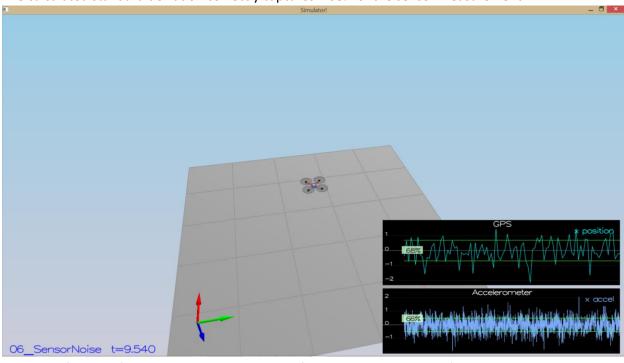
Building an Estimator

Provide a Writeup / README that includes all the rubric points and how you addressed each one. You can submit your writeup as markdown or pdf.

a) You are reading the writeup ©

Determine the standard deviation of the measurement noise of both GPS X data and Accelerometer X data.

a) The calculated standard deviation correctly captures ~ 68% of the sensor measurement.



c) Took data from log file and exported it to Microsoft Excel. I then used Excel's standard deviation formula (sample) to calculate standard deviation given simulated sensor measurements. Microsoft Excel uses the following Standard deviation formula:

$$s_{x} = \sqrt{\frac{\sum_{i=1}^{n}(x_{i} - \bar{x})^{2}}{n-1}}$$

b)

e) Where n = number data observations, X_i is the ith data observation, X_bar is the mean of all data in sample.

Implement a better rate gyro attitude integration scheme in the UpdateFromIMU() function.

- a) The improved integration scheme resulted in an attitude estimator of less than 0.1 rad for 3 seconds. (The max absolute error seems to be 0.015 rad based on a visual inspection of the graphs).
- b) I created a Euler Rate Matrix based on the current Euler angles then converted the body rates to Euler rates by multiplying the matrix with the body rates. I then integrated that to update the current Euler angles. Euler Rate Matrix formula and code shown below:

$$[E'_{123}(\phi, \theta, \psi)]^{-1} = \frac{1}{c_{\theta}} \begin{bmatrix} c_{\theta} & s_{\phi}s_{\theta} & c_{\phi}s_{\theta} \\ 0 & c_{\phi}c_{\theta} & -s_{\phi}c_{\theta} \\ 0 & s_{\phi} & c_{\phi} \end{bmatrix}.$$

Implement all of the elements of the prediction step for the estimator.

a) The prediction step includes the state update element (PredictState() function) as seen in code below:

c) As can be seen implemented in the code below:

b)

a. Calculated the Rgb prime matrix by taking the partial derivatives wrt to yaw angle.

$$R'_{bg} = \begin{bmatrix} -\cos\theta\sin\psi & -\sin\phi\sin\theta\sin\psi - \cos\phi\cos\psi & -\cos\phi\sin\theta\sin\psi + \sin\phi\cos\psi \\ \cos\theta\cos\psi & \sin\phi\sin\theta\cos\psi - \cos\phi\sin\psi & \cos\phi\sin\theta\cos\psi + \sin\phi\sin\psi \\ 0 & 0 & 0 \end{bmatrix}$$

- b. The acceleration is accounted for as a command in the calculation of gPrime;
 VectorXf RbgPrime_u = RbgPrime * accel_vec;
- c. and the covariance update follows the classic EKF update equation: ekfCov_p = gPrime * ekfCov * gPrime_T + Q (for prediction) and ekfCov = (eye K*H)*ekfCov

```
MatrixXf QuadEstimatorEKF::GetRbgPrime(float roll, float pitch, float yaw)
                             MatrixXf RbgPrime(3, 3);
                             RbgPrime.setZero();
                            RbgPrime(1, 1) = sin(phi)*sin(theta)*cos(psi) - 1 * cos(phi)*sin(psi);
RbgPrime(1, 2) = cos(phi)*sin(theta)*cos(psi) + sin(phi)*sin(psi);
RbgPrime(2, 0) = 0;
                            RbgPrime(2, 2) = 0;
d)
                       ■void QuadEstimatorEKF::Predict(float dt, V3F accel, V3F gyro)
                           VectorXf newState = PredictState(ekfState, dt, accel, gyro);
                           MatrixXf RbgPrime = GetRbgPrime(rollEst, pitchEst, ekfState(6));
                           MatrixXf gPrime(QUAD_EKF_NUM_STATES);
                           gPrime.setIdentity();
                           MatrixXf ekfCov_p(QUAD_EKF_NUM_STATES, QUAD_EKF_NUM_STATES);
                284
                           VectorXf accel_vec(3);
                           accel_vec(0) = accel[0];
accel_vec(1) = accel[1];
accel_vec(2) = accel[2];
                           VectorXf RbgPrime_u = RbgPrime * accel_vec;
                           gPrime(0, 3) = dt;
                           gPrime(1, 4) = dt;
                          gPrime(3, 6) = RbgPrime_u(0) * dt;
gPrime(4, 6) = RbgPrime_u(1) * dt;
                           gPrime(5, 6) = RbgPrime_u(2) * dt;
                           MatrixXf gPrime_T = gPrime.transpose();
ekfCov_p = gPrime * ekfCov * gPrime_T + Q;
e)
```

Implement the magnetometer update.

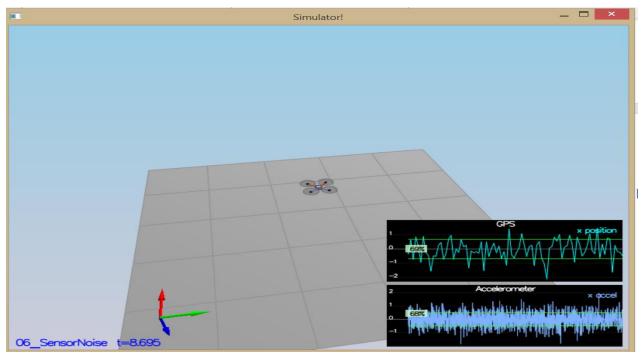
a) Incorporated magnetometer data into state and normalized the difference between current state and magnetometer value to ensure it takes the shorter way around.

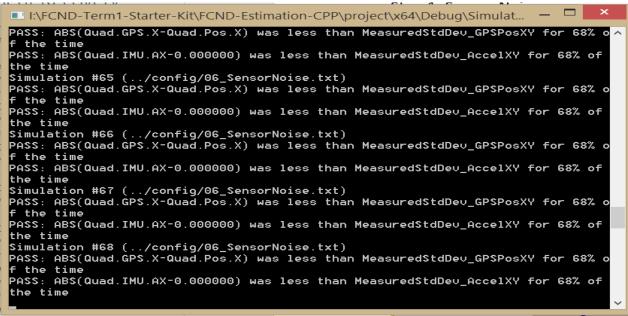
Implement the GPS update.

The estimator should correctly incorporate the GPS information to update the current state estimate as seen in image below:

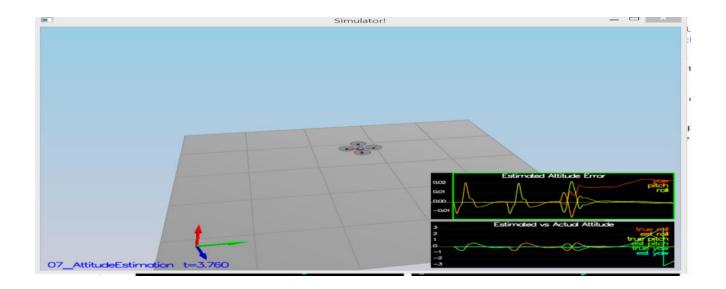
Meet the performance criteria of each step.

Step 1: Sensor Noise: standard deviations should accurately capture the value of approximately 68% of the respective measurements. This criterion was met as seen in images below:



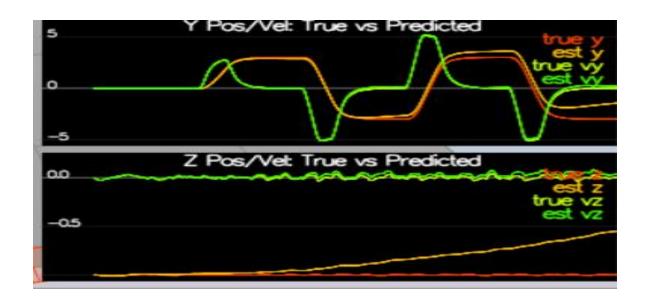


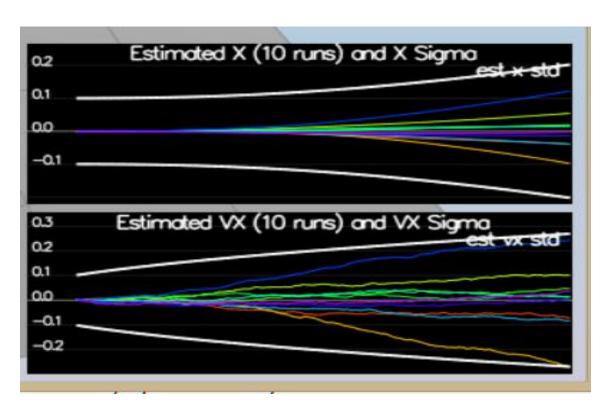
Step 2: Attitude Estimation : our attitude estimator needs to get within 0.1 rad for each of the Euler angles for at least 3 seconds. The criterion was met as can be seen in images below:



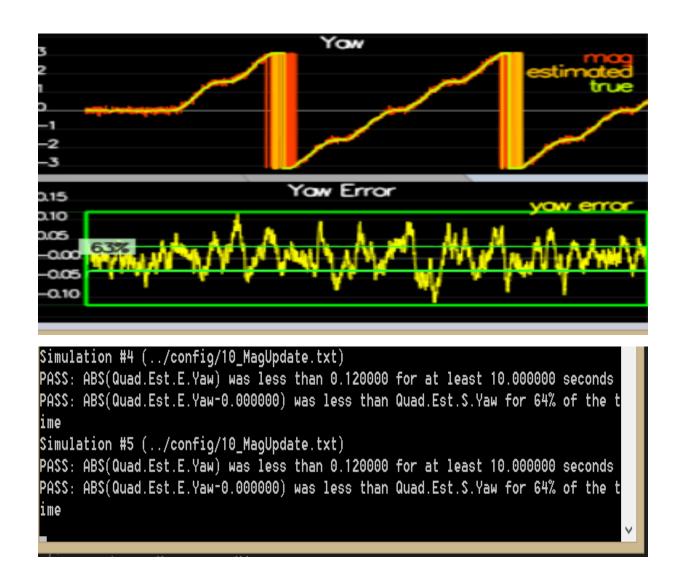
```
nds
Simulation #86 (../config/07_AttitudeEstimation.txt)
PASS: ABS(Quad.Est.E.MaxEuler) was less than 0.100000 for at least 3.000000 seconds
Simulation #87 (../config/07_AttitudeEstimation.txt)
PASS: ABS(Quad.Est.E.MaxEuler) was less than 0.100000 for at least 3.000000 seconds
Simulation #88 (../config/07_AttitudeEstimation.txt)
PASS: ABS(Quad.Est.E.MaxEuler) was less than 0.100000 for at least 3.000000 seconds
Simulation #89 (../config/07_AttitudeEstimation.txt)
PASS: ABS(Quad.Est.E.MaxEuler) was less than 0.100000 for at least 3.000000 seconds
```

Step 3: Prediction Step: This step doesn't have any specific measurable criteria being checked. However moderate/slow drift can be seen between true state and estimated state and covariance appears to grow appropriately like the data. Please see images below:

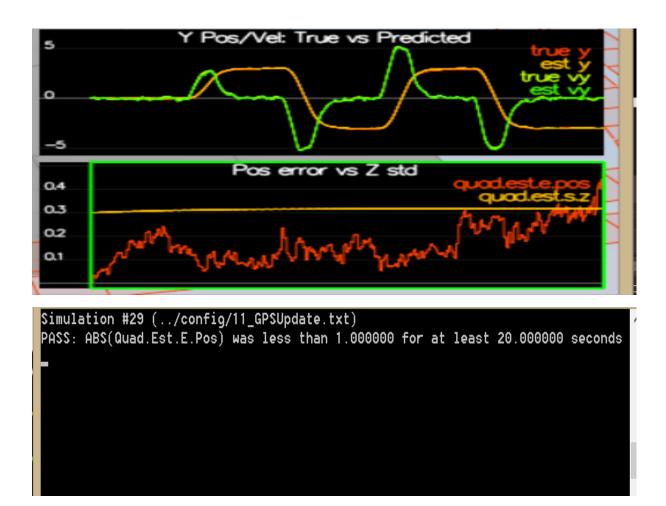




Step 4: Magnetometer Update: Your goal is to both have an estimated standard deviation that accurately captures the error and maintain an error of less than 0.1rad in heading for at least 10 seconds of the simulation. This goal was met as can be seen in the images below:



Step 5: Closed Loop + GPS Update: Your objective is to complete the entire simulation cycle with estimated position error of < 1m. The criterion was met as seen in images below:

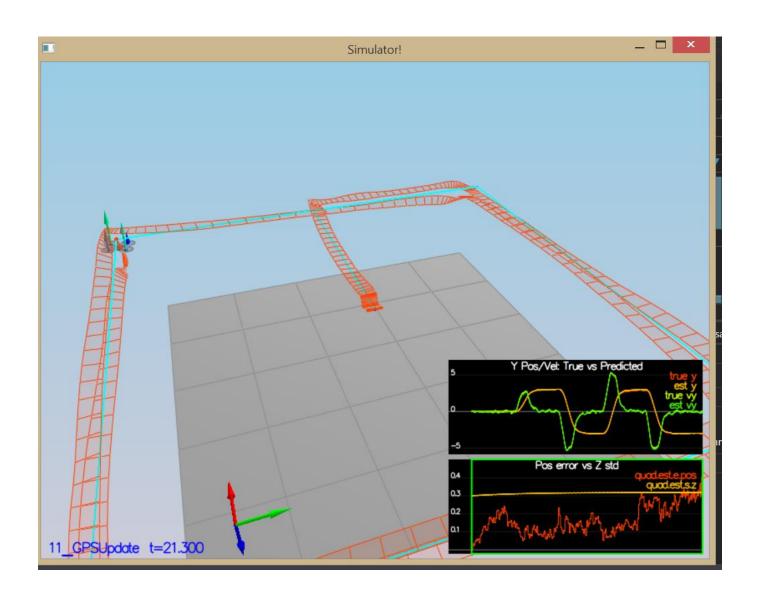


De-tune your controller to successfully fly the final desired box trajectory with your estimator and realistic sensors.

a) De-tuned controller gains and was able to successfully fly the entire box with <1 m error for box flight

```
10
     # Physical properties
     Mass = 0.5
     L = 0.17
     Ixx = 0.0023
     Iyy = 0.0023
     Izz = 0.0046
     kappa = 0.016
     minMotorThrust = .1
     maxMotorThrust = 4.5
     # Position control gains 36 , 60 , and 25
     kpPosXY = 2
kpPosZ = 10
KiPosZ = 1
22
23
24
     # Velocity control gains 15 and 25
27
28
     kpVelXY = 10
kpVelZ = 1
     # Angle control gains 8.3 and 1.8
31
     kpBank = 10.3
32
    kpYaw = 20
     # Angle rate gains 61.5, 25.6, 5
35 kpPQR = 60, 60, 2
```

b)



Simulation #7394 (../config/11_GPSUpdate.txt)

PASS: ABS(Quad.Est.E.Pos) was less than 1.000000 for at least 20.000000 seconds Simulation #7395 (../config/11_GPSUpdate.txt)

PASS: ABS(Quad.Est.E.Pos) was less than 1.000000 for at least 20.000000 seconds Simulation #7396 (../config/11_GPSUpdate.txt)

PASS: ABS(Quad.Est.E.Pos) was less than 1.000000 for at least 20.000000 seconds Simulation #7397 (../config/11_GPSUpdate.txt)

PASS: ABS(Quad.Est.E.Pos) was less than 1.000000 for at least 20.000000 seconds