i + v\Delta t + \frac{1}{2} * a * \Delta t^2$$

$$v_f = v_i + a\Delta t$$

$$\begin{pmatrix} x \\ v \end{pmatrix} = \begin{pmatrix} 1 & \Delta t \\ 0 & 1 \end{pmatrix} \cdot preState + \begin{pmatrix} \frac{1}{2}\Delta t^2 \\ \Delta t \end{pmatrix} \begin{pmatrix} a \end{pmatrix} + w \quad w = \begin{pmatrix} 5^2 \\ 0 \end{pmatrix}$$

state

A previ

previous state

U

Predict state:

$$state = A * preState + BU + w$$

Predict error covariance:

$$P_f = A * P_i * A^T + Q$$

Update stage:

Get kalman gain:

$$K = P * H^{T} (H * P * H^{T} + R)^{-1}$$

Update state:

$$state = prevState + K * (z - H * prevState)$$

Update error covariance:

$$P_f = P_i - K * H * P_i$$

^{*} Kalman filter updates at 1 hz, getting the gps info first, then waits for the accelerometer data.