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Linearization: Crane System Study

```
In [ ]:
# %%

import sympy
from sympy import *
from sympy.physics.vector.printing import vlatex
from IPython.display import Math, display

init_printing()

def dotprint(expr):
    display(Math(vlatex(expr)))

In [ ]:
# %%

t = symbols('t')
# Generalized coordinates
z, h, theta = symbols('z, h, \theta', cls=Function)
z = z(t)
h = h(t)
theta = theta(t)

z_dot = z.diff(t)
h_dot = h.diff(t)
theta_dot = theta.diff(t)

z_ddot = z.diff(t,2)
h_ddot = h.diff(t,2)
theta_ddot = theta.diff(t,2)

mc, mr, Jc, d, mu, g, F, tau = symbols('m_c, m_r, J_c, d, \mu, g, F, \tau', real=True)

In [ ]:
# %%

M = Matrix([[mc+2*mr, 0, 0],
            [0, mc+2*mr, 0],
            [0, 0, Jc+2*mr*d**2]])

RHS = Matrix([[-F*sin(theta)-mu*z_dot],
              [-(mc+2*mr)*g+F*cos(theta)],
              [tau]])

EOM_dd = M.inv()*RHS
dynamics = Eq(Matrix([z_ddot, h_ddot, theta_ddot]), EOM_dd)
dotprint(dynamics)


$$\begin{bmatrix} \ddot{z} \\ \ddot{h} \\ \ddot{\theta} \end{bmatrix} = \begin{bmatrix} \frac{-F \sin(\theta) - \mu \dot{z}}{m_c + 2 m_r} \\ \frac{F \cos(\theta) + g(-m_c - 2 m_r)}{m_c + 2 m_r} \\ \frac{\tau}{J_c + 2 m_r d^2} \end{bmatrix}$$

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Deriving Nonlinear State Space Equations

Solve for our highest derivatives, \ddot{z} , \ddot{h} , $\ddot{\theta}$:

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solve_dict = solve(dynamics, (z_ddot, h_ddot, theta_ddot), simplify=True, dict=True)[0]
dotprint(solve_dict)
```

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$$\left\{ \ddot{\theta} : \frac{\tau}{J_c} + 2 d^2 m_r \right\}, \ddot{h} : \frac{F \cos(\theta)}{m_c + 2 m_r} - g m_c - 2 g m_r \right\}$$

$$\ddot{z} : - \frac{F \sin(\theta)}{m_c + 2 m_r} + \mu \dot{z} \right\}$$

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Create our state vector $x = [x_1, x_2, x_3, x_4, x_5, x_6]$:

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```
state = MatrixSymbol('x', 6, 1)
dotprint(state)
```

```

$$x$$

```

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We have $x = [x_1, x_2, x_3, x_4, x_5, x_6] = [z, h, \theta, \dot{z}, \dot{h}, \dot{\theta}]$, so $\dot{x} = [\dot{z}, \dot{h}, \dot{\theta}, \ddot{z}, \ddot{h}, \ddot{\theta}]$. Let's put that in a vector:

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```
state_deriv = Matrix([
    z_dot,
    h_dot,
    theta_dot,
    solve_dict[z_ddot],
    solve_dict[h_ddot],
    solve_dict[theta_ddot]
])
dotprint(state_deriv)

$$\left[ \begin{matrix} \dot{z} \\ \dot{h} \\ \dot{\theta} \\ - \frac{F \sin(\theta)}{m_c + 2 m_r} + \mu \dot{z} \\ \frac{F \cos(\theta)}{m_c + 2 m_r} - g m_c - 2 g m_r \\ \frac{\tau}{J_c} + 2 d^2 m_r \end{matrix} \right]$$

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Now we can substitute in our $x_1, x_2, x_3, x_4, x_5, x_6$ values:

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In [ ]:
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# Dictionary for substitutions
```

```
subs_dict = {
    z: state[0],
    h: state[1],
    theta: state[2],
    z_dot: state[3],
    h_dot: state[4],
    theta_dot: state[5]
}
```

```
f_expr = state_deriv.subs(subs_dict)
dotprint(f_expr)
```

```

$$\left[ \begin{matrix} x_{3,0} \\ x_{4,0} \\ x_{5,0} \\ \frac{F \sin(x_{2,0})}{m_c + 2 m_r} + \mu x_{3,0} \\ \frac{F \cos(x_{2,0})}{m_c + 2 m_r} - g m_c - 2 g m_r \\ \frac{\tau}{J_c} + 2 d^2 m_r \end{matrix} \right]$$

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We now have our nonlinear state space equations!

Linearization

First, we need to find an equilibrium point. We can either do this by hand or use Sympy's `solve` function to do this.

Note: It's a good idea to practice doing this by hand in addition to using Sympy!

```
In [ ]:
# %%

# This equation represents  $f(x,u) = 0$ 
equilibrium_equation = Eq(f_expr, Matrix([0, 0, 0, 0, 0, 0]))

eq_solve_dict = solve(equilibrium_equation, (state[0], state[1], state[2], state[3], state[4], state[5], F, tau), simplify=True, dict=True)[0]
dotprint(eq_solve_dict)


$$\left( F : -g \left( m_c + 2 m_r \right), \tau : 0, x_{2,0} : \pi, x_{3,0} : 0, x_{4,0} : 0, x_{5,0} : 0 \right)$$

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We can see that at equilibrium:

1. z and h can be any value
2. $\theta = 0$ or π (we'll use 0)
3. All velocities are zero
4. $F = (m_c + 2m_r)g$
5. $\tau = 0$

Below we define our equilibrium points $x_e = [z_e, h_e, \theta_e, \dot{z}_e, \dot{h}_e, \dot{\theta}_e]$ and $u_e = [F_e, \tau_e]$.

```
In [ ]:
# %%

z_e, h_e = symbols('z_e h_e')
u_eq = Matrix([(m_c + 2*m_r)*g, 0])
theta_eq = 0
z_dot_eq = h_dot_eq = theta_dot_eq = 0
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Define A, B Jacobians

We can use Sympy's `jacobian` function to find the jacobians of $f(x,u)$.

First we find $A = \frac{\partial f}{\partial x}$:

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In [ ]:
# %%

A = f_expr.jacobian(state)
dotprint(A)


$$\begin{bmatrix} 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & -\frac{F}{\cos(\left( x_{2,0} \right))} (m_c + 2 m_r) & -\frac{\mu}{\cos(\left( x_{2,0} \right))} (m_c + 2 m_r) & 0 & 0 & 0 & 0 & -\frac{F \sin(\left( x_{2,0} \right))}{m_c + 2 m_r} & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

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We can evaluate at our specific (x_e, u_e) point using `subs`.

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In [ ]:
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# %%%

A_subs = {
    state[0]: z_e,
    state[1]: h_e,
    state[2]: theta_eq,
    state[3]: z_dot_eq,
    state[4]: h_dot_eq,
    state[5]: theta_dot_eq,
    F: u_eq[0],
    tau: u_eq[1]
}

A_eq = A.subs(A_subs)
dotprint(A_eq)


$$\begin{bmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} m_c + 2 m_r \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} -g \\ -\frac{\mu}{m_c + 2 m_r} \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

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Now we do a similar process to find $B = \frac{\partial f}{\partial u}$

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In [ ]:
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B = f_expr.jacobian(Matrix([F, tau]))
dotprint(B)


$$\begin{bmatrix} 0 & 0 & 0 & 0 & -\frac{\sin(x_{2,0})}{m_c + 2 m_r} & 0 \\ 0 & 0 & 0 & 0 & \frac{\cos(x_{2,0})}{m_c + 2 m_r} & 0 \\ 0 & 0 & 0 & 0 & 0 & \frac{1}{J_c + 2 d^2} \end{bmatrix} \begin{bmatrix} m_c + 2 m_r \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

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Like for the A matrix, we need to substitute in our equilibrium values.

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In [ ]:
# %%%

B_subs = {
    state[0]: z_e,
    state[1]: h_e,
    state[2]: theta_eq,
    state[3]: z_dot_eq,
    state[4]: h_dot_eq,
    state[5]: theta_dot_eq,
    F: u_eq[0],
    tau: u_eq[1]
}

B_eq = B.subs(B_subs)
dotprint(B_eq)


$$\begin{bmatrix} 0 & 0 & 0 & 0 & -\frac{1}{m_c + 2 m_r} & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{J_c + 2 d^2} & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} m_c + 2 m_r \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

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Decoupled Dynamics

We can separate the longitudinal and lateral dynamics:

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A_lin_lon = A_eq[[1, 4], [1, 4]]

B_lin_lon = B_eq[[1, 4], [0]]

print("Linearized A matrix for longitudinal dynamics:")
dotprint(A_lin_lon)

print("Linearized B matrix for longitudinal dynamics:")
dotprint(B_lin_lon)

A_lin_lat = A_eq[[0, 2, 3, 5], [0, 2, 3, 5]]

B_lin_lat = B_eq[[0, 2, 3, 5], [1]]

print("Linearized A matrix for lateral dynamics:")
dotprint(A_lin_lat)

print("Linearized B matrix for lateral dynamics:")
dotprint(B_lin_lat)

Linearized A matrix for longitudinal dynamics:

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Linearized B matrix for longitudinal dynamics:

$$\begin{bmatrix} 0 \\ \frac{1}{m_c} + 2 \frac{m_r}{m_c} \end{bmatrix}$$

Linearized A matrix for lateral dynamics:

$$\begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & -g & -\frac{\mu}{m_c} + 2 \frac{m_r}{m_c} \end{bmatrix}$$

Linearized B matrix for lateral dynamics:

$$\begin{bmatrix} 0 \\ 0 \\ \frac{1}{J_c} + 2 \frac{d^2}{m_r} \end{bmatrix}$$

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Final Equations

To obtain our final equations of motion, we define new variables measuring our offset from the equilibrium:

$$\tilde{x} = x - x_e \quad \tilde{u} = u - u_e$$

Our final equations of motion are:

$$\dot{\tilde{x}} = A\tilde{x} + B\tilde{u}$$

Where A and B are the linearized matrices we derived above, and the system is decoupled into longitudinal and lateral dynamics.