



 POLITECNICO DI MILANO



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Longitudinal dynamics: slip control

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"Single-corner" model

The following slides are adapted from those used in the course "Automazione nei mezzi di trasporto"

(M.Sc. course at the Politecnico di Milano, Prof. Sergio M. Savaresi)

$$\begin{cases} J\dot{\omega} = rF_x - T_b \\ m\dot{v} = -F_x \end{cases}$$

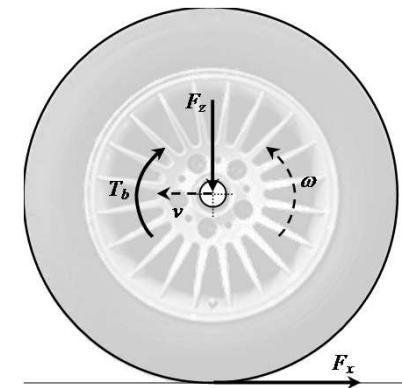
$$F_x = F_z \mu(\lambda, \alpha; \theta_r)$$

$$\lambda = (v - \omega r) / v = 1 - r\omega / v$$

$$\begin{cases} J\dot{\omega} = rF_z \mu\left(\frac{v - \omega r}{v}\right) - T_b \\ m\dot{v} = -F_z \mu\left(\frac{v - \omega r}{v}\right) \end{cases}$$

- ω : angular speed of the wheel ([rad/s]; $\omega > 0$ is assumed);
- v : longitudinal speed of the vehicle body;
- T_b : braking torque (control/input variable);
- F_x : longitudinal road-tire contact force;
- F_z : vertical road-tire contact force;
- J, m and r are the momentum of inertia of the wheel, the quarter-car mass, and the wheel radius

(e.g. $J = 1 \text{ Kg m}^2$, $m = 225 \text{ Kg}$, $r = 0.28 \text{ m}$).



SISO system, 2nd order ($n=2$), non-linear, time-invariant



Change of state variables

$$\begin{cases} J\dot{\omega} = rF_z\mu\left(\frac{v - \omega r}{v}\right) - T_b \\ m\dot{v} = -F_z\mu\left(\frac{v - \omega r}{v}\right) \end{cases} \quad \begin{aligned} \dot{\lambda} &= -\frac{r}{v}\dot{\omega} + \frac{r\omega}{v^2}\dot{v} \\ \omega &= \frac{v}{r}(1 - \lambda) \end{aligned}$$

CHANGE OF STATE VARIABLES

$$\begin{cases} \dot{\lambda} = -\frac{1}{v}\left(\frac{(1 - \lambda)}{m} + \frac{r^2}{J}\right)F_z\mu(\lambda) + \frac{r}{vJ}T_b \\ m\dot{v} = -F_z\mu(\lambda) \end{cases}$$

SISO system, 2nd order ($n=2$), non-linear, time-invariant, strictly proper (or:
SISO, non-linear, 1st order, time-varying)



Equilibrium points (I)

$$\begin{cases} \dot{\lambda} = -\frac{1}{v} \left(\frac{(1-\lambda)}{m} + \frac{r^2}{J} \right) F_z \mu(\lambda) + \frac{r}{vJ} T_b \\ m\dot{v} = -F_z \mu(\lambda) \end{cases}$$

$$\dot{\lambda} = 0, \quad \eta = \bar{\eta}$$

$$\text{not } \dot{v} = 0$$

Linear, normalized, wheel deceleration

$$\eta = -\frac{\dot{\omega}r}{g}$$

$$\bar{T}_b = F_z \left(r + \frac{J}{rm} (1 - \bar{\lambda}) \right) \mu(\bar{\lambda})$$

Note: DO NOT depend on v



Equilibrium points

$$\begin{cases} \dot{\lambda} = -\frac{1}{v} \left(\frac{(1-\lambda)}{m} + \frac{r^2}{J} \right) F_z \mu(\lambda) + \frac{r}{vJ} T_b \\ m\dot{v} = -F_z \mu(\lambda) \end{cases}$$

$$\dot{\lambda} = 0, \quad \eta = \bar{\eta}$$

not $\dot{v} = 0$

$$\lambda = 1 - r\omega / v$$

$$(\dot{\omega}v - \dot{v}\omega) / v^2 = 0$$

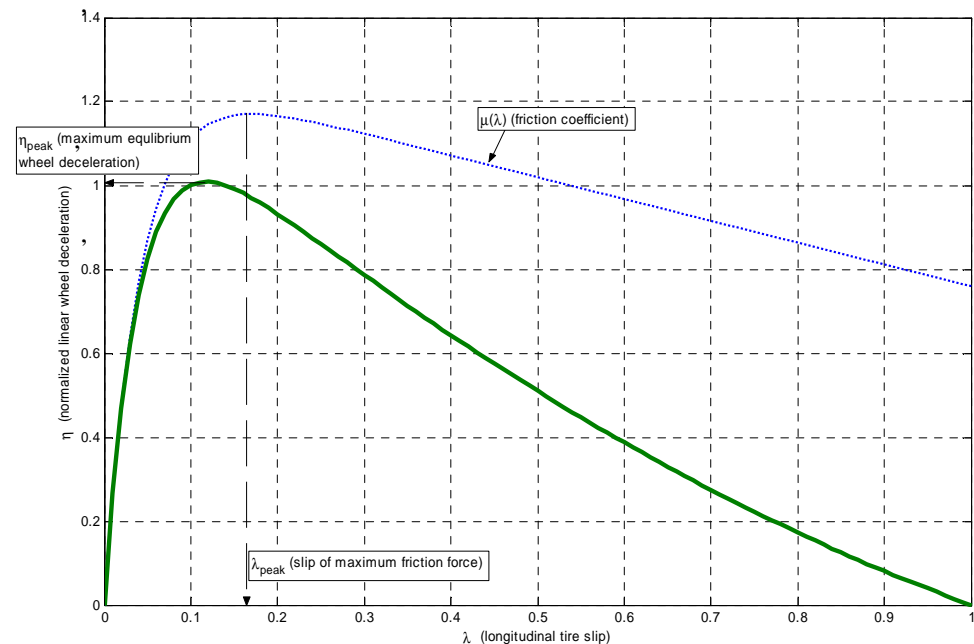
$$\dot{\omega} = \omega \dot{v} / v$$

$$\omega = v(1-\lambda) / r$$

$$m\dot{v} = -F_z \mu(\lambda)$$

$$\eta = -\dot{\omega}r / g$$

$$\bar{\eta} = (1 - \bar{\lambda}) \frac{F_z}{mg} \mu(\bar{\lambda})$$



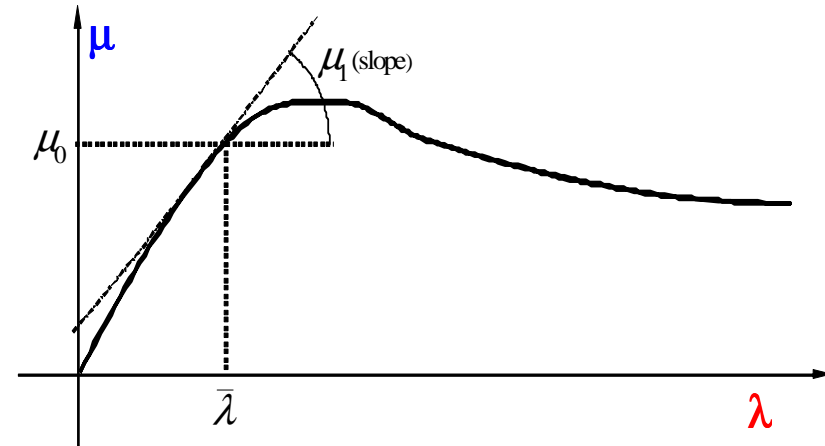


Linearization model (method I)

$$\delta T_b = T_b - \bar{T}_b \quad \delta \lambda = \lambda - \bar{\lambda}$$

$$\mu_1 = \mu_1(\bar{\lambda}) = \partial \mu(\lambda) / \partial \lambda|_{\lambda=\bar{\lambda}}$$

$$\mu_0 = \mu(\bar{\lambda})$$



$$\begin{cases} \dot{\lambda} = -\frac{1}{v} \left(\frac{(1-\lambda)}{m} + \frac{r^2}{J} \right) F_z \mu(\lambda) + \frac{r}{vJ} T_b \\ m\dot{v} = -F_z \mu(\lambda) \end{cases}$$

Assumption: v dynamics (Body Dynamics) are much slower than λ or ω dynamics (Wheel Dynamics).
Consider v as slowly varying.



Linearization model (method I)

Equation of λ :

$$\dot{\lambda} = -\frac{1}{v} \left(\frac{(1-\lambda)}{m} + \frac{r^2}{J} \right) F_z \mu(\lambda) + \frac{r}{vJ} T_b$$

$$\dot{\lambda} = -\frac{F_z}{v} \left(\frac{(1-\lambda)}{m} \mu(\lambda) + \frac{r^2}{J} \mu(\lambda) \right) + \frac{r}{vJ} T_b$$

Linearizing around $\bar{\lambda}, \bar{T}_b$

$$\delta \dot{\lambda} = -\frac{F_z}{v} \left(\frac{-1}{m} \mu_0 + \frac{(1-\bar{\lambda})}{m} \mu_1 + \frac{r^2}{J} \mu_1 \right) \delta \lambda + \frac{r}{vJ} \delta T_b$$

$$\delta \dot{\lambda} = -\frac{F_z}{v} \left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J} \right) \mu_1 - \frac{\mu_0}{m} \right) \delta \lambda + \frac{r}{vJ} \delta T_b$$

$$G_{\lambda}(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{F_z}{v} \left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J} \right) \mu_1 - \frac{\mu_0}{m} \right) \right]}$$



Linearization model (method I)

$$J\dot{\omega} = rF_z\mu(\lambda) - T_b \quad -\frac{r}{g}J\dot{\omega} = -\frac{r}{g}(rF_z\mu(\lambda) - T_b) \quad J\eta = -\frac{r^2}{g}F_z\mu(\lambda) + \frac{r}{g}T_b$$

$$\eta = -\frac{r^2}{Jg}F_z\mu(\lambda) + \frac{r}{Jg}T_b$$

$$\delta\eta = -\frac{r^2F_z}{Jg}\mu_1\delta\lambda + \frac{r}{Jg}\delta T_b \quad G_\eta(s) = -\frac{r^2F_z}{Jg}\mu_1G_\lambda(s) + \frac{r}{Jg} = \frac{r}{Jg} \left\{ 1 - \frac{rF_z\mu_1\left[\frac{r}{vJ}\right]}{s + \left[\frac{F_z}{v}\left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J}\right)\mu_1 - \frac{\mu_0}{m}\right)\right]} \right\}$$

$$G_\eta(s) = \frac{r}{Jg} \frac{s + \left[\frac{F_z}{v}\left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J}\right)\mu_1 - \frac{\mu_0}{m}\right)\right] - rF_z\mu_1\left[\frac{r}{vJ}\right]}{s + \left[\frac{F_z}{v}\left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J}\right)\mu_1 - \frac{\mu_0}{m}\right)\right]}$$

$$G_\eta(s) = \frac{r}{Jg} \frac{s + \frac{F_z}{v}\left[\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J}\right)\mu_1 - \frac{\mu_0}{m} - \mu_1\frac{r^2}{J}\right]}{s + \left[\frac{F_z}{v}\left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J}\right)\mu_1 - \frac{\mu_0}{m}\right)\right]}$$

$$G_\eta(s) = \frac{r}{Jg} \frac{s + \frac{F_z}{mv}[(1-\bar{\lambda})\mu_1 - \mu_0]}{s + \left[\frac{F_z}{v}\left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J}\right)\mu_1 - \frac{\mu_0}{m}\right)\right]}$$



Linearization model (method II)

Alternative method of linearization

Do not make assumptions on the dynamic decoupling between the wheel and body.
Linearize around a **point (non-equilibrium)** defined by

$$\bar{v} \quad \bar{\omega} \quad \bar{\lambda} = (\bar{v} - \bar{\omega}r) / \bar{v}$$

$$\mu(\lambda(v, \omega)) = \mu(\bar{\lambda}) + \left. \frac{\partial \mu}{\partial v} \right|_{\lambda=\bar{\lambda}} (v - \bar{v}) + \left. \frac{\partial \mu}{\partial \omega} \right|_{\lambda=\bar{\lambda}} (\omega - \bar{\omega}) + O((v - \bar{v})^2, (\omega - \bar{\omega})^2)$$

,

$$\mu(\lambda(v, \omega)) \approx \mu(\bar{\lambda}) + \left[\frac{\partial \mu}{\partial \lambda} \frac{\partial \lambda}{\partial v} \right]_{\lambda=\bar{\lambda}} (v - \bar{v}) + \left[\frac{\partial \mu}{\partial \lambda} \frac{\partial \lambda}{\partial \omega} \right]_{\lambda=\bar{\lambda}} (\omega - \bar{\omega})$$

$$\mu(\lambda) \approx \mu_0 + \mu_1 \frac{\bar{\omega}r}{\bar{v}^2} \delta v - \mu_1 \frac{r}{\bar{v}} \delta \omega$$



Linearization model (method II)

$$\begin{cases} J\dot{\omega} = rF_z \left[\mu_0 + \mu_1 \frac{\bar{\omega}r}{\bar{v}^2} \delta v - \mu_1 \frac{r}{\bar{v}} \delta \omega \right] - T_b \\ m\dot{v} = -F_z \left[\mu_0 + \mu_1 \frac{\bar{\omega}r}{\bar{v}^2} \delta v - \mu_1 \frac{r}{\bar{v}} \delta \omega \right] \end{cases}$$

$$, \quad \begin{cases} J\bar{\dot{\omega}} = rF_z [\mu_0] - \bar{T}_b \\ m\bar{\dot{v}} = -F_z [\mu_0] \end{cases} \quad \text{equilibrium}$$

$$\begin{cases} \delta\dot{\omega} = \left[-\mu_1 \frac{r^2 F_z}{\bar{v}J} \right] \delta\omega + \left[\mu_1 \frac{\bar{\omega}r^2}{J\bar{v}^2} F_z \right] \delta v - \frac{1}{J} \delta T_b \\ \delta\dot{v} = \left[\mu_1 \frac{r}{m\bar{v}} F_z \right] \delta\omega + \left[-\mu_1 \frac{\bar{\omega}r}{m\bar{v}^2} F_z \right] \delta v \end{cases}$$

Is the final relation identical to the previous one ...?



Linearization model (method II)

$$\begin{cases} \delta\dot{\omega} = \begin{bmatrix} -\mu_1 \frac{r^2 F_z}{\bar{v} J} \end{bmatrix} \delta\omega + \begin{bmatrix} \mu_1 \frac{\bar{\omega} r^2}{J \bar{v}^2} F_z \end{bmatrix} \delta v - \frac{1}{J} \delta T_b \\ \delta\dot{v} = \begin{bmatrix} \mu_1 \frac{r}{m \bar{v}} F_z \end{bmatrix} \delta\omega + \begin{bmatrix} -\mu_1 \frac{\bar{\omega} r}{m \bar{v}^2} F_z \end{bmatrix} \delta v \end{cases}$$

```
>> A=sym(' [-m1*r^2*Fz/(v*J),m1*w*r^2*Fz/(J*v^2);m1*r*Fz/(v*m),-m1*w*r*Fz/(m*v^2)] ');
>> B=sym(' [-1/J;0] ');
>> C=sym(' [1,0;0,1] ');
>> I=sym(eye(2));
>> F=C*inv(sym('s')*I-A)*B;
>> pretty(F)
```

$$\begin{bmatrix} \frac{s^2 m v^2 + m_1 w r^2 F_z}{s^2 (s v^2 J m + J m_1 w r F_z + m_1 r^2 F_z m v)} & \frac{m_1 r F_z v}{s^2 (s v^2 J m + J m_1 w r F_z + m_1 r^2 F_z m v)} \end{bmatrix}$$

$$G_{\omega}(s) = -\frac{1}{Js} \frac{s + \frac{\mu_1(\bar{\lambda}) F_z}{m \bar{v}} (1 - \bar{\lambda})}{s + \left[\frac{\mu_1(\bar{\lambda}) F_z}{m \bar{v}} \left((1 - \bar{\lambda}) + \frac{m r^2}{J} \right) \right]}$$

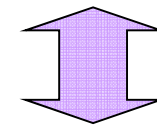
$$G_v(s) = -\frac{1}{Js} \frac{\frac{\mu_1(\bar{\lambda}) r F_z}{m \bar{v}}}{s + \left[\frac{\mu_1(\bar{\lambda}) F_z}{m \bar{v}} \left((1 - \bar{\lambda}) + \frac{m r^2}{J} \right) \right]}$$



Linearization model (method II)

$$\delta\lambda = -\frac{r}{\bar{v}}\delta\omega + \frac{\bar{\omega}r}{\bar{v}^2}\delta v$$

$$G_{\lambda}(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$



$$G_{\lambda}(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{F_z}{v} \left(\left(\frac{(1 - \bar{\lambda})}{m} + \frac{r^2}{J} \right) \mu_1 - \frac{\mu_0}{m} \right) \right]}$$



Linearization model

Method I

$$G_{\lambda}(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{F_z}{v} \left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J} \right) \mu_1 - \frac{\mu_0}{m} \right) \right]}$$

$$G_{\eta}(s) = \frac{r}{Jg} \frac{s + \frac{F_z}{m\bar{v}} \left[(1-\bar{\lambda})\mu_1 + \mu_0 \right]}{s + \left[\frac{F_z}{v} \left(\left(\frac{(1-\bar{\lambda})}{m} + \frac{r^2}{J} \right) \mu_1 - \frac{\mu_0}{m} \right) \right]}$$

Linearization around an equilibrium point

Assumption of v-slowly-varying

Method II

$$G_{\lambda}(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1-\bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$

$$G_{\eta}(s) = \frac{r}{Jg} \frac{s + \frac{F_z}{m\bar{v}} (1-\bar{\lambda})\mu_1(\bar{\lambda})}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1-\bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$

Linearization around a non-equilibrium point

No v-slowly-varying assumption



Analysis: open-loop

$$G_{\lambda}(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$

$$G_{\eta}(s) = \frac{r}{Jg} \frac{s + \frac{F_z}{m\bar{v}}(1 - \bar{\lambda})\mu_1(\bar{\lambda})}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$

Pole (stability):

$$\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) > 0 \quad \mu_1(\bar{\lambda}) > 0$$

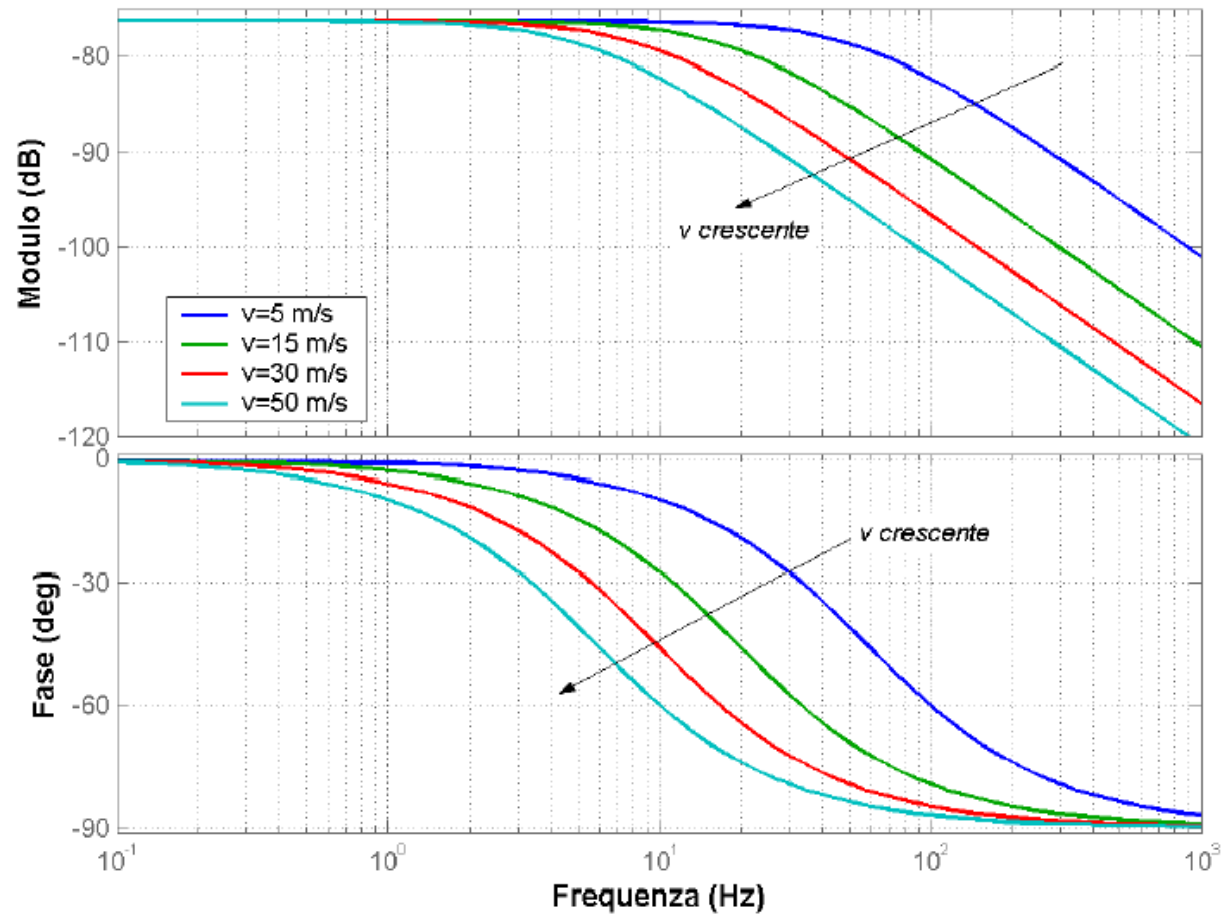
Zero (minimum phase):

$$\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} (1 - \bar{\lambda}) > 0 \quad \mu_1(\bar{\lambda}) > 0$$



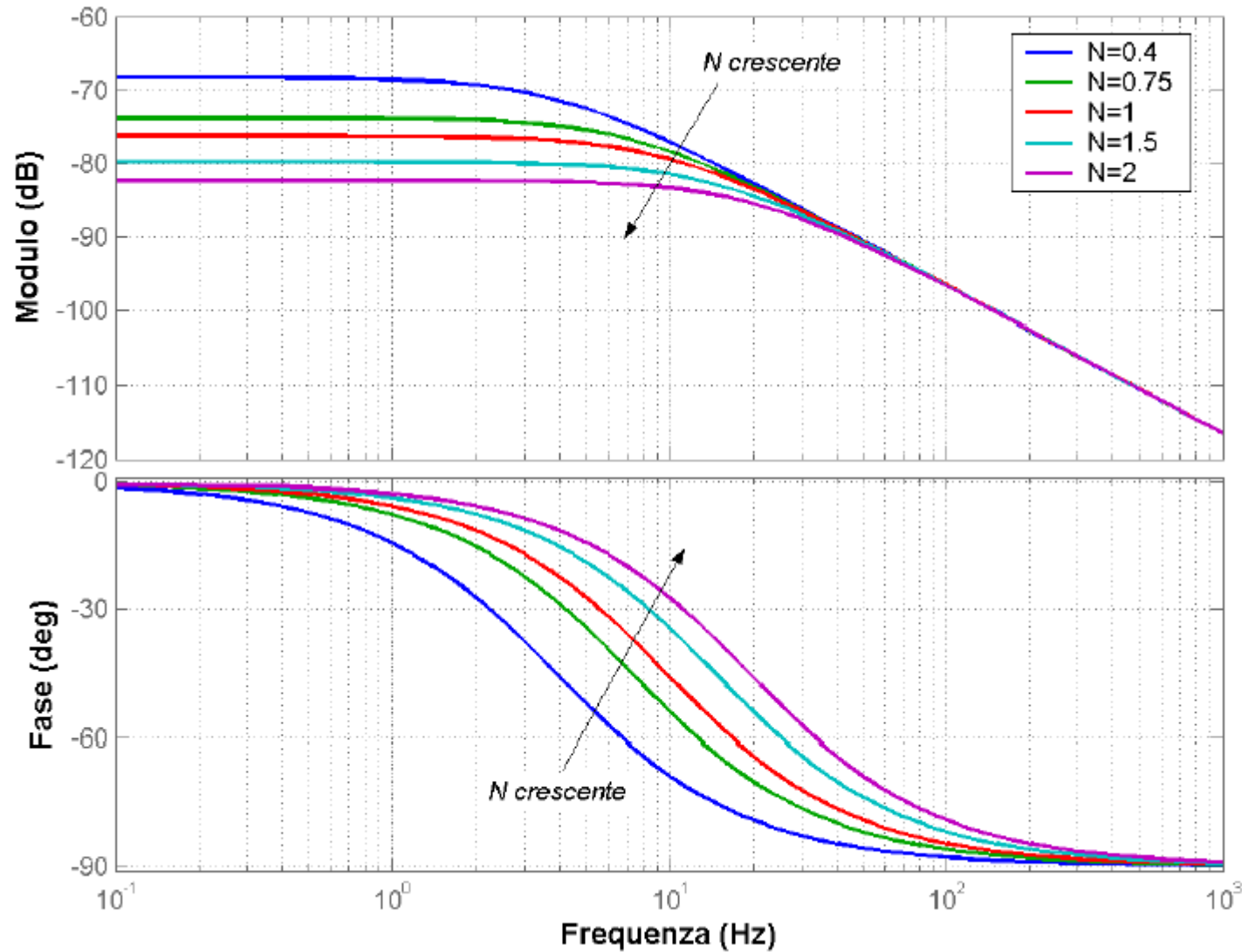
Sensitivity with respect to the longitudinal vehicle speed

F.d.t. From T_b to λ (nominal $F_z = Mg$)





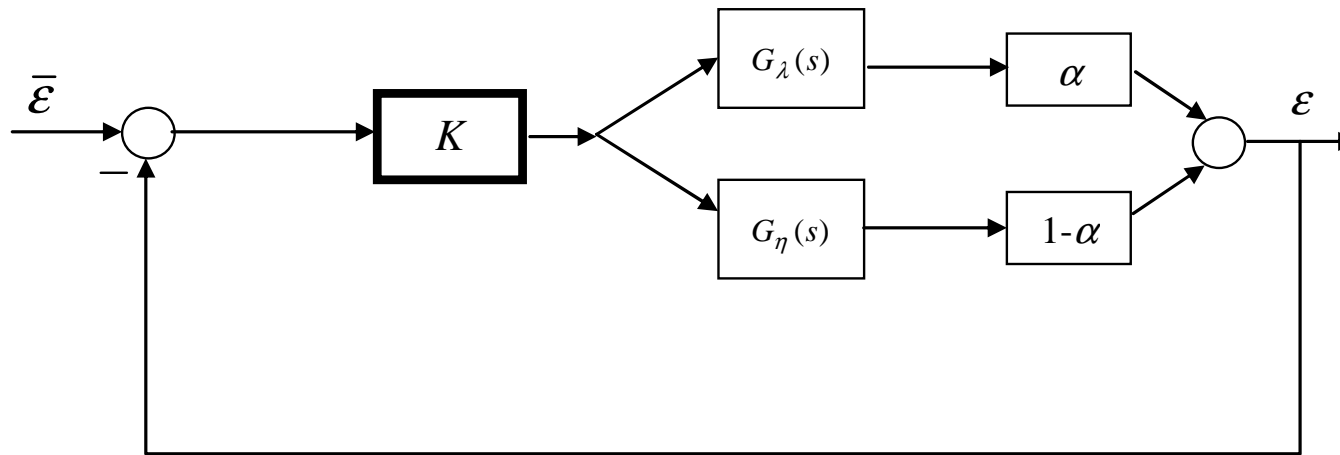
Sensitivity with respect to vertical load



$$N = \frac{F_z}{mg}$$



Control schemes



$$G_\lambda(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$

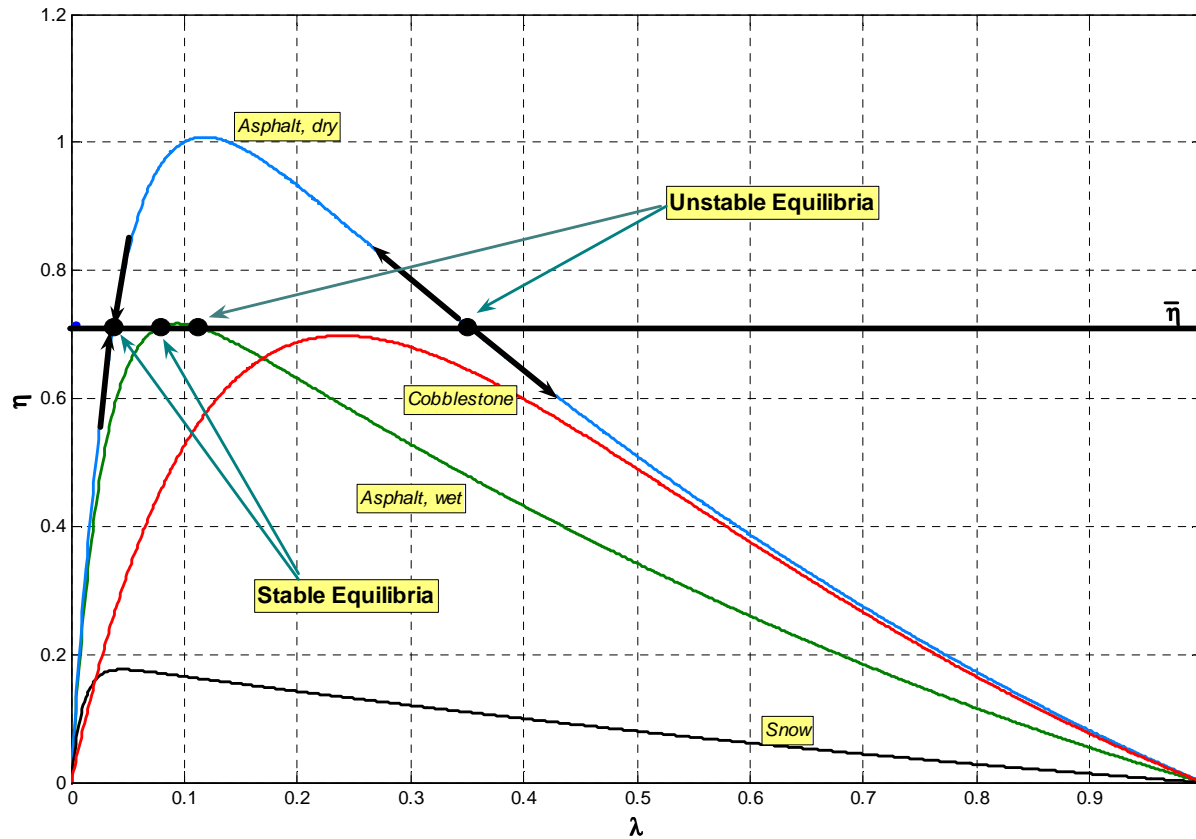
$\alpha = 1$ (Slip control)

$\alpha = 0$ (Deceleration control)

$$G_\eta(s) = \frac{\frac{r}{Jg} \left(s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} ((1 - \bar{\lambda})) \right] \right)}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$



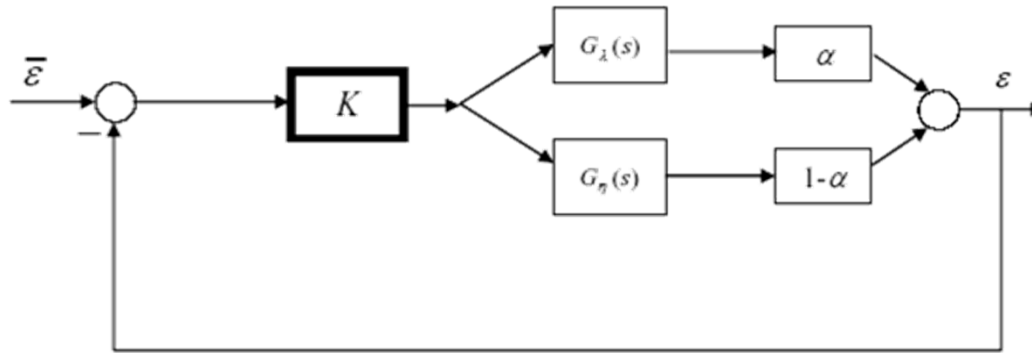
$\alpha = 0$ (deceleration control)



- No fixed set-point
- No unique equilibrium
- deceleration measurement very simple and robust



$\alpha = 0$ (deceleration control)



$$G_{\eta}(s) = \frac{\frac{r}{Jg} \left(s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} ((1-\bar{\lambda})) \right] \right)}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1-\bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$

$$\chi_{\eta}(s) = \left(1 + K \frac{r}{Jg} \right) s + \frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1-\bar{\lambda}) \left(1 + K \frac{r}{Jg} \right) + \frac{mr^2}{J} \right)$$

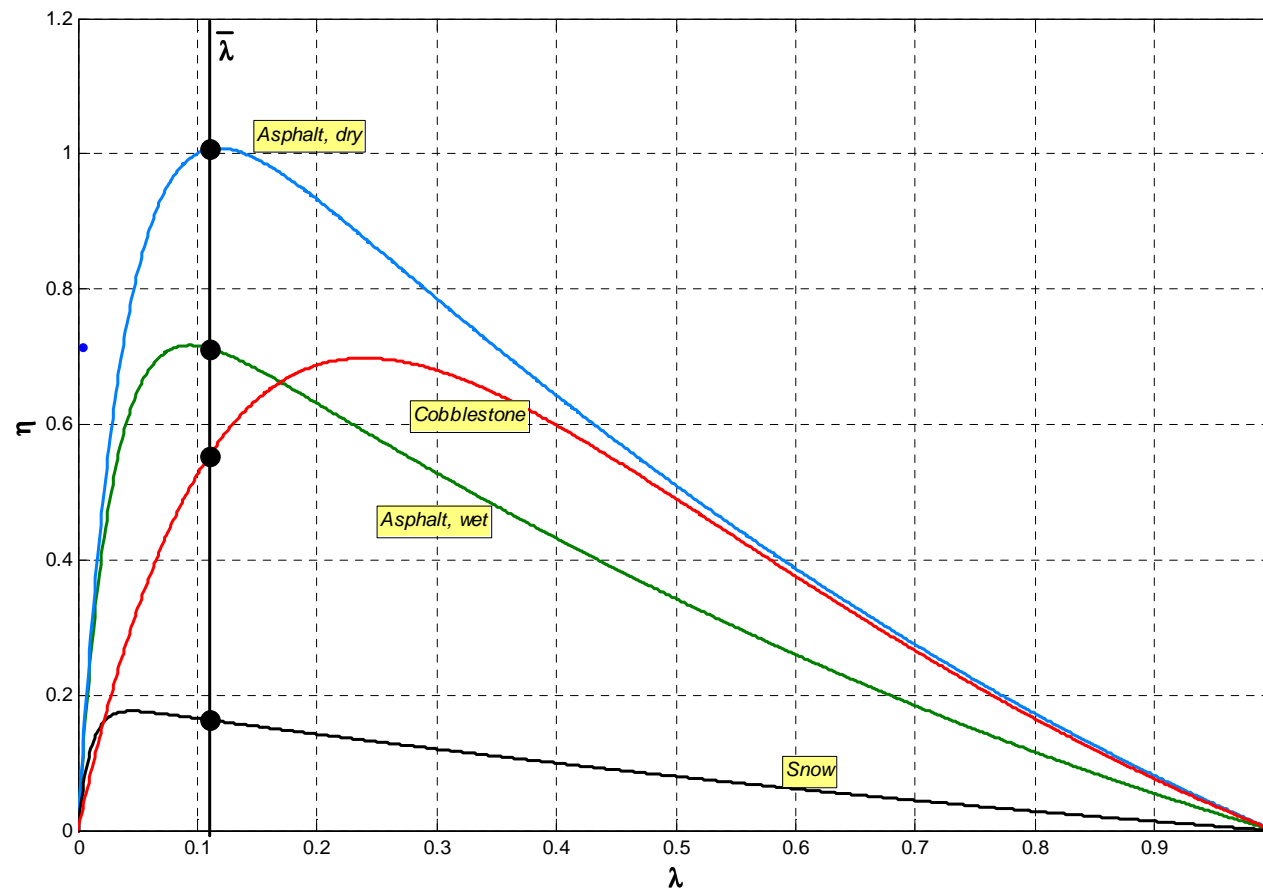
\Downarrow

$$\mu_1(\bar{\lambda}) \left((1-\bar{\lambda}) \left(1 + K \frac{r}{Jg} \right) + \frac{mr^2}{J} \right) > 0$$

Impossible to find a stabilizing **K** in all working conditions



$\alpha = 1$ (slip control)



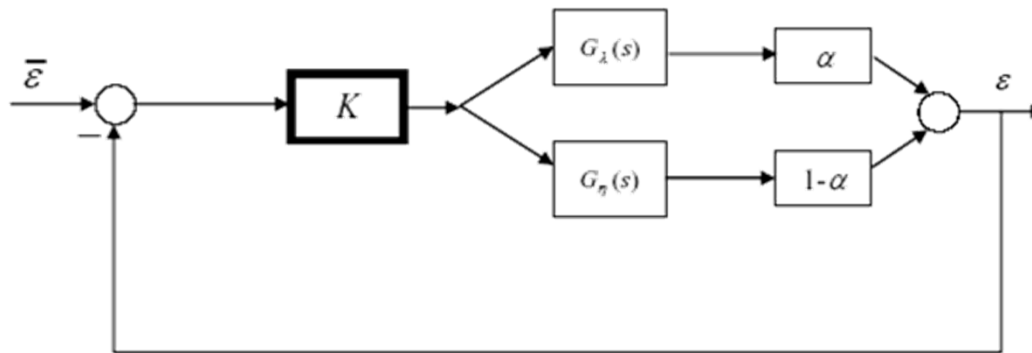
- Set-point (almost) fixed

- Unique equilibrium

- Sensitivity to measurement noise (slip measurement is critical)



$\alpha = 1$ (slip control)



$$G_\lambda(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) \right]}$$

$$\chi_\lambda(s) = s + \frac{1}{\bar{v}} \left[\frac{\mu_1(\bar{\lambda})F_z}{m} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) + K \frac{r}{J} \right]$$

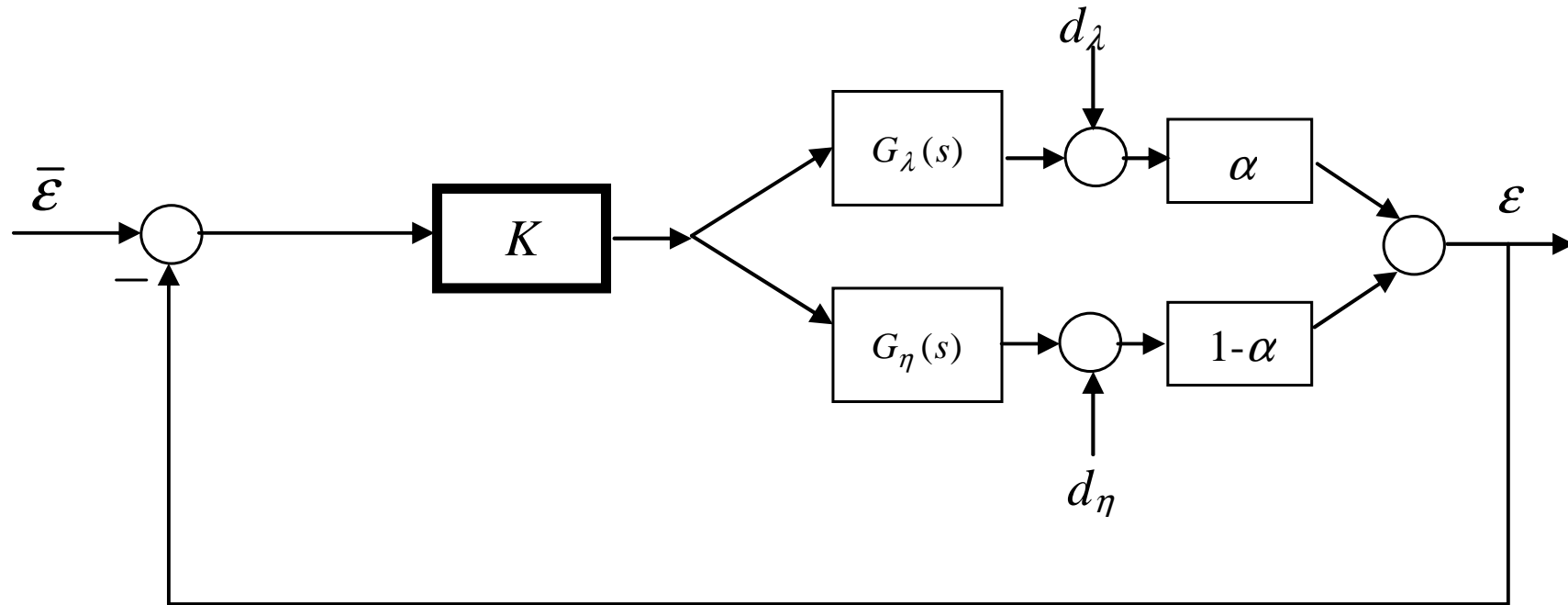
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$$K \frac{r}{J} > - \frac{\mu_1(\bar{\lambda})F_z}{m} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right)$$

Possible to find a stabilizing K in all working conditions



Mixed Slip-Deceleration control (MSD-control)



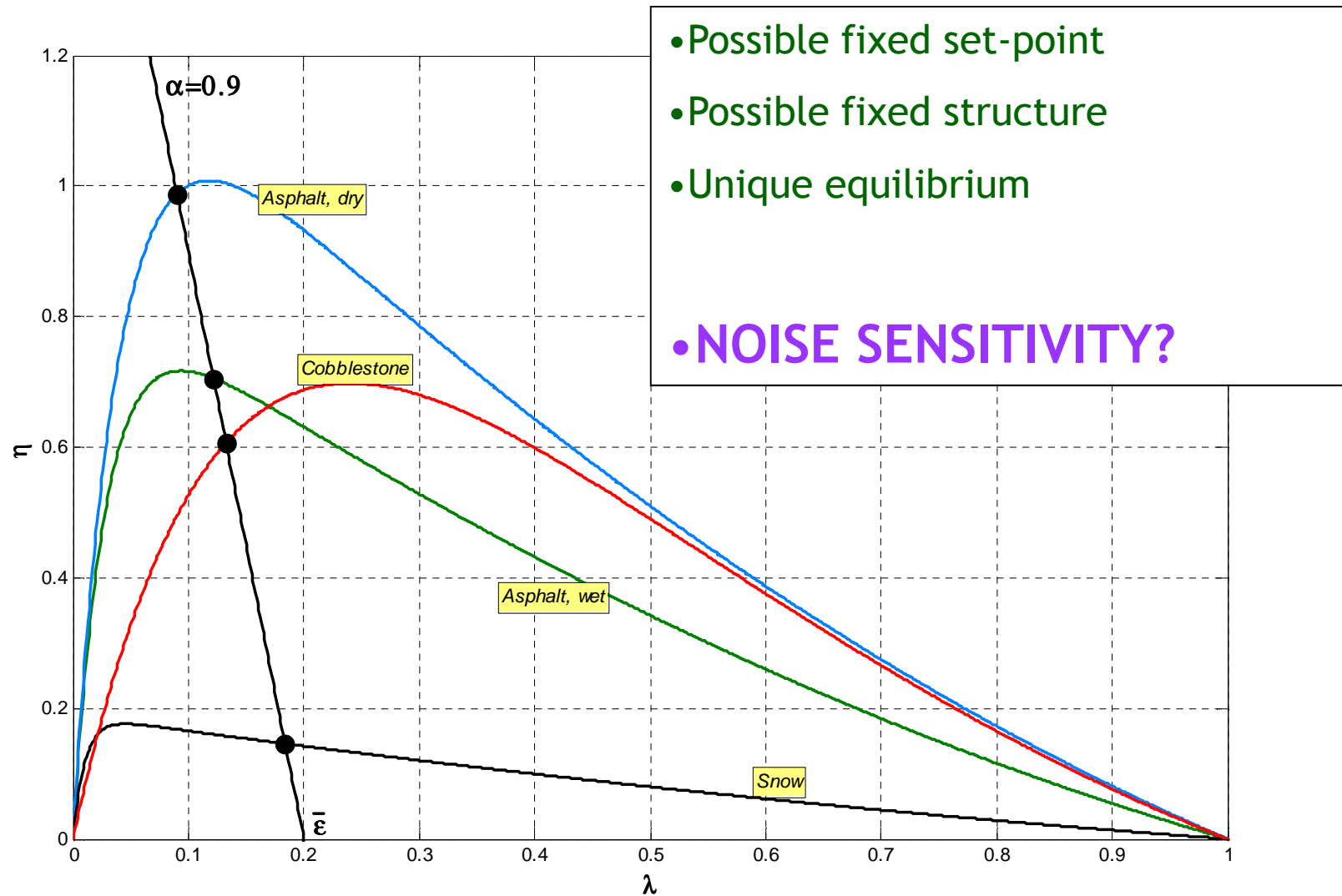
$$\epsilon = \alpha\lambda + (1 - \alpha)\eta, \quad \alpha \in [0, 1]$$

Savaresi s.m., M. Tanelli, C. Cantoni (2007). Mixed slip-deceleration control in automotive braking systems. ASME Transactions: Journal of Dynamic Systems, Measurement and Control, Vol.129, No. 1, pp.20-31 [2008 ASME Dynamic Systems and Control Rudolf Kalman Best Paper Award]

PCT/EP2005/050820 (2005). "Sistema di controllo automatico della frenatura", Freni Brembo S.p.A. (Cantoni C., Savaresi S., Charalambakis D., Tanelli M.).



$0 < \alpha < 1$ (mixed slip-deceleration control - MSD-control)

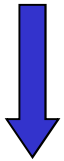




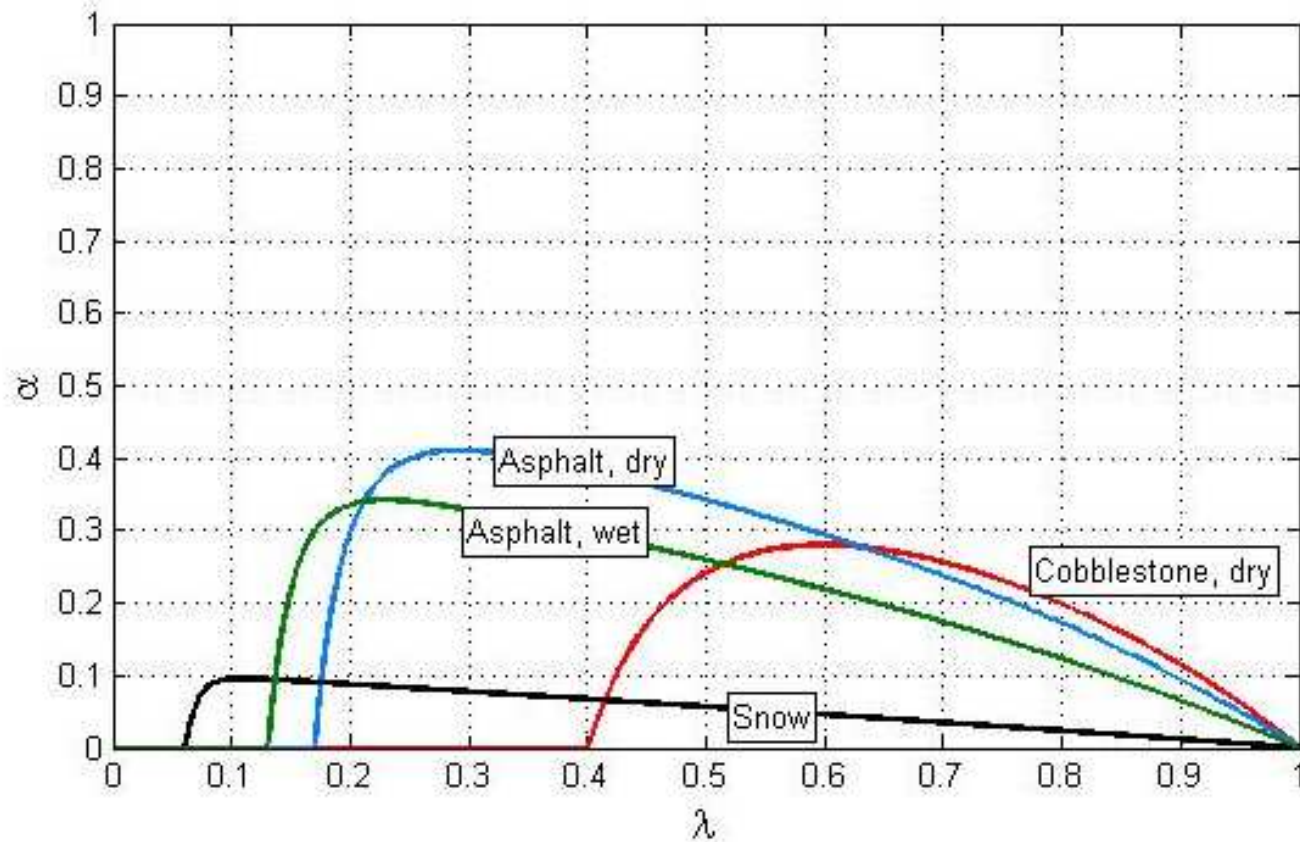
Limits in the choice of α

$$\alpha_{min} = \max_{\bar{\lambda}, \theta_r} \frac{\mu_1(\bar{\lambda}; \theta_r)(\bar{\lambda} - 1) \frac{F_z}{mg}}{1 + \left[\mu_1(\bar{\lambda}; \theta_r)(\bar{\lambda} - 1) \right] \frac{F_z}{mg}}$$

$$F_z = m g$$

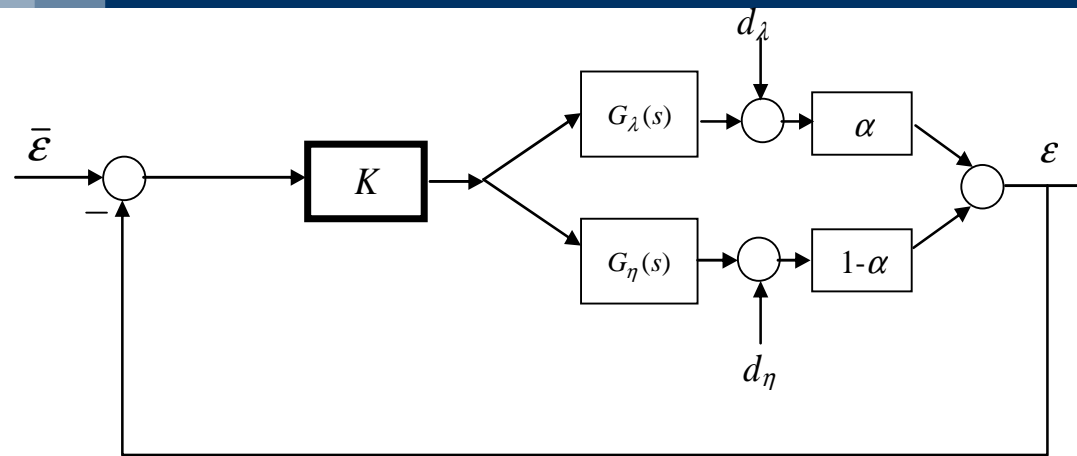


$$\alpha_{min} \approx 0.45$$





Noise sensitivity



$$\sigma_\lambda^2 > \sigma_\eta^2$$

$$D_\varepsilon(s; \alpha) = (\alpha D_\lambda(s) + (1 - \alpha) D_\eta(s)) S_\varepsilon(s; \alpha)$$

$$\gamma(\alpha) = \frac{\text{var}[d_\varepsilon(\alpha)]}{\text{var}[d_\varepsilon(1)]}$$



Noise sensitivity

$$\gamma(\alpha) = \frac{\text{var}[d_{\varepsilon}(\alpha)]}{\text{var}[d_{\varepsilon}(1)]} = \Psi(\alpha)\Phi(\alpha)$$

$$\Psi(\alpha) = \frac{\alpha^2 \sigma_{\lambda}^2 + (1-\alpha)^2 \sigma_{\eta}^2}{\sigma_{\lambda}^2}$$

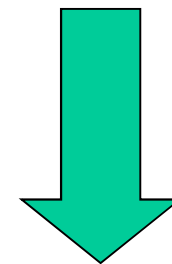
$$0.80 \leq \alpha \leq 0.95$$

$$\sigma_{\lambda}^2 > \sigma_{\eta}^2$$



$$\Psi(\alpha) \approx \alpha^2$$

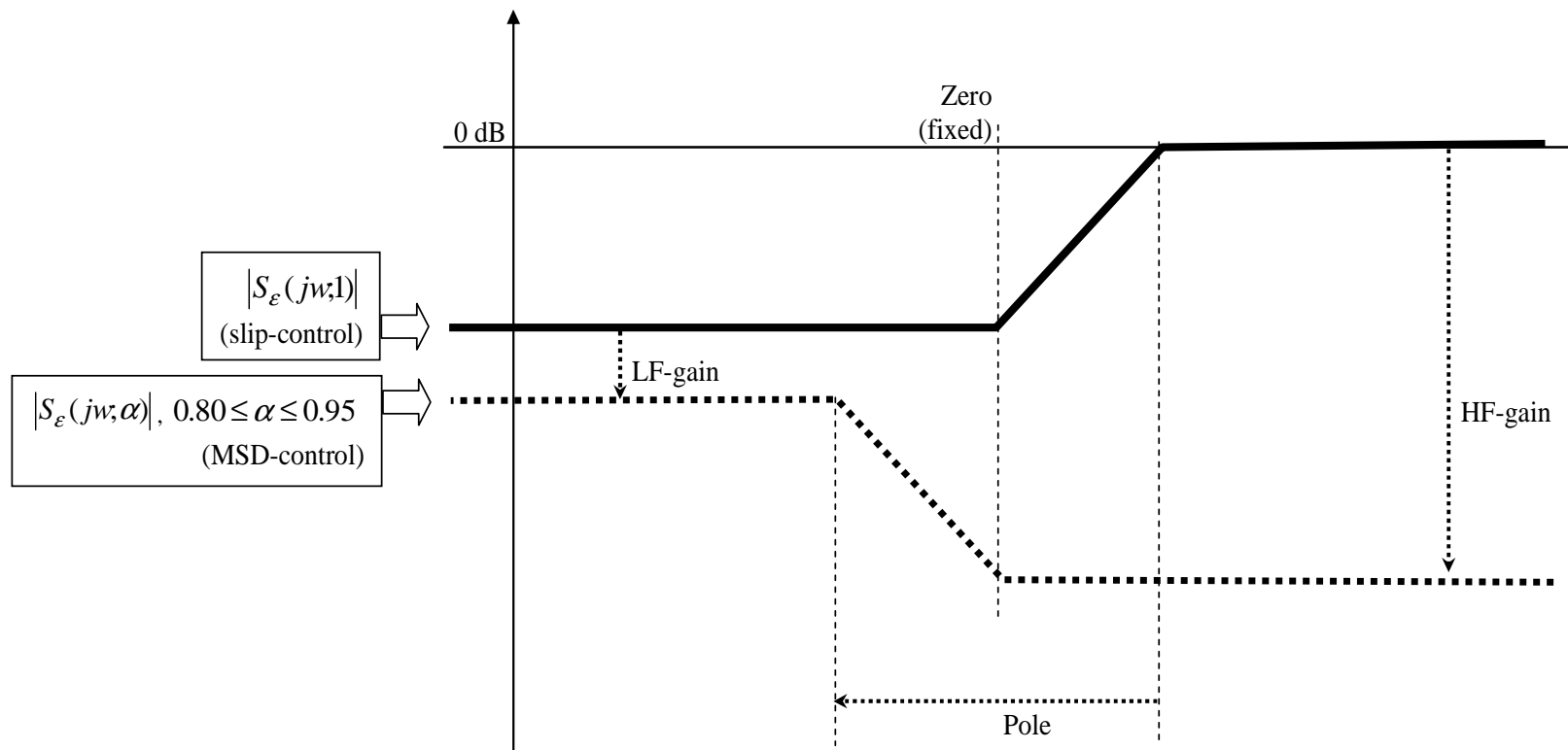
$$\Phi(\alpha) = \frac{\int_{w=0}^{\Omega_N} |S_{\varepsilon}(jw, \alpha)|^2 dw}{\int_{w=0}^{\Omega_N} |S_{\varepsilon}(jw, 1)|^2 dw}$$





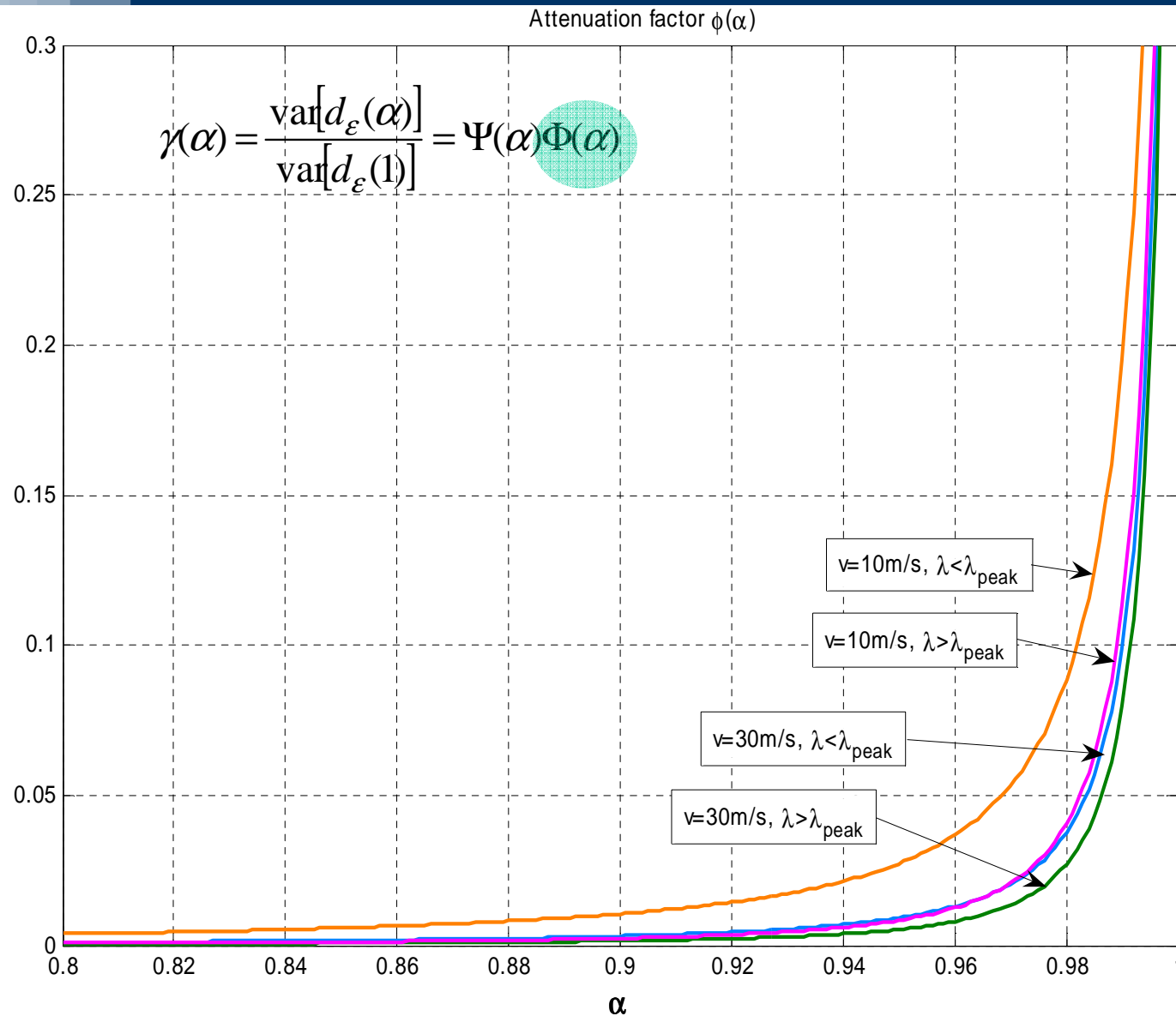
Noise sensitivity

$$S_{\varepsilon}(s; \alpha) = \frac{s + \frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right)}{s \left[1 + K \frac{r}{Jg} (1 - \alpha) \right] + \frac{1}{\bar{v}} \left[\frac{\mu_1(\bar{\lambda})F_z}{m} \left((1 - \bar{\lambda}) \left(1 + K \frac{r}{Jg} (1 - \alpha) \right) + \frac{mr^2}{J} \right) + K\alpha \frac{r}{J} \right]}$$



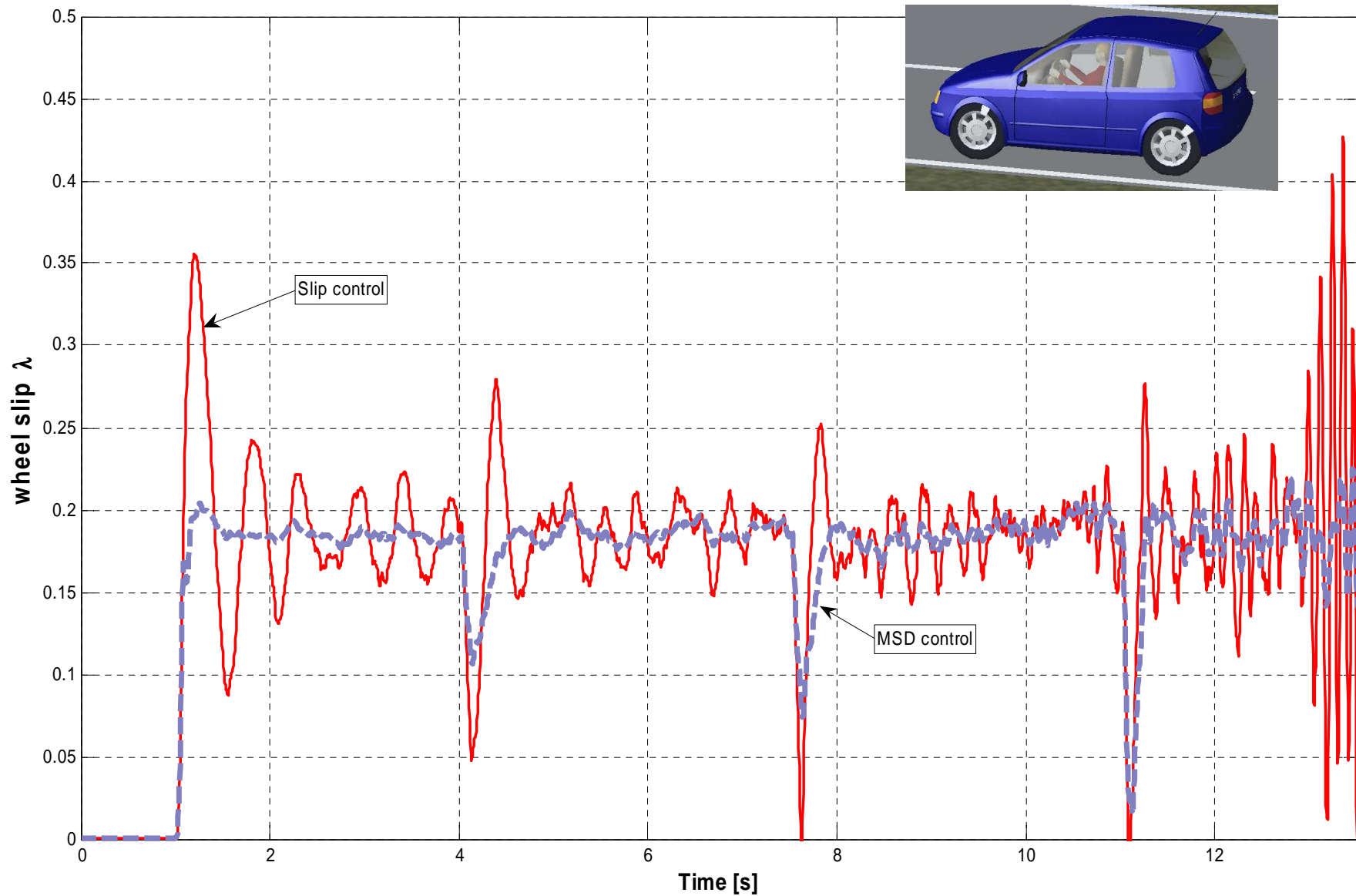


Noise sensitivity





Slip Vs. MSD ($\alpha = 0.9$)





Slip-control: linear stability analysis

$$[m=225, J=1, r=0.28]$$

$$G_{\lambda}(s) = \frac{\left[\frac{r}{vJ} \right]}{s + \left[\frac{\mu_1(\bar{\lambda})F_z}{m\bar{v}} \left((1 - \bar{\lambda}) + \frac{mr^2}{J} \right) \right]} \rightarrow$$

$$G_{\lambda}(s) = \frac{\frac{0.28}{\bar{v}}}{s + \frac{1}{\bar{v}} \left[\mu_1(\bar{\lambda})9.8((1 - \bar{\lambda}) + 17.6) \right]}$$

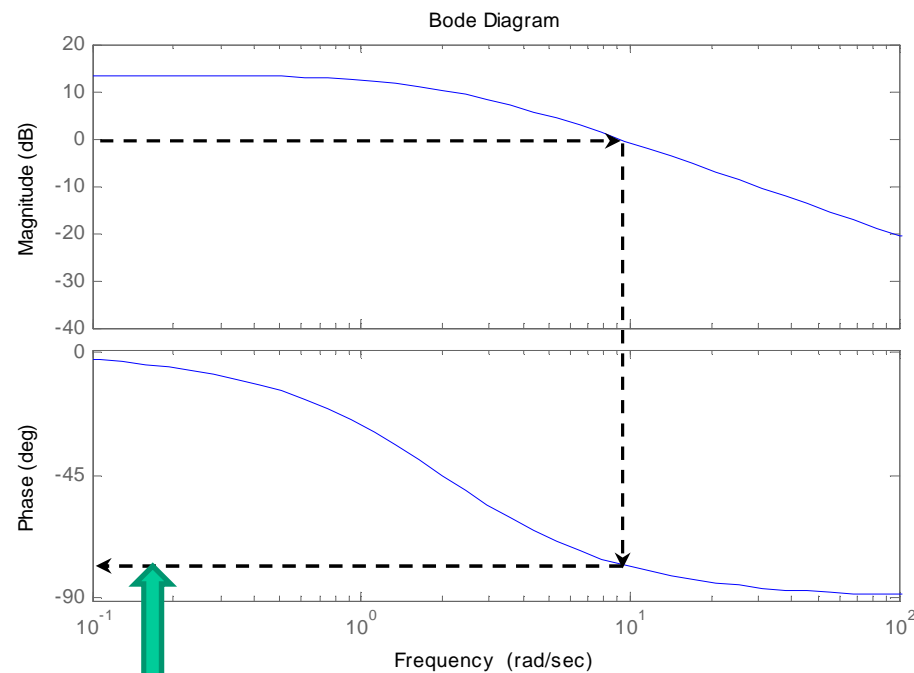
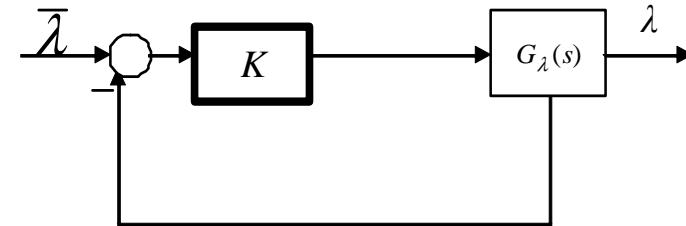


Slip-control: linear stability analysis

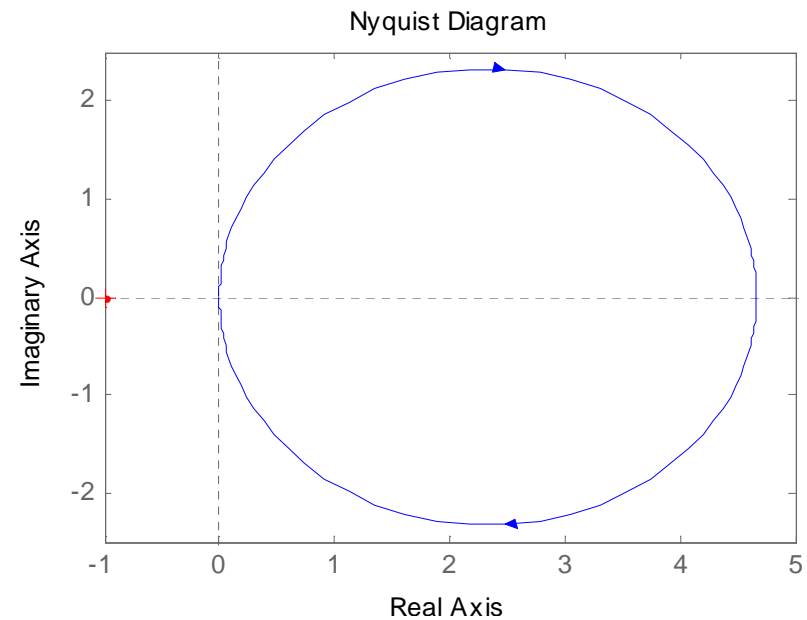
$$1000 \left[\begin{array}{c} \frac{0.28}{30} \\ s + \frac{60}{30} \end{array} \right]$$

No unstable open-loop pole
No (-1,0) (counter-clock-wise)
encircling

Asymptotic stability
Bode criterion is applicable

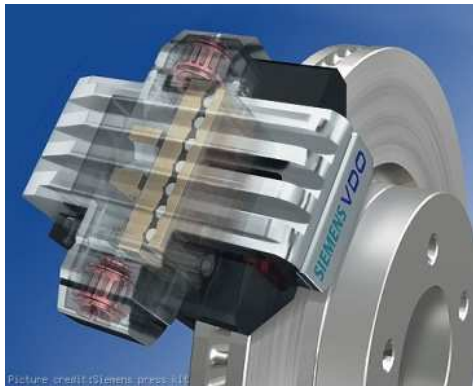


Phase margin = distance to “-180°”



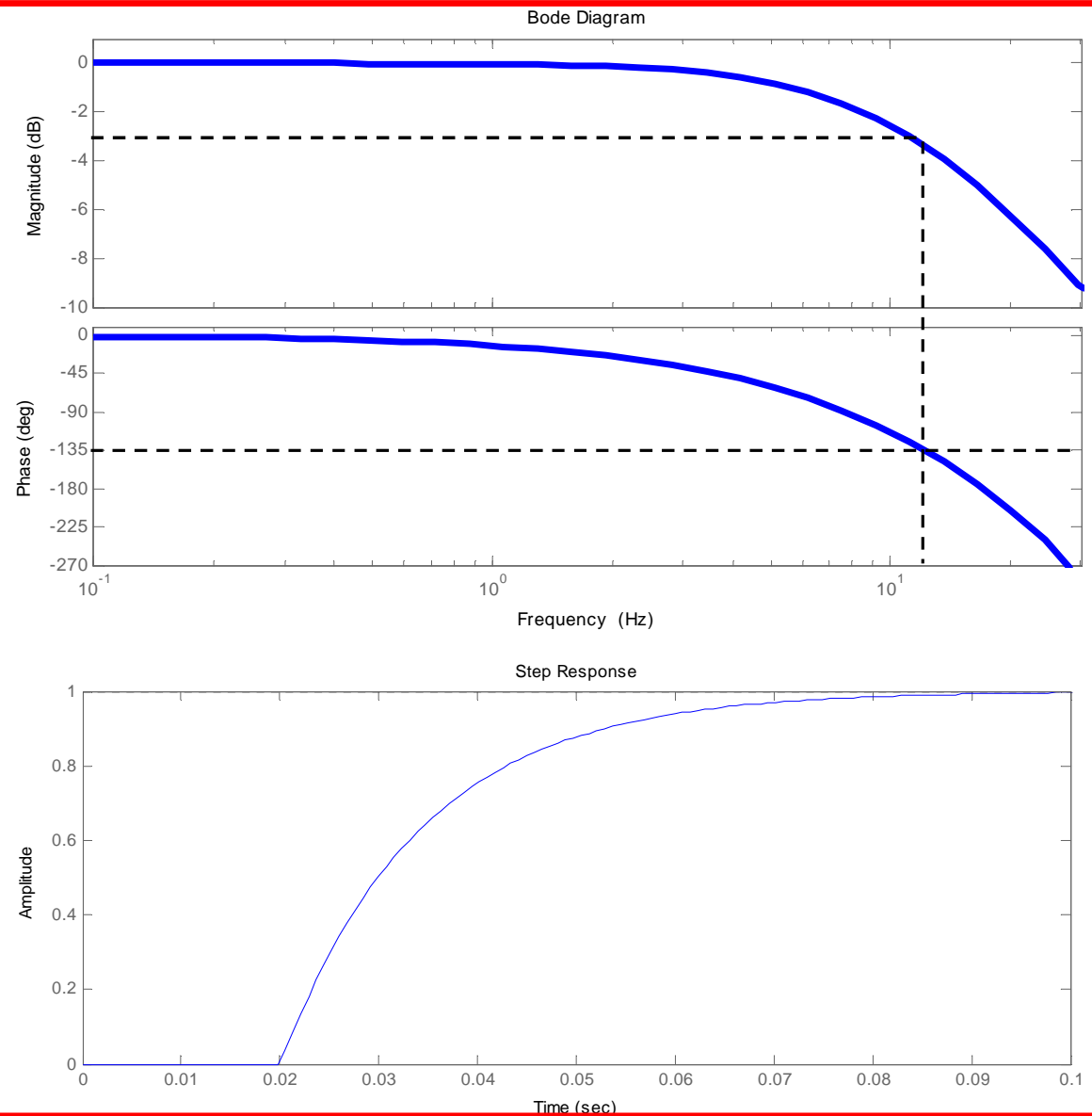


Actuator dynamics



Added dynamics (example)

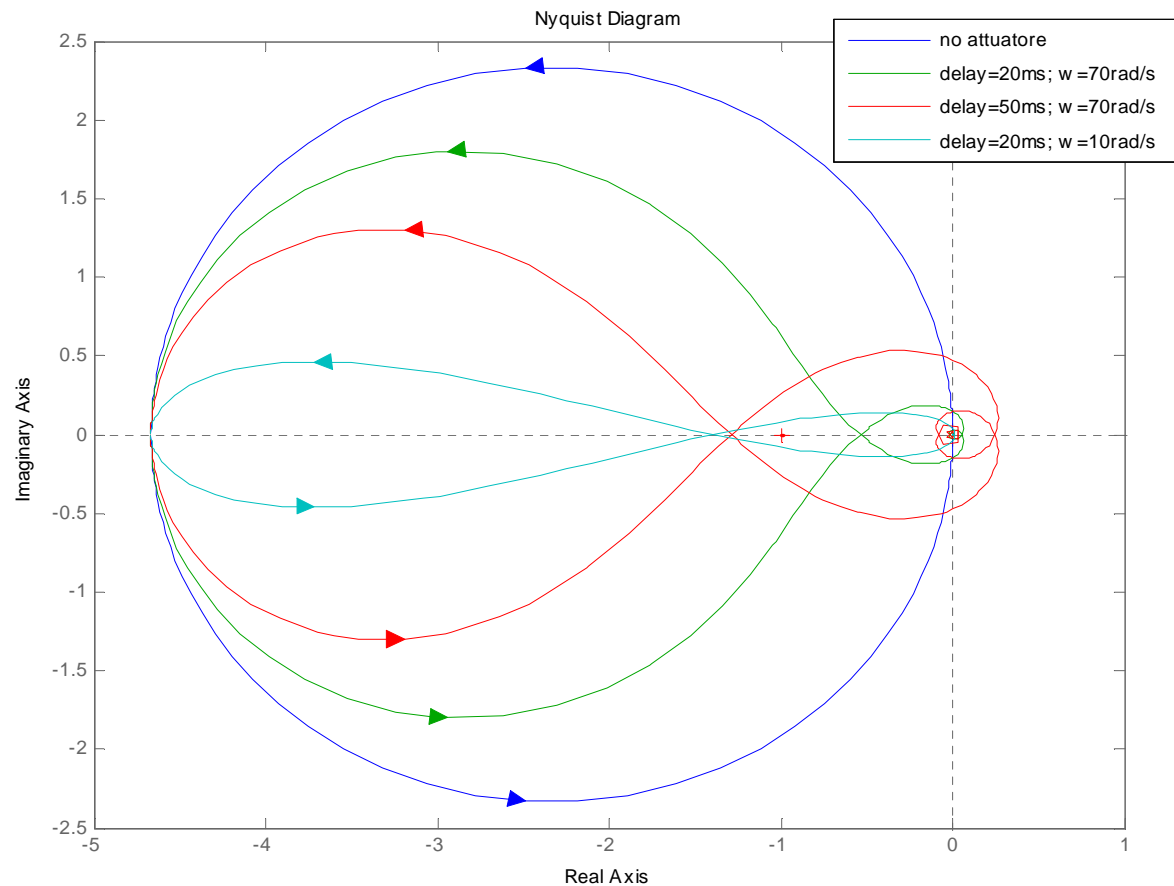
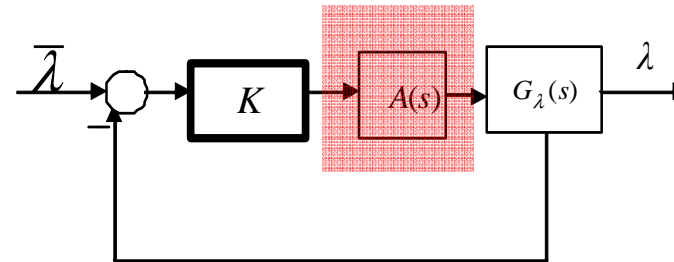
$$F_{caliper}(s) = \frac{70}{s + 70} e^{-(0.02)s}$$





Example - effect of the actuator dynamics

$$1000 \left[\begin{array}{c} \frac{0.28}{10} \\ s - \frac{60}{10} \end{array} \right] \left[\begin{array}{c} 70 \\ s + 70 \end{array} \right] \left[e^{-(0.02)s} \right]$$





Slip-control: example of a PID design

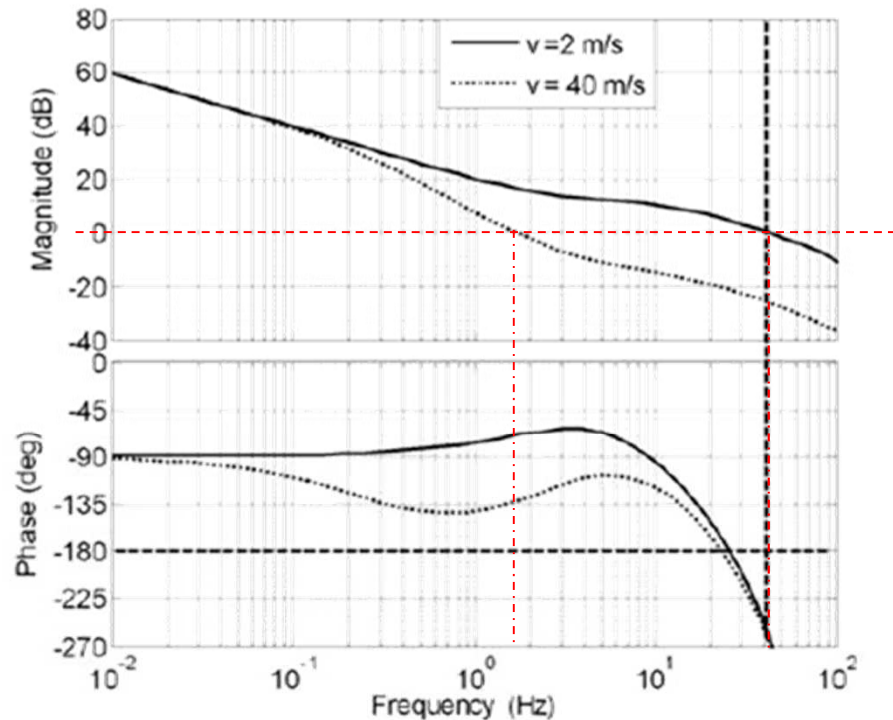


Figure 3.8 Magnitude and phase Bode diagrams of the frequency response associated with the loop transfer function $L_{2\lambda}(s)$ for $v = 40$ m/s (dotted line) and $v = 2$ m/s (solid line)

Cut-off frequency: between 2 and 40Hz about

At 2m/s becomes unstable

with $\omega_{act} = 70$ rad/s and $\tau = 10$ ms.

$$L_{2\lambda}(s) = R_{\lambda}(s)D^{st}(s),$$

$$D^{st}(s) = G_{caliper}(s)G_{\lambda}^{st}(s)$$

$$G_{\lambda}^{st}(s) = \frac{0.3/\bar{v}}{s + 60/\bar{v}},$$

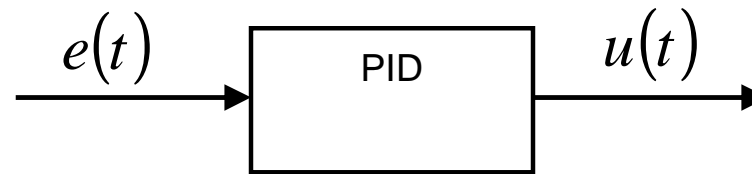
$$R_{\lambda}(s) = 12000 \frac{(1 + \frac{1}{20}s)^2}{s(1 + \frac{1}{500}s)}.$$

$$G_{caliper}(s) = \frac{\omega_{act}}{s + \omega_{act}} e^{-s\tau}, \quad (1.4)$$



Recall: “Ideal” PID

Controllers **PID** → action **Proportional**, **Integral**, **Derivative**.



$$u(t) = K_p e(t) + K_i \int_0^t e(\tau) d\tau + K_d \frac{de(t)}{dt}$$

Coeff.
Proportional

Coeff.
Integral
Action

Coeff.
Derivative
Action

$$u(t) = K_p \left[e(t) + \frac{1}{T_i} \int_0^t e(\tau) d\tau + T_d \dot{e}(t) \right]$$

Integral action
time

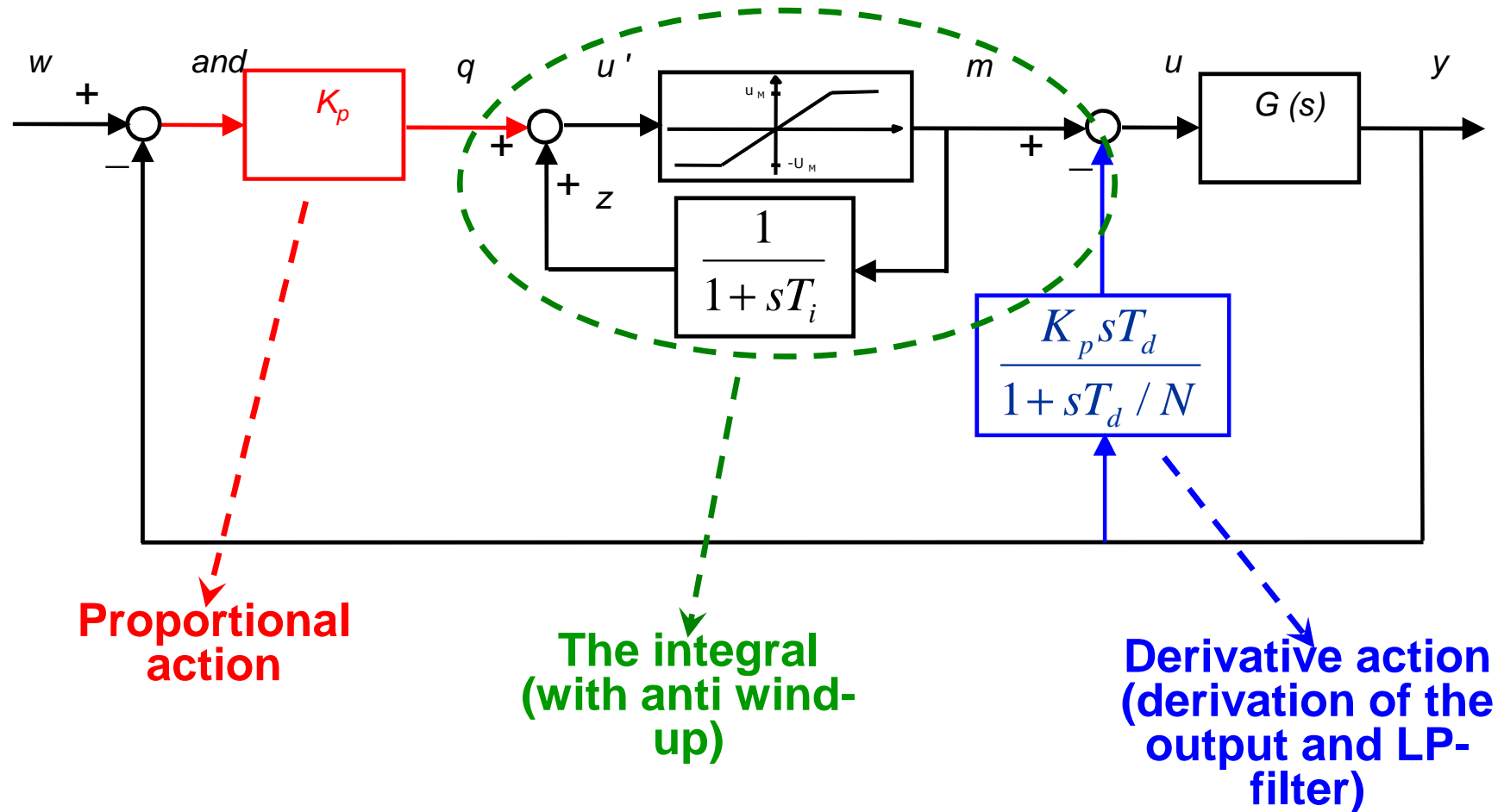
Derivative
action time

$$T_i = \frac{K_p}{K_i}$$

$$T_d = \frac{K_d}{K_p}$$

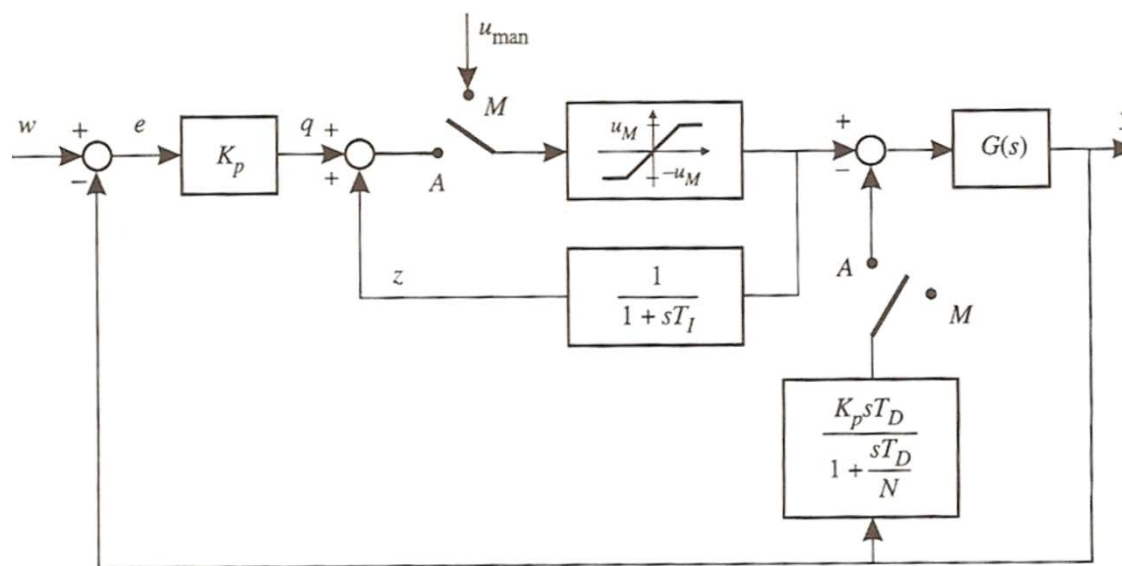
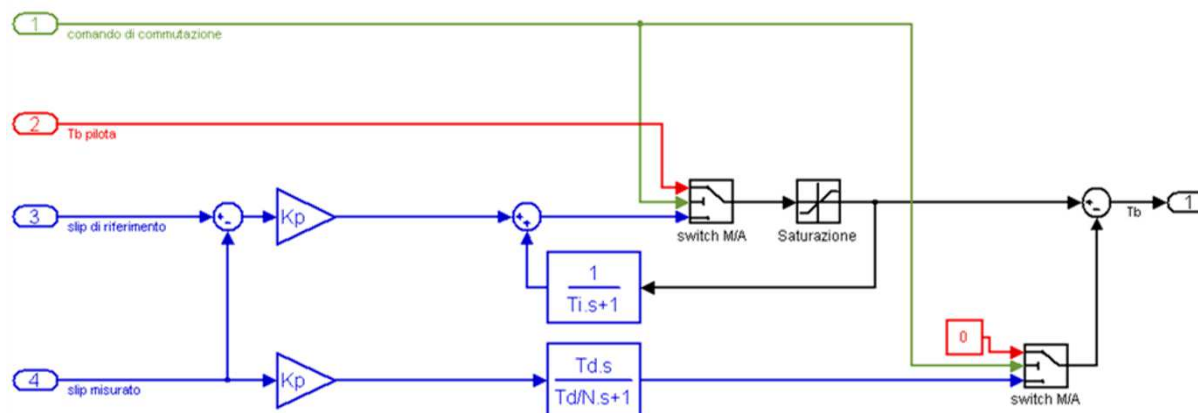


Recall: “real” PID





Logic activation-deactivation on a PID



activation and deactivation variable(s)?

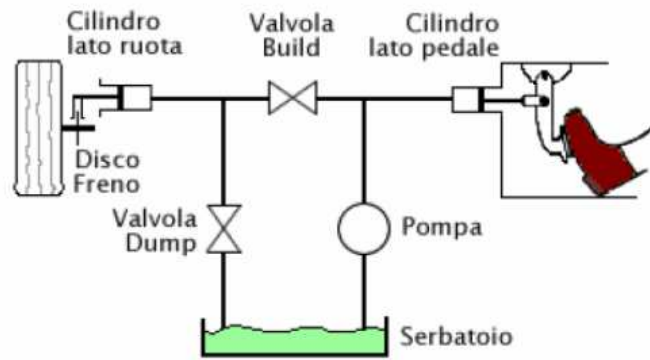


Slip-control algorithms (ABS) with three-states- actuators



Actuator

HAB: Hydraulic Actuated Brakes (increase, hold, decrease)



Constraints:

1. Rate limiter $dT_b/dT=u$
2. Control action: $u \in \{k, 0, -k\}$

Control actions:

Open Build - Close dump \rightarrow Increase pressure ($u = k$)

Close Build - Open dump \rightarrow decrease pressure ($u = -k$)

Build Close - Close dump \rightarrow hold pressure ($u = 0$)



Model (recall)

$$\begin{cases} \dot{\lambda} = -\frac{1}{v} \left(\frac{(1-\lambda)}{m} + \frac{r^2}{J} \right) F_z \mu(\lambda) + \frac{r}{vJ} T_b \\ m\dot{v} = -F_z \mu(\lambda) \end{cases}$$



Hp: v is a parameter ("slowly varying")



$$\dot{\lambda} = -\frac{1-\lambda}{\omega r} \left(\frac{(1-\lambda)}{m} + \frac{r^2}{J} \right) F_z \mu(\lambda) + \frac{r}{J} \frac{1-\lambda}{\omega r} T_b$$



Equilibria (open loop)

$$\dot{\lambda} = -\frac{1-\lambda}{\omega r} \left(\frac{(1-\lambda)}{m} + \frac{r^2}{J} \right) F_z \mu(\lambda) + \frac{r}{J} \frac{1-\lambda}{\omega r} T_b$$

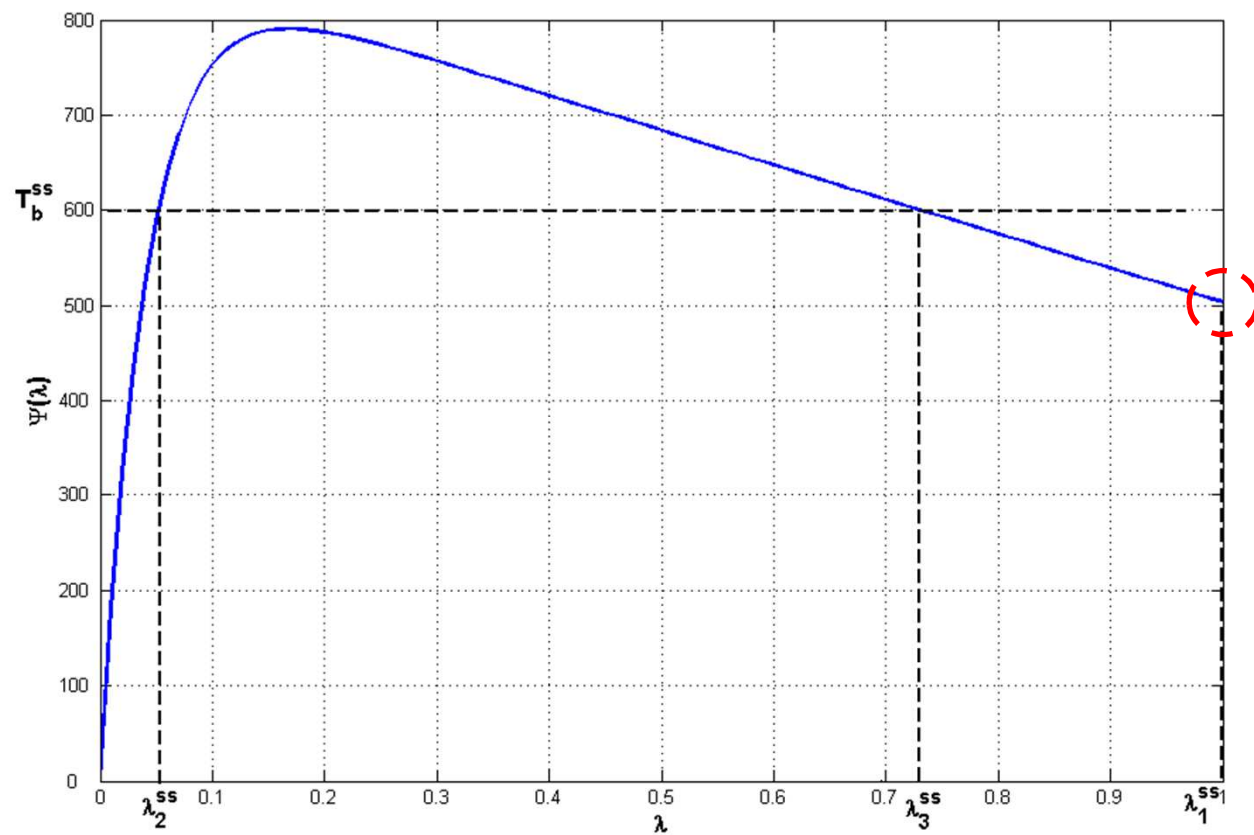
$$\lambda = \lambda^{ss}$$

$$T_b = T_b^{ss} = \psi(\lambda) = F_z \mu(\lambda) \left(r + \frac{J}{rm} (1-\lambda) \right)$$



Equilibria (open loop)

$$T_b^{ss} > \max_{\lambda} \Psi(\lambda) \rightarrow \lambda_1^{ss} = 1$$

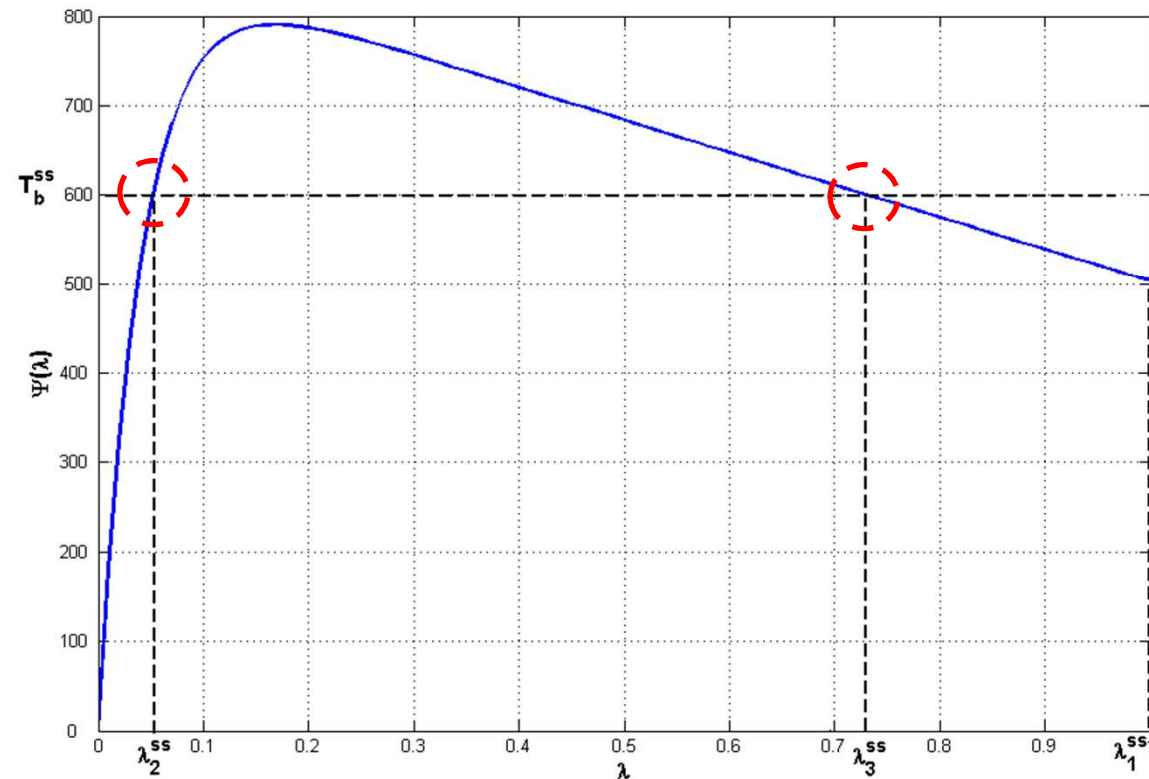




Equilibria (open loop)

$$T_b^{ss} \leq \max_{\lambda} \psi(\lambda)$$

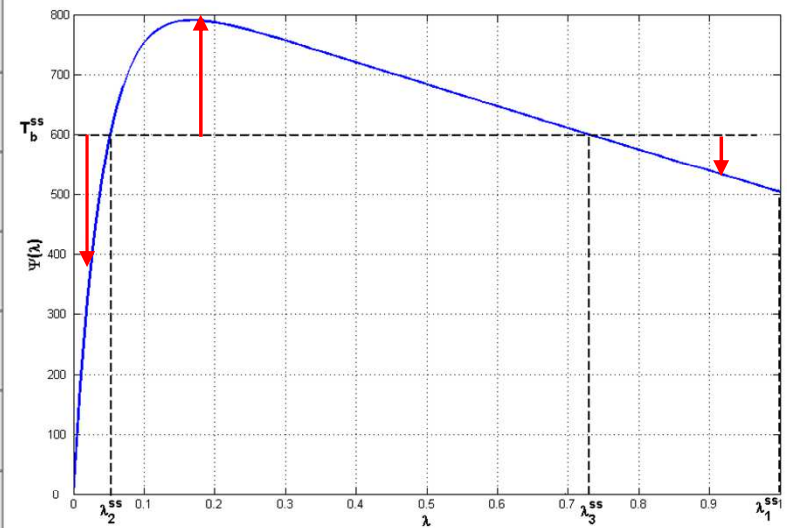
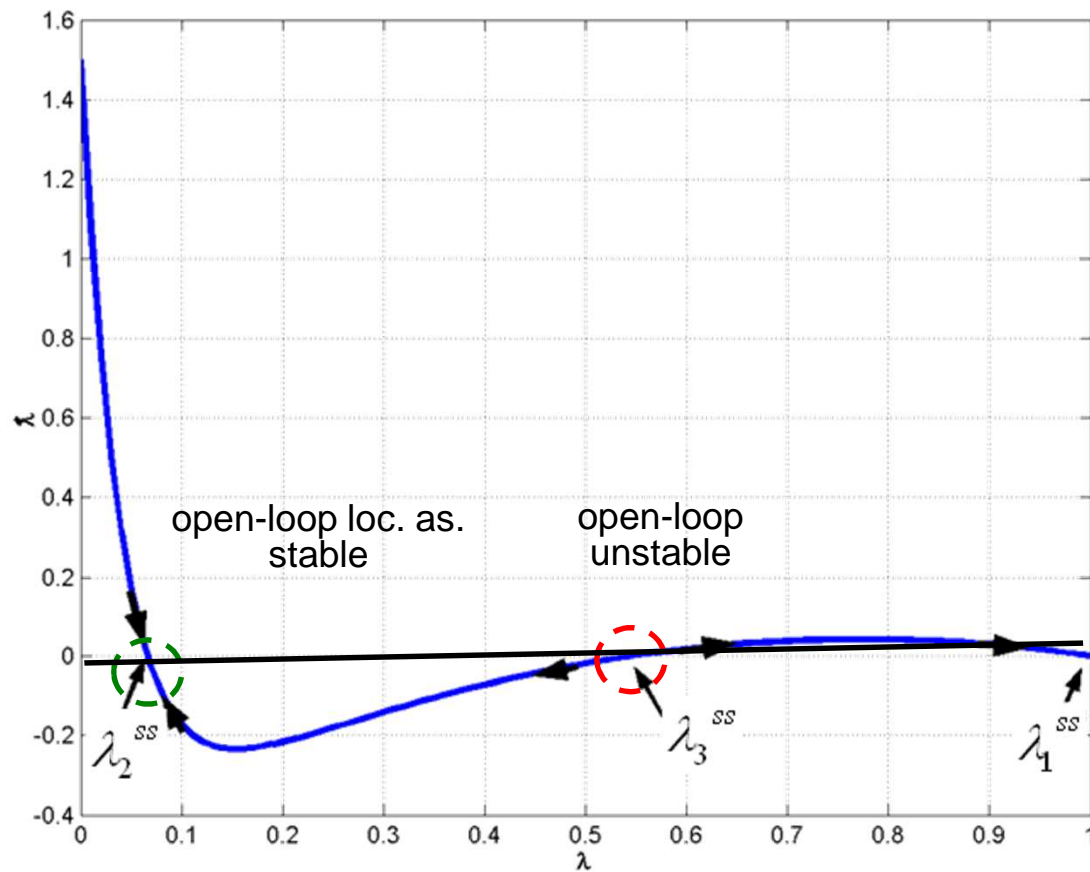
$$\lambda_2^{ss} = \bar{\lambda}_2, \quad \lambda_3^{ss} = \bar{\lambda}_3$$





Phase diagrams (open-loop)

$$\dot{\lambda} = -\frac{(1-\lambda)}{J\omega} (\psi(\lambda) - T_b)$$





Analysis of closed loop



$$\begin{cases} \dot{\lambda} = -\frac{1-\lambda}{\omega r} \left(\frac{(1-\lambda)}{m} + \frac{r^2}{J} \right) F_z \mu(\lambda) + \frac{r}{J} \frac{1-\lambda}{\omega r} T_b \\ \dot{T}_b = u \end{cases}$$

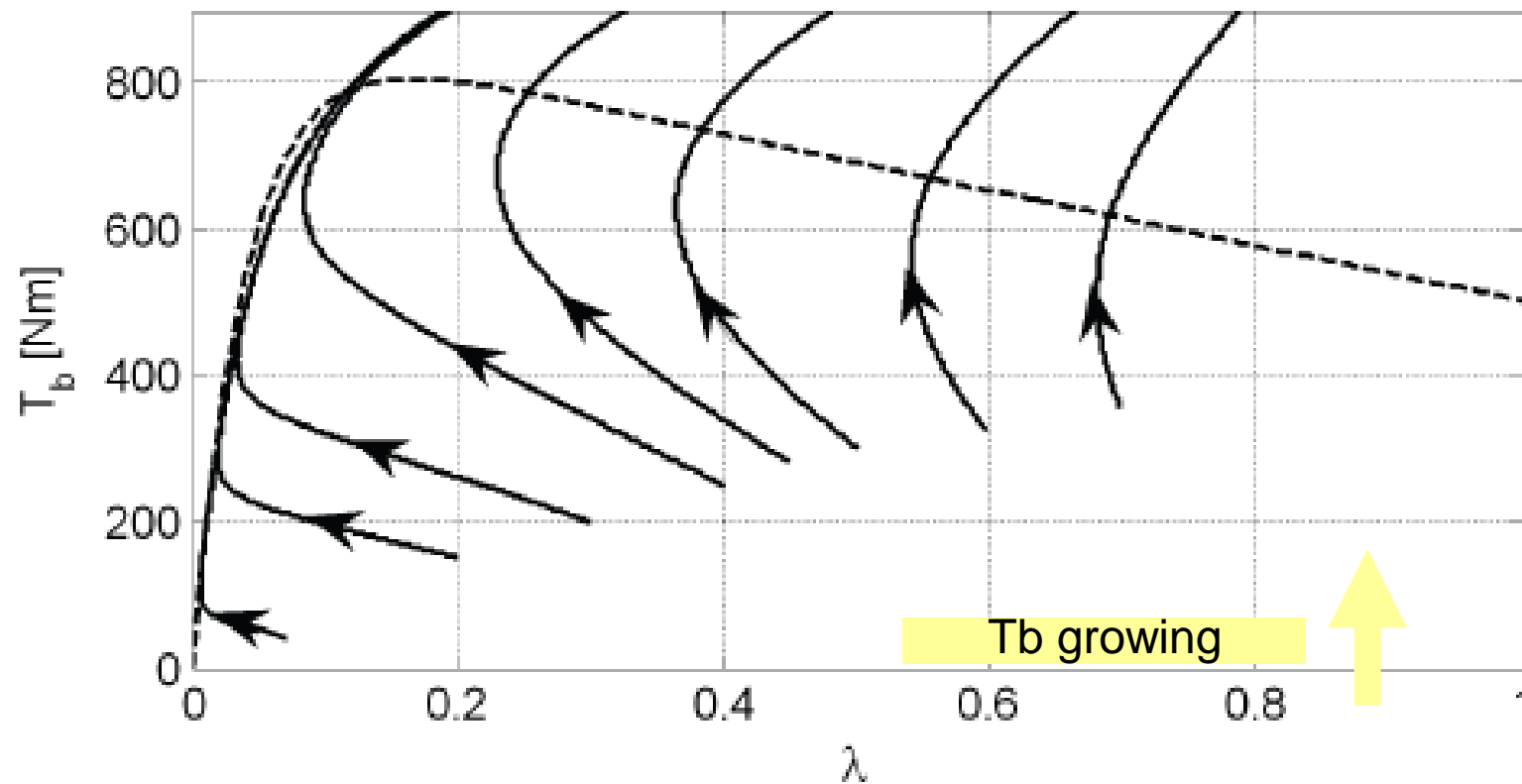
We analyze the phase diagrams: $u \in \{k, 0, -k\}$



Increase: $u = k$

$$\dot{\lambda} = -\frac{(1-\lambda)}{J\omega} (\psi(\lambda) - T_b)$$

System trajectories with control action $u = k$ (increase)

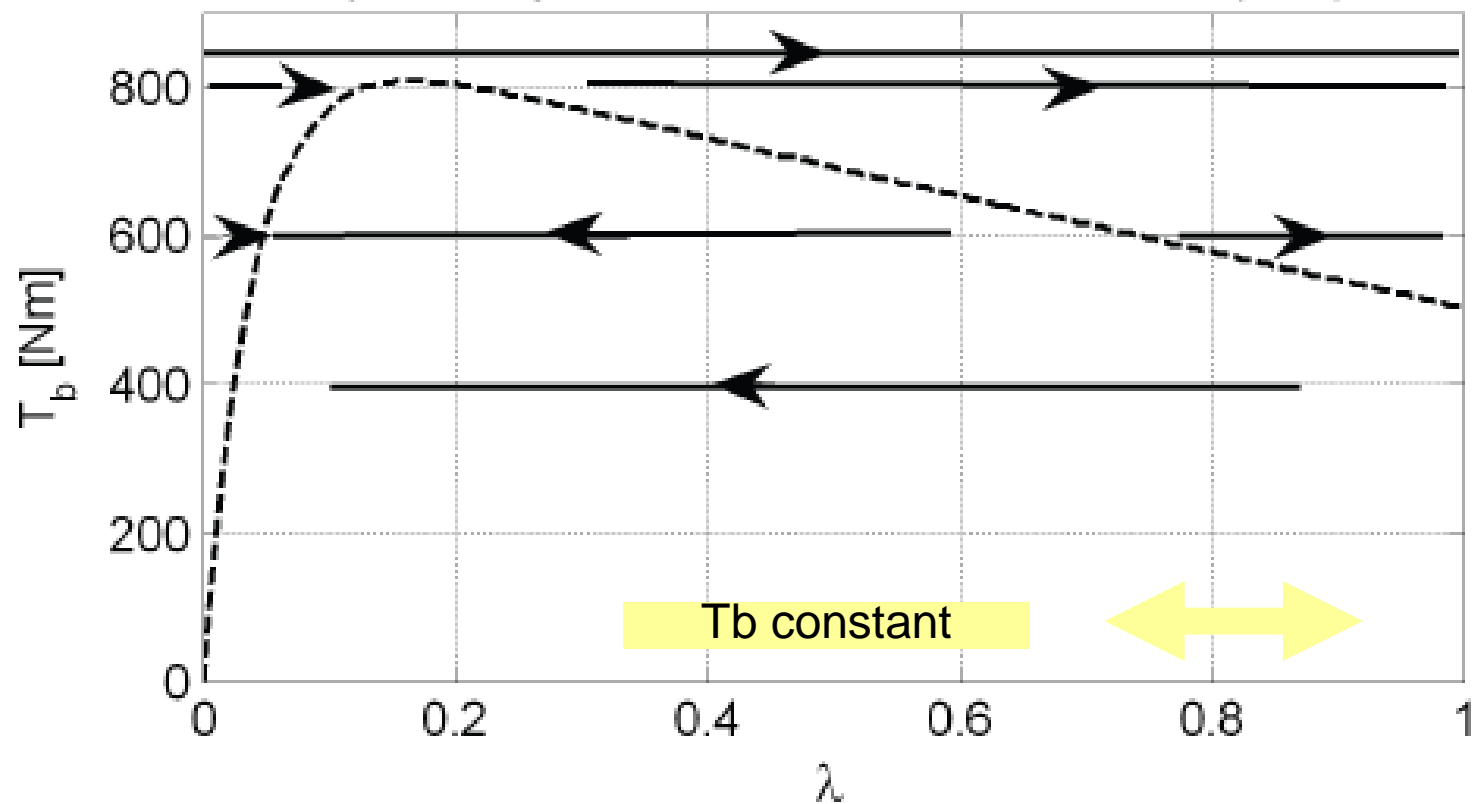




Hold: $u = 0$

$$\dot{\lambda} = -\frac{(1 - \lambda)}{J\omega} (\psi(\lambda) - T_b)$$

System trajectories with control action $u = 0$ (hold)

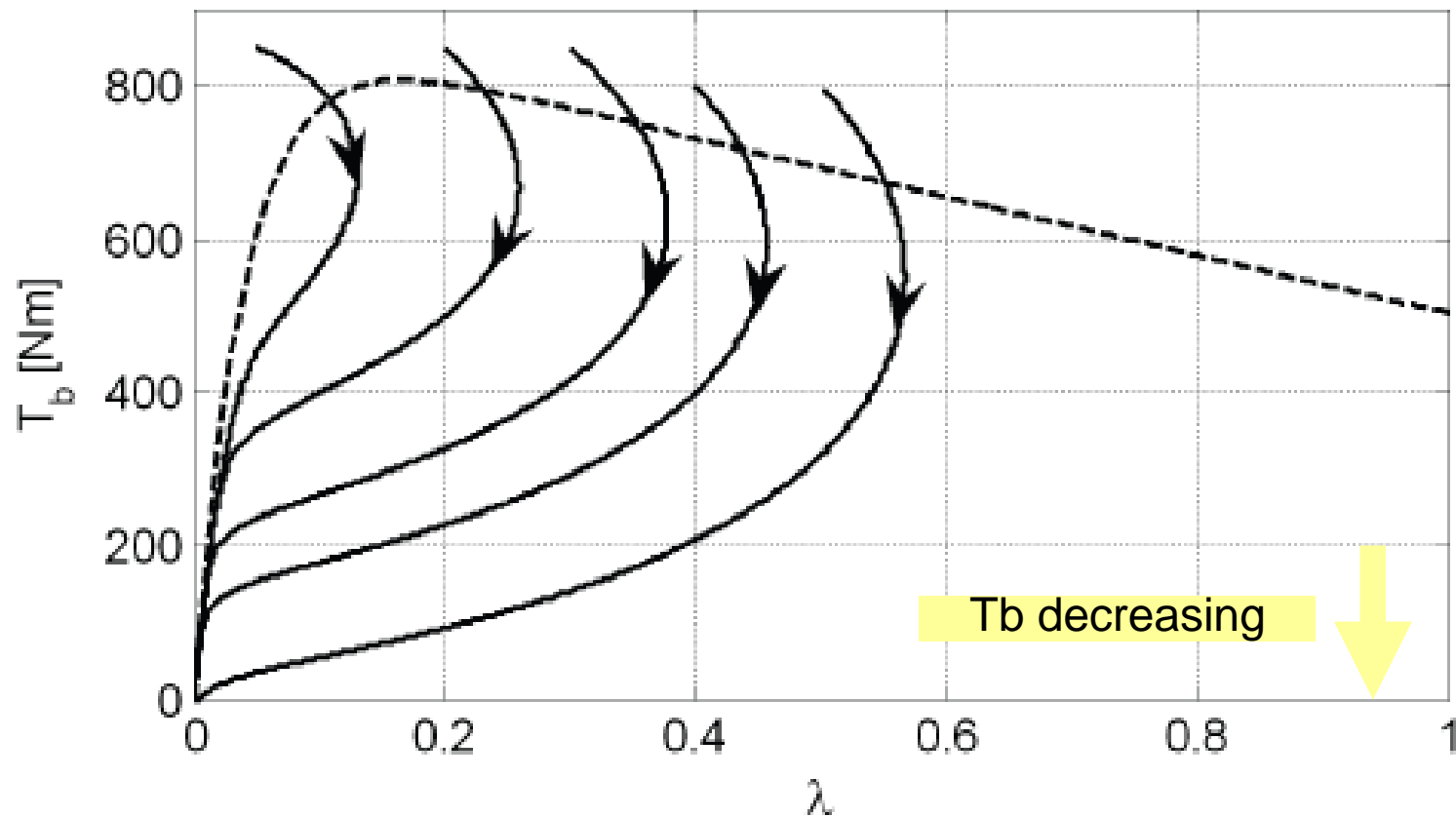




Decrease: $u = -k$

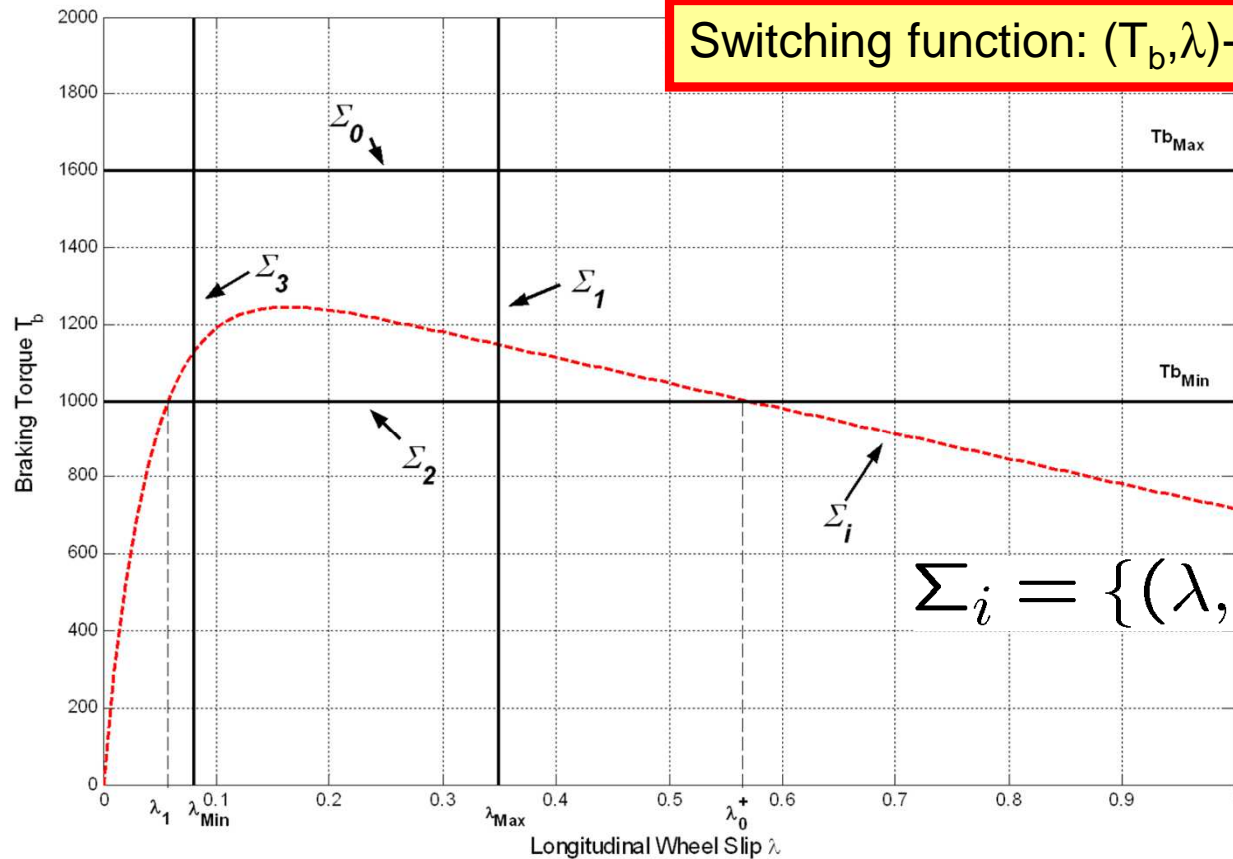
$$\dot{\lambda} = -\frac{(1 - \lambda)}{J\omega} (\psi(\lambda) - T_b)$$

System trajectories with control action $u = -k$ (decrease)





Switching Control Logic



$$\Sigma_i = \{(\lambda, T_b) : T_b = \psi(\lambda)\}$$

$$\Sigma_0 = \{(\lambda, T_b) : H_0(\lambda, T_b) := T_b - T_{bMax} = 0\},$$

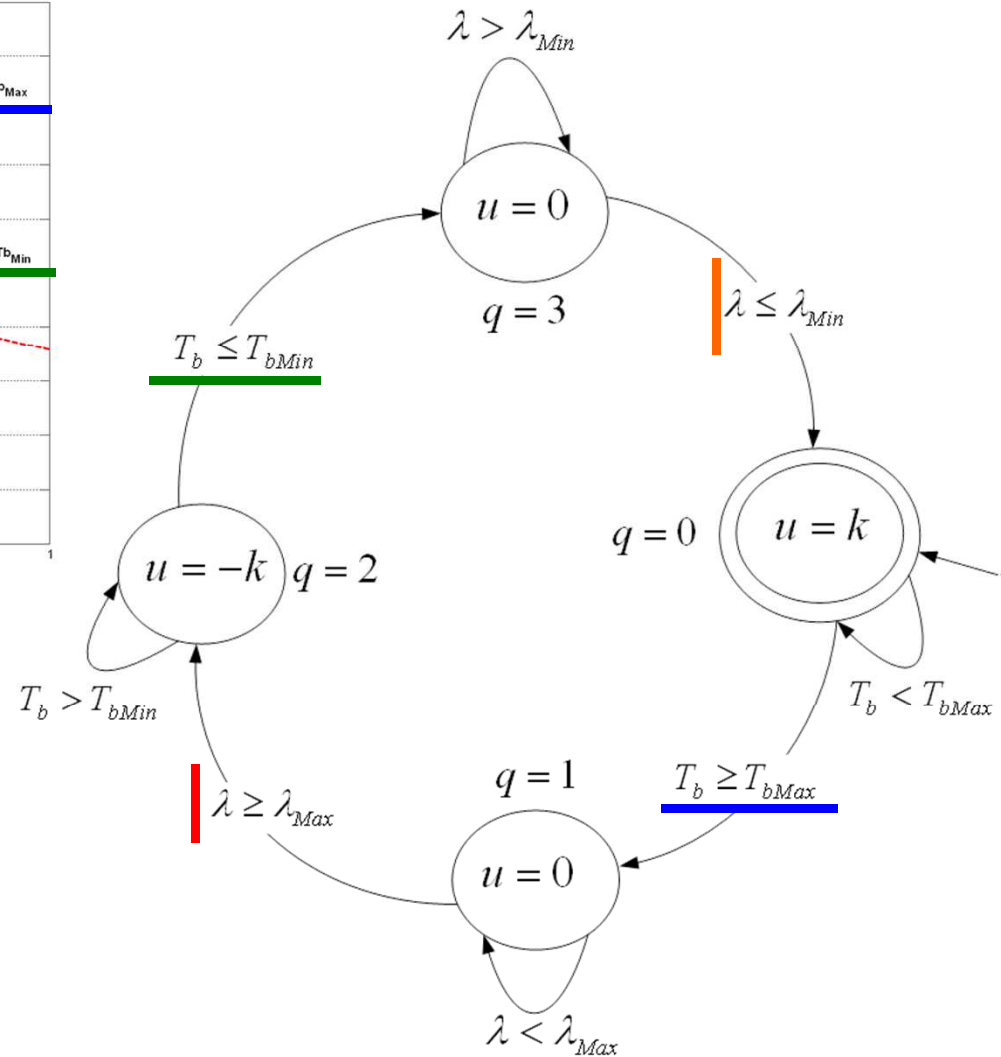
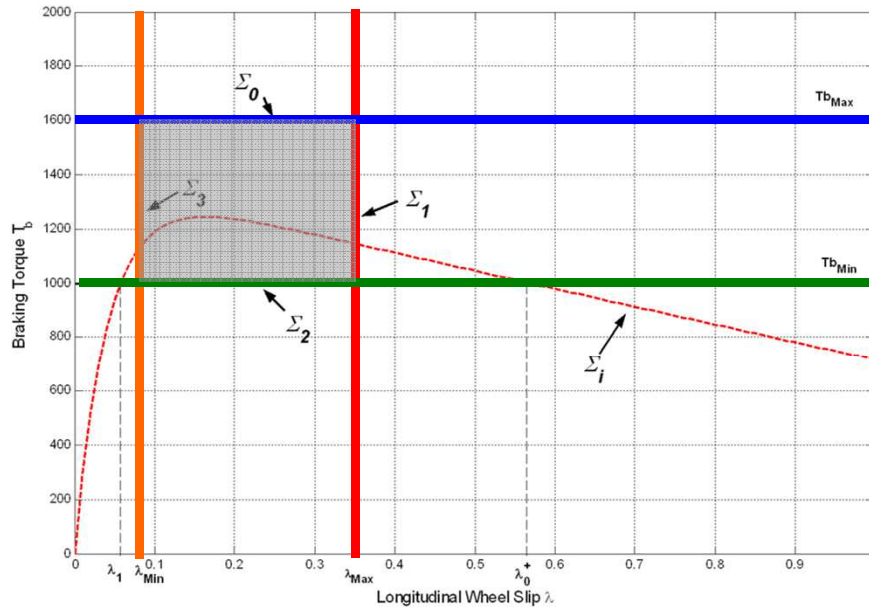
$$\Sigma_1 = \{(\lambda, T_b) : H_1(\lambda, T_b) := \lambda - \lambda_{Max} = 0\},$$

$$\Sigma_2 = \{(\lambda, T_b) : H_2(\lambda, T_b) := T_b - T_{bMin} = 0\},$$

$$\Sigma_3 = \{(\lambda, T_b) : H_3(\lambda, T_b) := \lambda - \lambda_{Min} = 0\}.$$

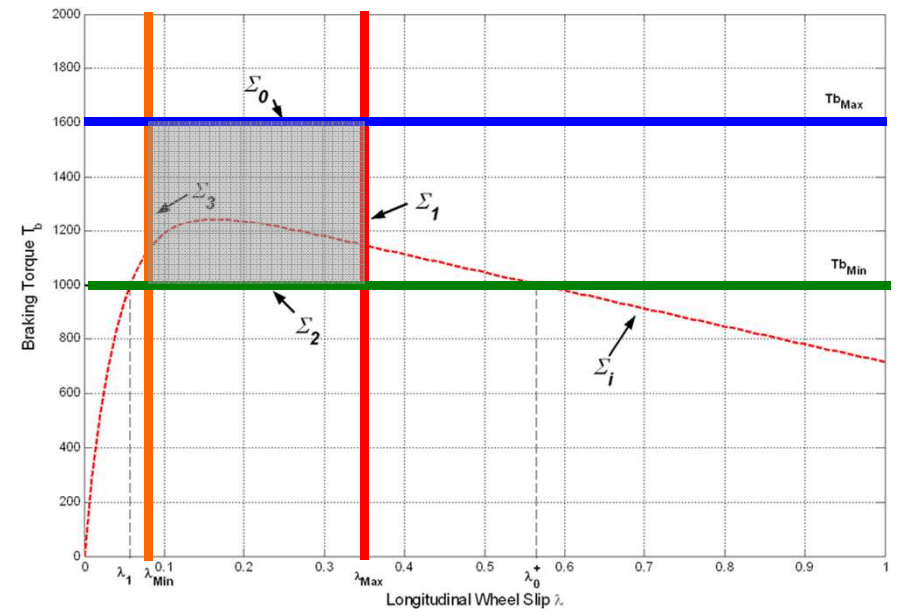
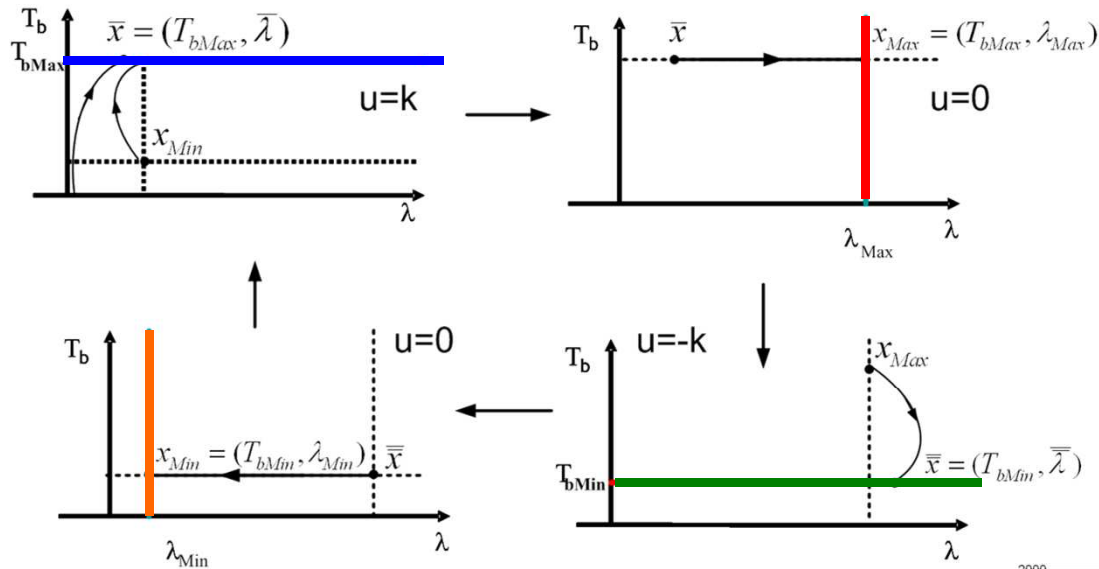


Switching Control Logic



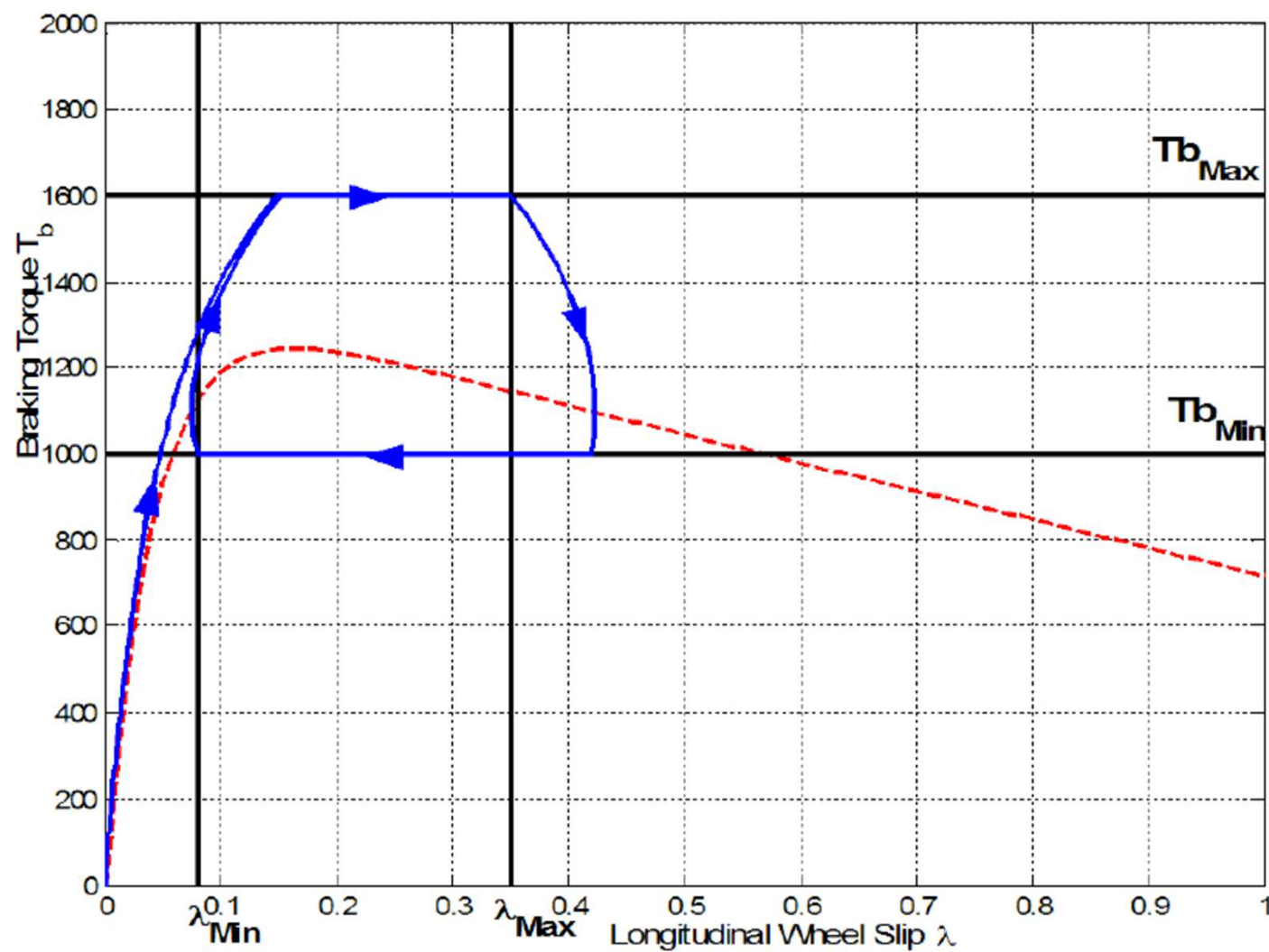


Limit-cycle: analysis



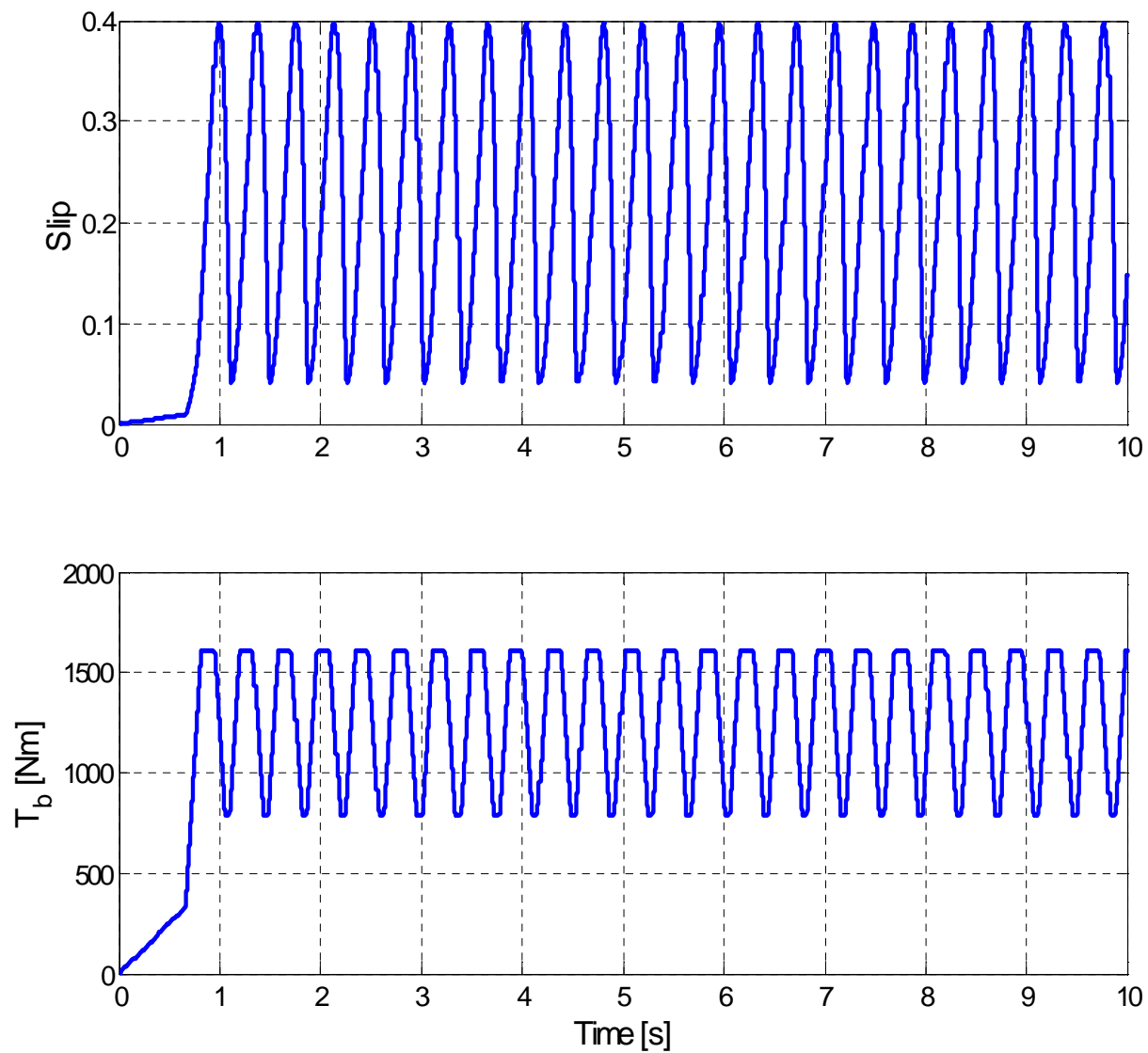


Limit-cycle: analysis





Limit-cycle: analysis





Limit-cycle: view in the hybrid space

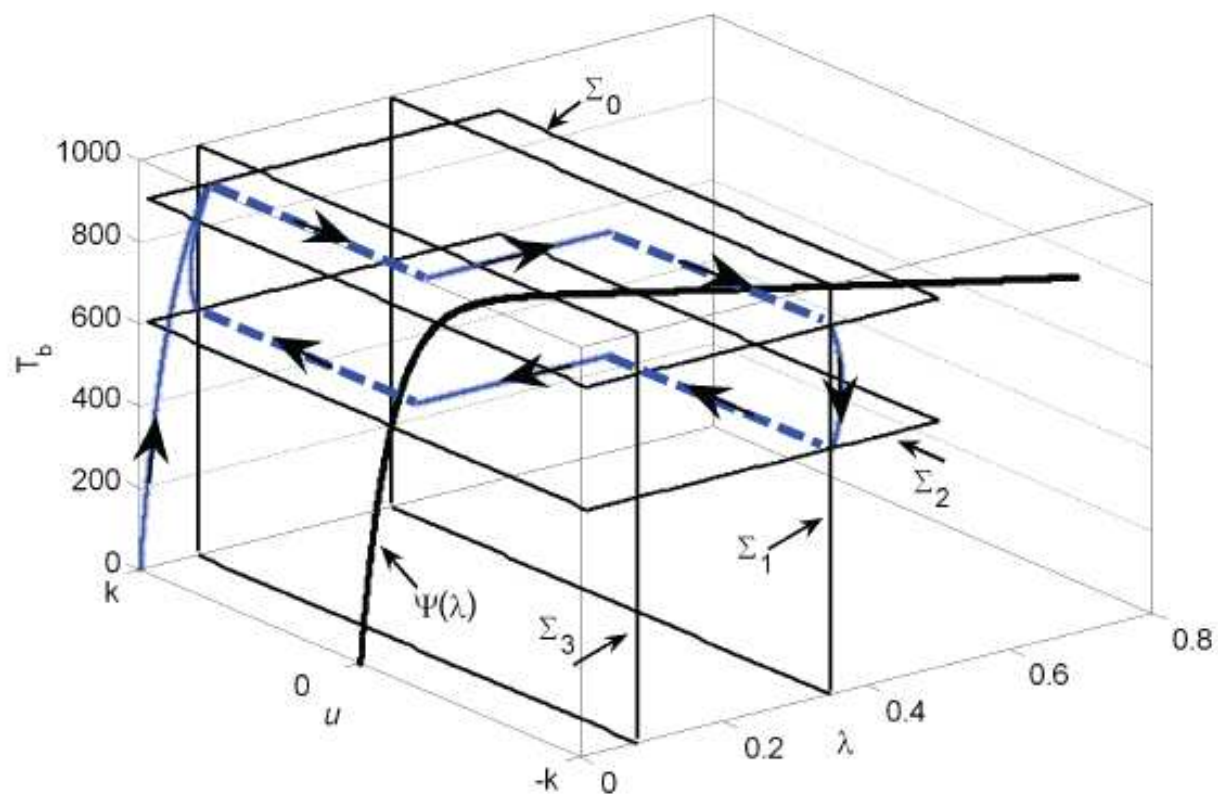


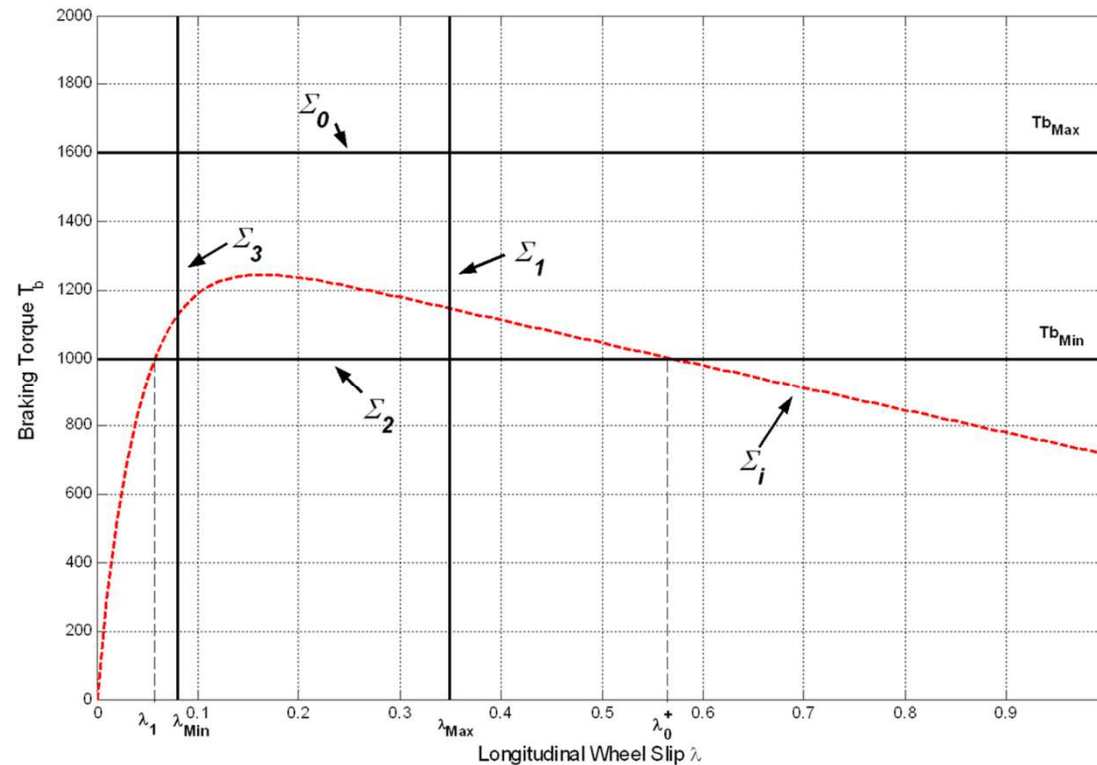
Figure 4.3 Simulated closed-loop trajectory of system (4.1) with the given control logic in the *hybrid* state space (T_b, λ, u)



Existence of limit cycles: necessary conditions

$$\Sigma_0 \cap \Sigma_i = \emptyset$$

$$\Sigma_2 \cap \Sigma_i \neq \emptyset$$



if $\Sigma_2 \cap \Sigma_i = \{(T_{bMin}, \lambda_1), (T_{bMin}, \lambda_0^+)\}$, $\lambda_1 < \lambda_0^+$
 then the intersection between the system trajectory

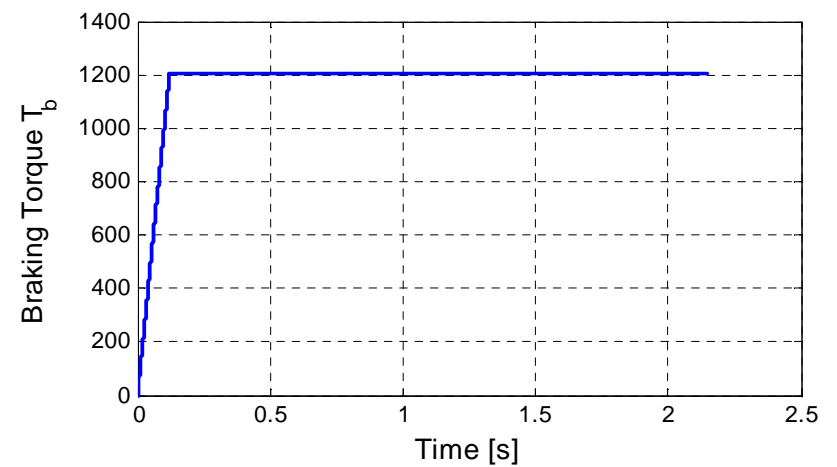
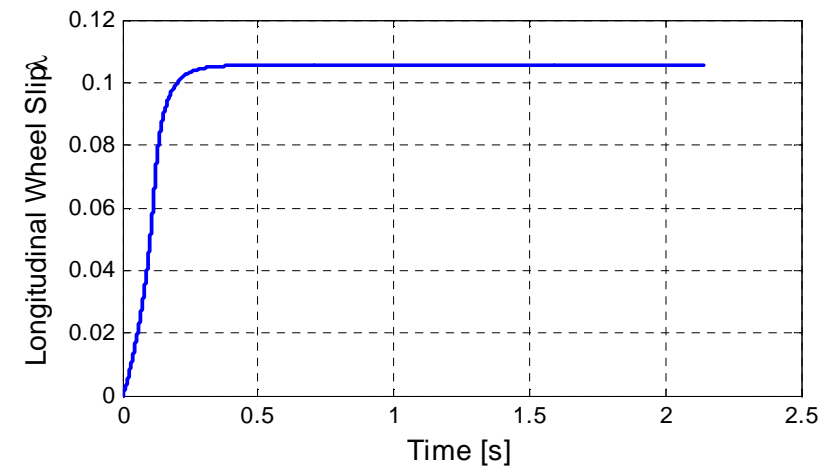
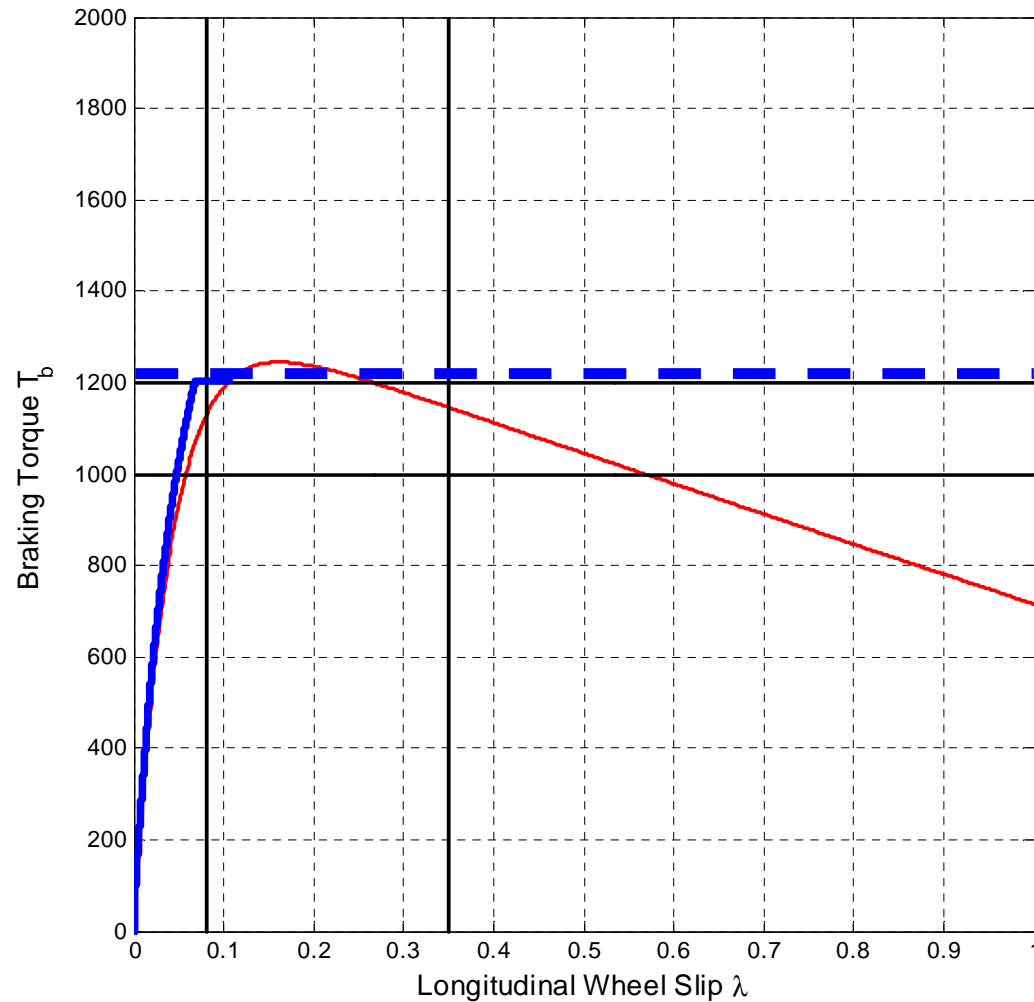
$$\Phi(t, [\lambda_{Max}, T_{bMax}; q = 2])$$

must occur at a point $(T_{bMin}, \bar{\lambda})$ such that $\bar{\lambda} < \lambda_0^+$



Limit-cycle: analysis

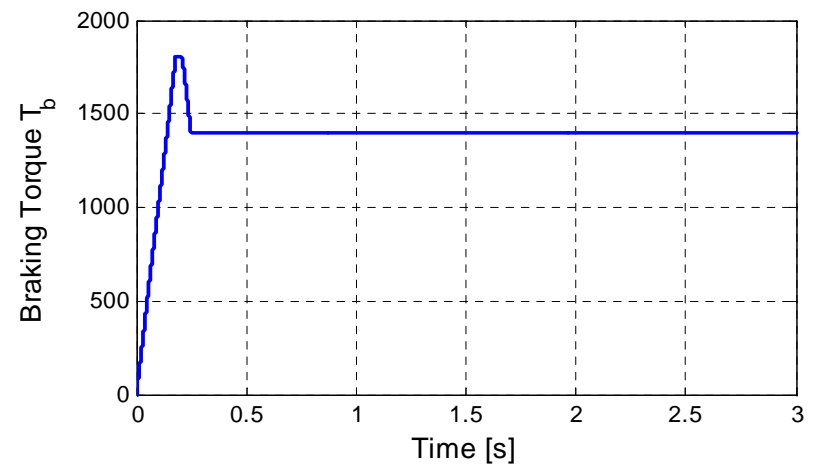
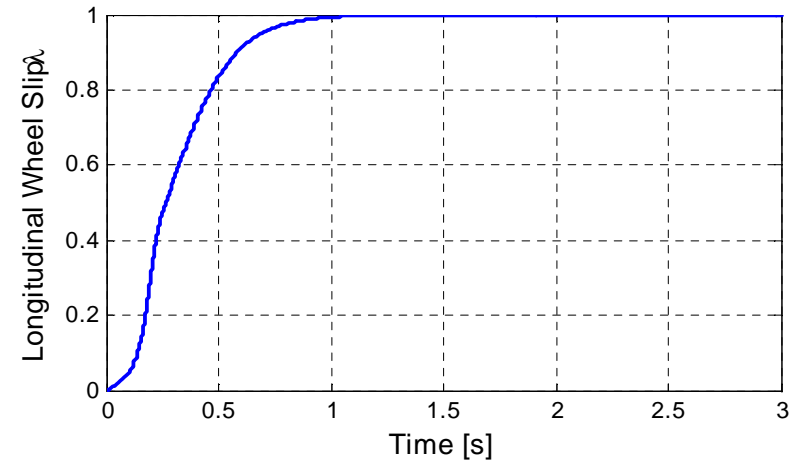
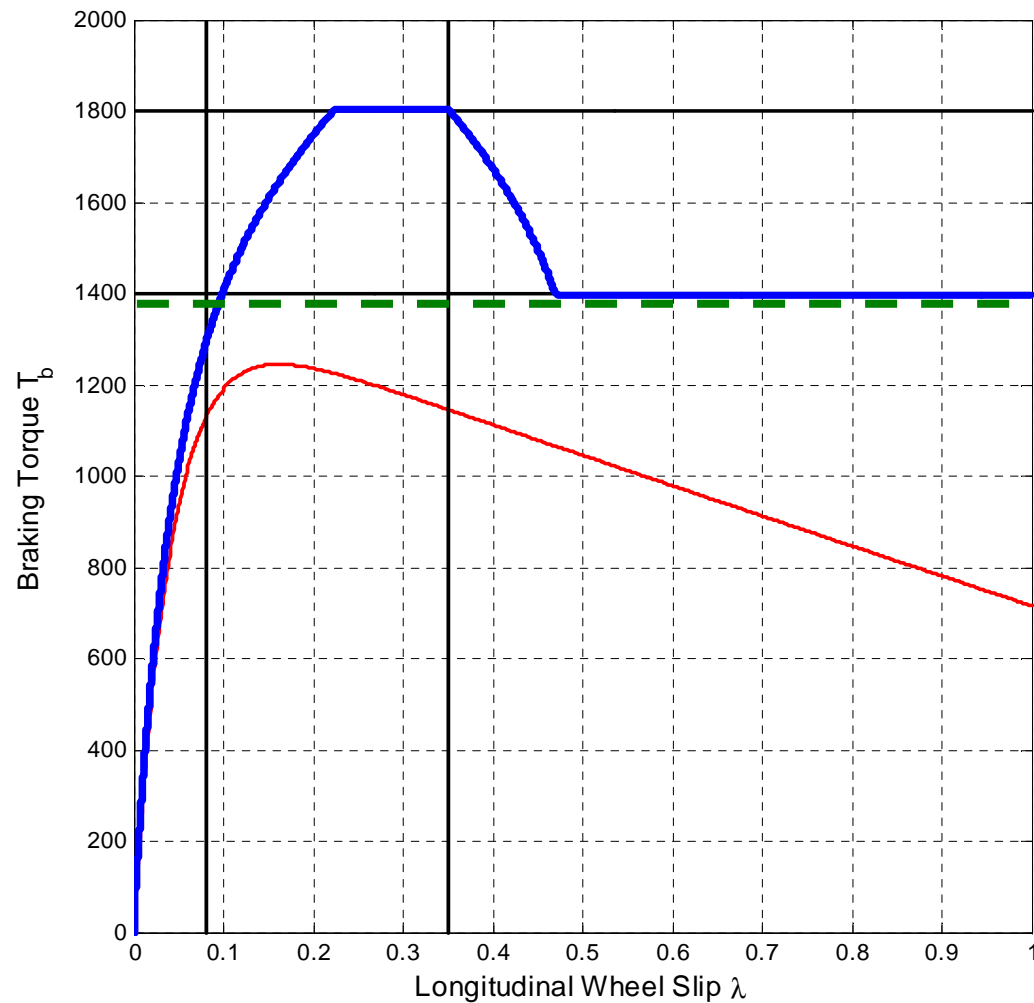
First condition violated: NOT a limit cycle but a steady-state condition





Limit-cycle: analysis

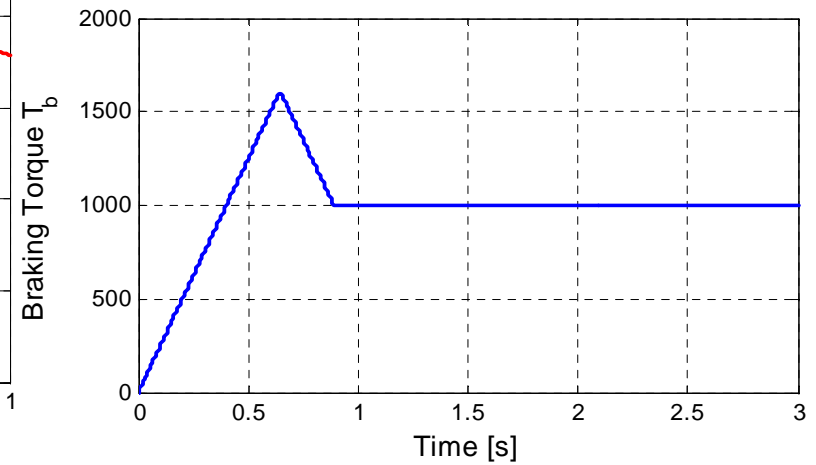
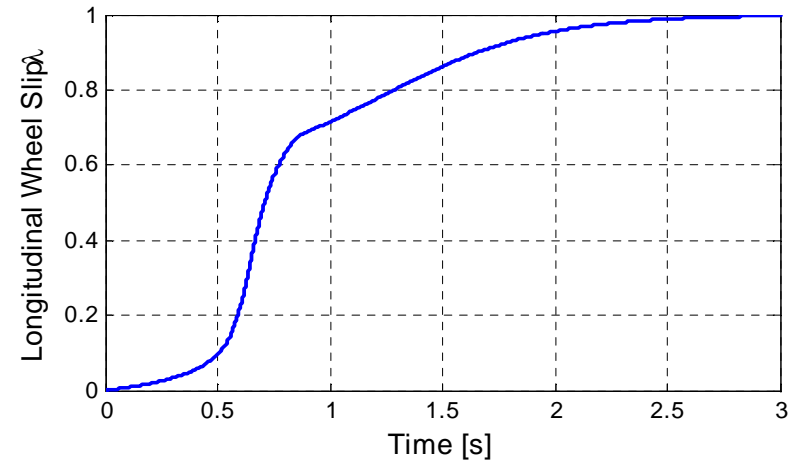
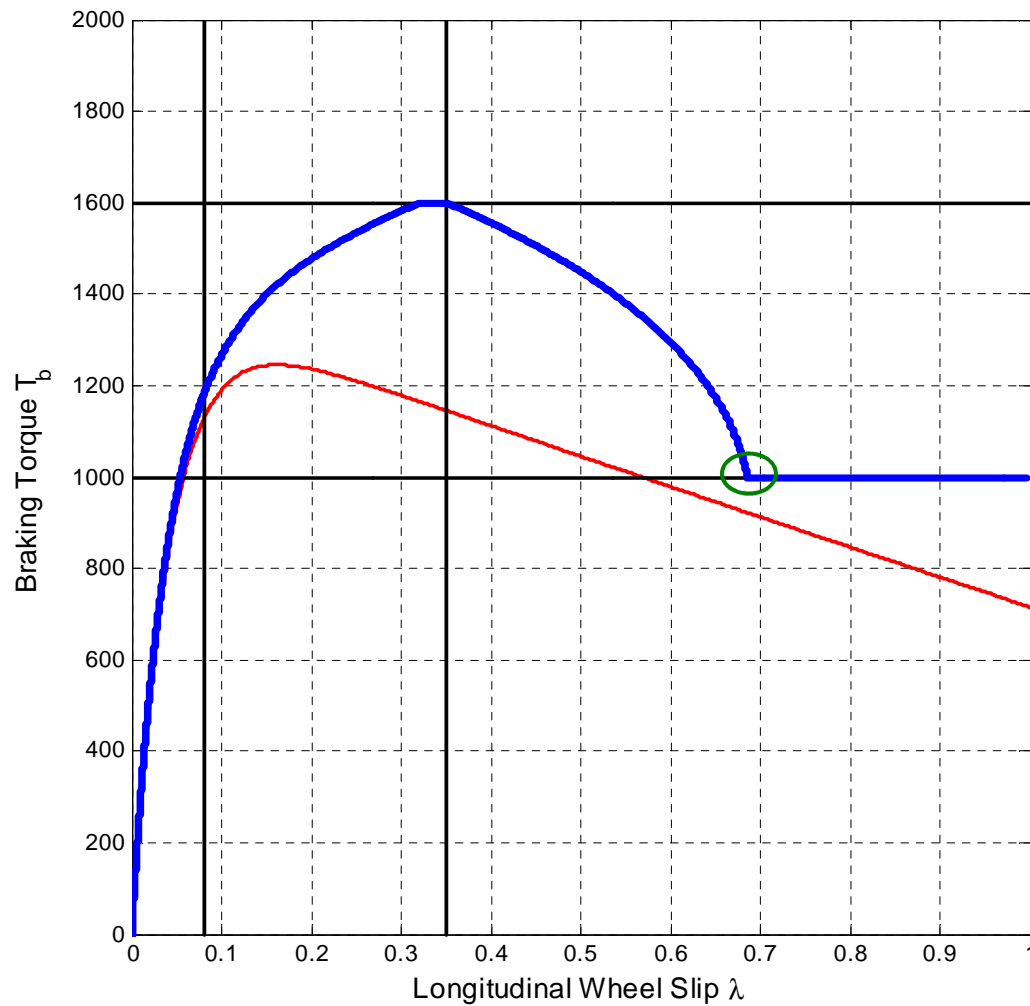
The second condition is violated: NOT a limit cycle but wheel locks





Limit-cycle: analysis

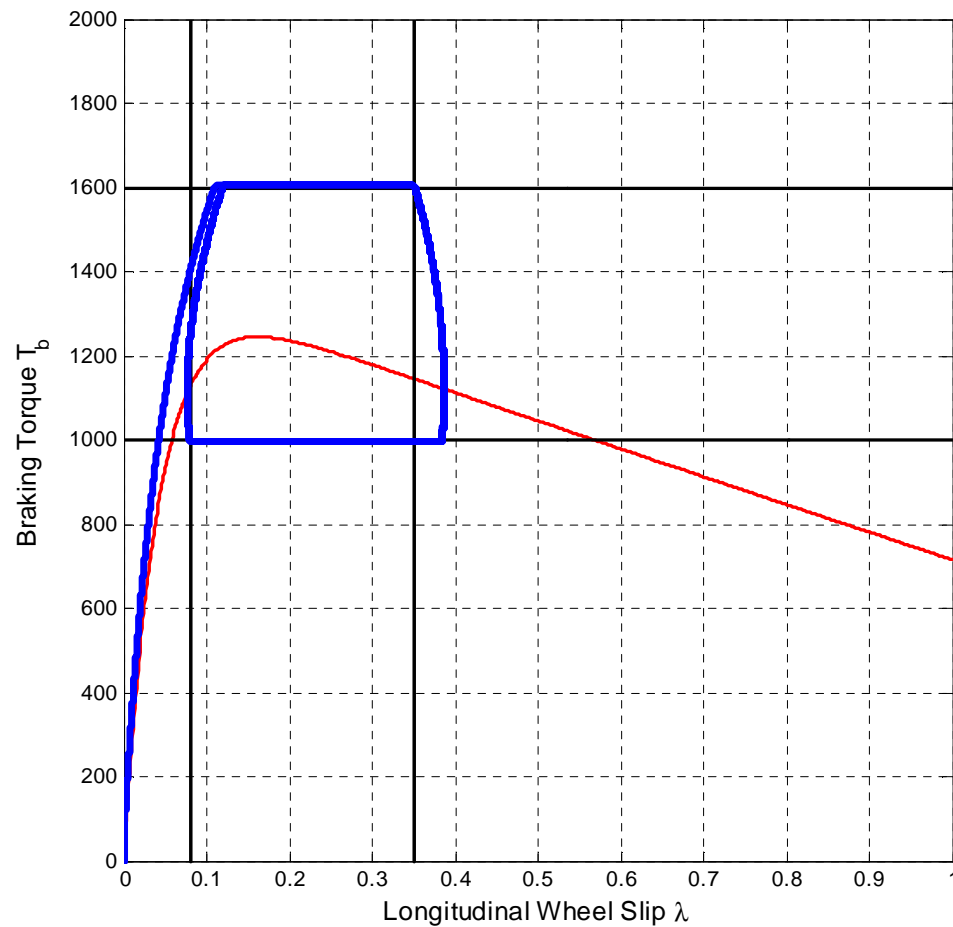
Third condition violated: NOT a limit cycle but wheel locks



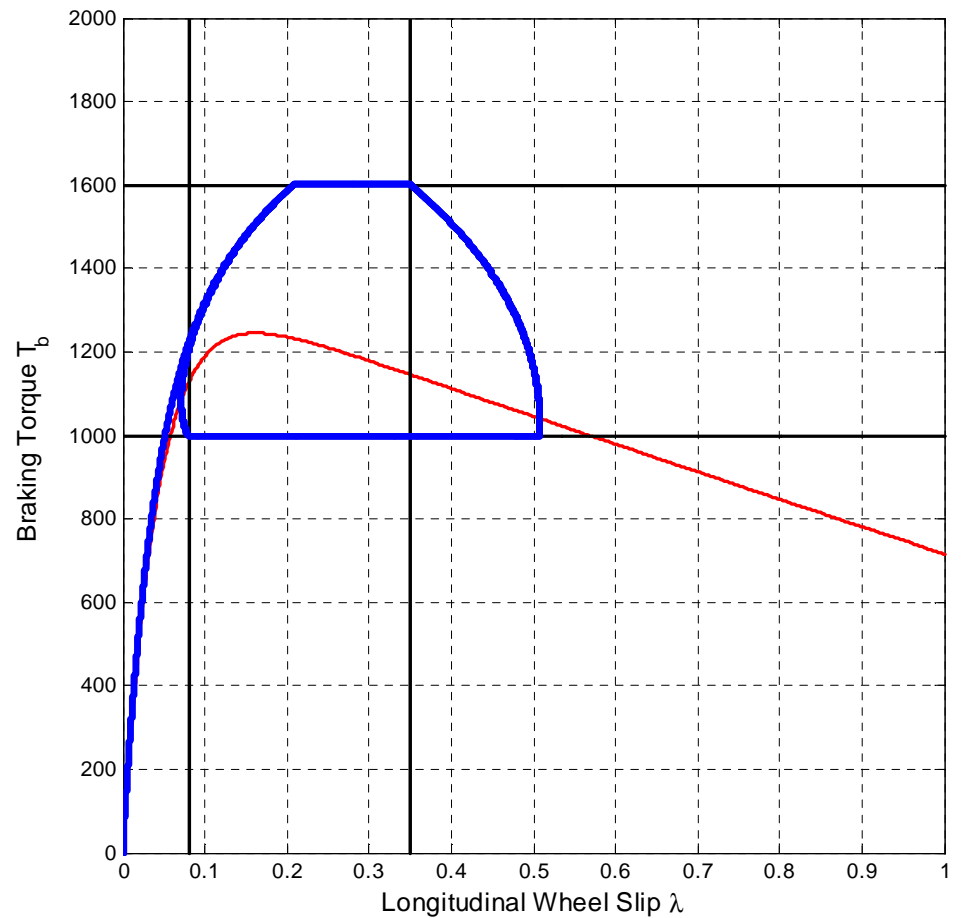


Limit-cycle: analysis

Impact of k large / small



K large

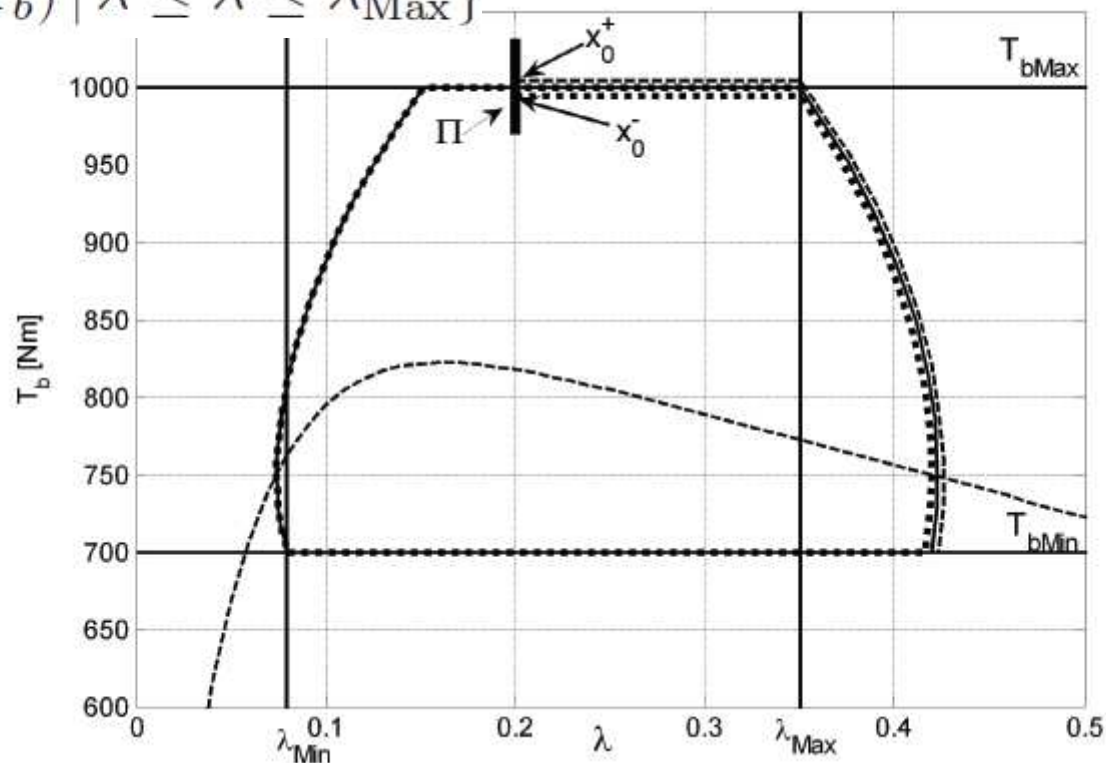


K small



Limit-cycle: stability

$$\Pi := \{(\lambda, T_b) \mid \bar{\lambda} \leq \lambda \leq \lambda_{\text{Max}}\}$$



$$x_{n+1} = P(x_n), \quad x_n \in \mathcal{B}_r(x^*) \cap \Pi$$



$$P(x_n) = x^* \quad \forall x_n \in \mathcal{B}_r(x^*) \cap \Pi$$



Comments

Ideal behavior:

"Box" very tight around the peak of the curve of friction
Transitions very fast (high-k)

In practice we have to find a compromise between:

k small / large

Box small (performance) / large (robustness)

Robustness (fixed) & Adaptation (varying) Box

