

Design of advanced resolver-to-digital converters

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Introduction

What can be done when the sensor yields two signals proportional to the sine and cosine of the shaft position ?



This happens

- when a resolver is used
- when a magnetic position encoder is used
- when injecting high frequency voltages into salient PMSM machines for sensorless control

This paper both recalls the classical solution, with its **possible settings and performances**, and **two new estimators**.

The classical angle tracking observer

The classical solution is based on a second-order state space model with a “nearly constant speed” (the acceleration is considered as a random noise) and a linear measurement of the position.

- State space model :

$$\dot{X}(t) = A_1 X(t) + G_1 \alpha(t), \text{ with } X = \begin{pmatrix} \theta \\ \Omega \end{pmatrix},$$

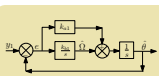
$$y_1(t) = C_1 X(t) + w(t), \text{ with } \Omega = \dot{\theta}$$

$$A_1 = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, C_1^T = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \text{ and } G_1 = \begin{pmatrix} 0 \\ 1 \end{pmatrix},$$

- Luenberger observer

$$\begin{aligned} \dot{\hat{X}}(t) &= A_1 \hat{X}(t) + K_{c1} e(t), \quad \text{with } e(t) = y_1(t) - C_1 \hat{X}(t) = y_1(t) - \hat{\theta}(t) \\ \text{with } K_{c1}^T &= (k_{a1} \quad k_{b1}) \text{ and } \hat{X}^T = \begin{pmatrix} \hat{\theta} & \hat{\Omega} \end{pmatrix} \end{aligned}$$

The classical angle tracking observer



The resulting position and speed estimator is equivalent to :

- **an integrator in closed-loop with a PI controller**

$$K_p = k_{a1} \quad \text{and} \quad T_i = k_{a1}/k_{b1}$$

a common structure in automatic control

- **a filtering process applied to the measured position**

$$\hat{\Theta}(s) = \frac{k_{a1} s + k_{b1}}{s^2 + k_{a1} s + k_{b1}} Y_1(s)$$

$$\hat{\Omega}(s) = \frac{k_{b1}}{s^2 + k_{a1} s + k_{b1}} s Y_1(s)$$

$\hat{\Theta}(s)$ results from a 2nd-order lowpass filter with a slope of only -20 dB/dec

The classical angle tracking observer

Properties

The classical ATO is unbiased when the speed is constant **but biased when the speed is not constant.**

- if $Y_1(s) = \Theta(s) = \frac{\alpha}{s^2}$,

$$\lim_{t \rightarrow +\infty} y_1(t) - \hat{\theta}(t) = 0$$

- if $Y_1(s) = \Theta(s) = \frac{\alpha}{s^3}$,

$$\lim_{t \rightarrow +\infty} y_1(t) - \hat{\theta}(t) = \frac{\alpha}{k_{b1}}$$

The classical angle tracking observer

Possible settings of the ATO

The setting of the ATO parameters should be deduced **from desired performances in the time domain** rather than in the frequency domain.

For a linearly increasing speed with acceleration equal to α ,

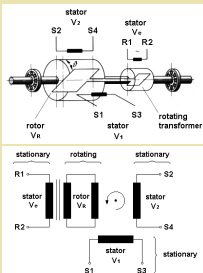
- $k_{b1} = \frac{\alpha}{\theta_{\text{true}} - \hat{\theta}}$ and $k_{a1} = \frac{\Omega_{\text{true}} - \hat{\Omega}}{\theta_{\text{true}} - \hat{\theta}}$
- $k_{b1} = \frac{\alpha}{\theta_{\text{true}} - \hat{\theta}}$ and $k_{a1} = 2m\sqrt{k_{b1}}$,

where m is a damping ratio. $m = 1.945$ provides a 5 % overshoot when the actual position abruptly changes from 0 to 180°. A **“Butterworth” setting** ($m = \sqrt{2}/2$) leads to an **overshoot of 20 %** !

The higher the accuracy, the longer the transients

The classical angle tracking observer

resolver :



source :

<http://data.bolton.ac.uk>.

nonlinear measurement equation

if the sensor provides two noisy signals

$$y_c(t) = \cos(\theta(t)) + w_c(t)$$

$$y_s(t) = \sin(\theta(t)) + w_s(t),$$

the error term $e(t) = y_1(t) - \hat{\theta}(t)$ should be **simply replaced** by

$$\begin{aligned} \epsilon(t) &= y_s(t) \cos(\hat{\theta}(t)) - y_c(t) \sin(\hat{\theta}(t)) \\ &= \sin(\theta(t) - \hat{\theta}(t)) + w_s(t) \cos(\hat{\theta}(t)) - w_c(t) \sin(\hat{\theta}(t)) \\ &\approx \theta(t) - \hat{\theta}(t) + w_s(t) \cos(\hat{\theta}(t)) - w_c(t) \sin(\hat{\theta}(t)) \end{aligned}$$

The resulting observer **remains stable** [Harnefors2000].

A new continuous-time third-order ATO

Another angle tracking observer can be derived from on a third-order state space model with a “nearly constant acceleration” (the jerk is considered as a random noise) and a linear position measurement.

- State space model :

$$\dot{X}(t) = A_2 X(t) + G_2 \beta(t), \text{ with } X = \begin{pmatrix} \theta \\ \Omega \\ \alpha \end{pmatrix}$$

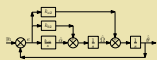
$$y_1(t) = C_2 X(t) + w(t),$$

$$A_2 = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}, C_2^T = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \text{ and } G_2 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix},$$

- Luenberger observer

$$\begin{aligned} \dot{\hat{X}}(t) &= A_2 \hat{X}(t) + K_{c2} e(t), \quad \text{with } e(t) = y_1(t) - C_2 \hat{X}(t) = y_1(t) - \hat{\theta}(t) \\ \text{with } K_{c2}^T &= (k_{a2} \quad k_{b2} \quad k_{c2}) \text{ and } \hat{X}^T = (\hat{\theta} \quad \hat{\Omega} \quad \hat{\alpha}) \end{aligned}$$

A new continuous-time third-order ATO



The resulting position, speed and acceleration estimator is equivalent to :

- **an double integrator in closed-loop with a PID controller**

$$K_p = k_{b2}, \quad T_i = \frac{k_{b2}}{k_{c2}} \quad \text{and} \quad T_d = \frac{k_{a2}}{k_{b2}}$$

still a common structure in automatic control

- **a third-order filtering process applied to the measured position**

$$\hat{\Theta}(s) = \frac{k_{a2} s^2 + k_{b2} s + k_{c2}}{s^3 + k_{a2} s^2 + k_{b2} s + k_{c2}} Y_1(s)$$

$$\hat{\Omega}(s) = \frac{(k_{b2} s + k_{c2})}{s^3 + k_{a2} s^2 + k_{b2} s + k_{c2}} s Y_1(s)$$

$$\hat{\alpha}(s) = \frac{k_{c2}}{s^3 + k_{a2} s^2 + k_{b2} s + k_{c2}} s^2 Y_1(s)$$

$\hat{\Theta}(s)$ results from a third-order lowpass filter with a slope of only -20 dB/dec

A new continuous-time third-order ATO

Properties

This estimator is unbiased when the speed is constant **but also when the speed is linearly increasing**

- if $Y_1(s) = \Theta(s) = \frac{\varepsilon}{s^2}$,

$$\lim_{t \rightarrow +\infty} y_1(t) - \hat{\theta}(t) = 0$$

- if $Y_1(s) = \Theta(s) = \frac{\alpha}{s^3}$,

$$\lim_{t \rightarrow +\infty} y_1(t) - \hat{\theta}(t) = 0$$

This estimator provides an improved position and speed estimation during transients.

A new continuous-time third-order ATO

Possible settings of this ATO

Placing the poles of the transfer functions at $-K/T$, $-1/T + \gamma\psi/T$ and $-1/T - \gamma\psi/T$ to obtain a desired settling time, a desired peak overshoot and a desired natural frequency of oscillation.

$$k_{a2} = \frac{K + 2}{T}, \quad k_{b2} = \frac{\psi^2 + 2K + 1}{T^2}, \quad k_{c2} = \frac{K(\psi^2 + 1)}{T^3}$$

For example, simulation results show that $\psi = 3\pi/2$ and $K = 39.04$ leads to a peak overshoot of 10 % when the actual position abruptly changes from 0 to 180° .

A “Butterworth” setting ($k_{a2} = 2/T_c$, $k_{b2} = 2/T_c^2$, $k_{c2} = 1/T_c^3$) would lead to an overshoot of 30.9 %!

The higher the accuracy, the longer the transients

A new discrete-time third-order ATO

Another angle tracking observer can be derived from on a discrete-time third-order state space model, using a **statistical state-space estimator**.

State space model : linear transition equation and nonlinear measurement equation

$$X[k+1] = A_3 X[k] + G v[k]$$

$$Y[k+1] = \begin{pmatrix} y_c[k+1] \\ y_s[k+1] \end{pmatrix} = \begin{pmatrix} y_c((k+1) T_s) \\ y_s((k+1) T_s) \end{pmatrix} = \mathcal{H}(X[k+1]) + W[k+1],$$

$$\text{with } X[k] = \begin{pmatrix} x_1[k] \\ x_2[k] \\ x_3[k] \end{pmatrix} = \begin{pmatrix} \theta(k T_s) \\ T_s \Omega(k T_s) \\ T_s^2 \alpha(k T_s) \end{pmatrix}, \quad A_3 = \begin{pmatrix} 1 & 1 & 1/2 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix},$$

$$G^T = (1/6 \quad 1/2 \quad 1), \quad v[k] = T_s^3 \beta(k T_s),$$

$$\mathcal{H}(X[k]) = \begin{pmatrix} \cos(x_1[k]) \\ \sin(x_1[k]) \end{pmatrix} \text{ and } W[k] = \begin{pmatrix} w_c[k] \\ w_s[k] \end{pmatrix}$$

A new discrete-time third-order ATO

Another angle tracking observer can be derived from on a discrete-time third-order state space model, using a **statistical state-space estimator**.

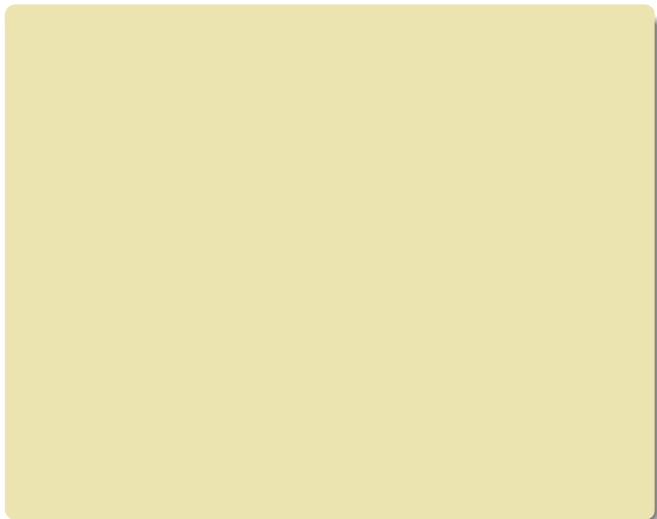
nonlinear Kalman estimator : using a third order Taylor expansion of the measurement function \mathcal{H} , a simple and computationally efficient **time-invariant** state estimator can be designed :

$$X_p[k] = A_3 X_e[k-1]$$

$$X_e[k] = X_p[k] + K_{\text{lin}} \left(y_s[k] \cos(\hat{\theta}_p[k]) - y_c[k] \sin(\hat{\theta}_p[k]) \right)$$

where K_{lin} is derived from the variances of the state noise q and the measurement noise r . This is a **new and very surprising result**, because “extended” Kalman filters are seldom simple.

Simulation Results



Conclusion

- Performances, setting, and digital implementation of the classical ATO
- Performances, setting, and digital implementation of a new continuous-time third-order ATO
- New discrete-time estimator based on estimation theory
- MATLAB/SIMULINK files available at <http://www.univ-nantes.fr/auger-f>

Conclusion

- Performances, setting, and digital implementation of the classical ATO
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- Thanks for your attention
 - Any questions ?