Robotics

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Chapter 1

Dynamics

1.1 Kinematics

Take a vector with components $\mathbf{P} = (a_x, b_y, c_z)$. A scale factor w can be added (to the matrix form) to give

$$\mathbf{P} = \begin{bmatrix} P_x \\ P_y \\ P_z \\ w \end{bmatrix} \tag{1.1}$$

where $(a_x, b_y, c_z) = (P_x/w, P_y/w, P_z/w)$. A direction vector can be represented by a scale factor of zero (w = 0).

A universe reference frame is represented by $F_{x,y,z}$ and a moving frame is represented by $F_{n,o,a}$ where the letters n, o, and a come from the words normal, orientation and approach. Relative to the gripper, the z-axis is the approach axis by which the gripper approaches an object. The orientation with which the gripper frame approaches the part is the orientation axis. The normal-axis or x-axis is normal to both. A fourth vector which gives the location of a frame relative to a reference frame can be added to the vectors representing the components of the n-, o-, and a-axes to give a homogeneous matrix representation of this relative frame

$$F = \begin{bmatrix} 1 & 0 & 0 & d_x \\ 0 & 1 & 0 & d_y \\ 0 & 0 & 1 & d_z \\ 0 & 0 & 0 & 1 \end{bmatrix}$$
 (1.2)

Pre-multiplying the frame matrix by the transformation matrix will yield the new location of the frame. Rotation matrices about the x-, y- and z-axes are given by

$$rot(x,\theta) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos\theta & \sin\theta \\ 0 & \sin\theta & \cos\theta \end{bmatrix}
rot(y,\theta) = \begin{bmatrix} \cos\theta & 0 & \sin\theta \\ 0 & 1 & 0 \\ -\sin\theta & 0 & \cos\theta \end{bmatrix}
rot(z,\theta) = \begin{bmatrix} \cos\theta & -\sin\theta & 0 \\ \sin\theta & \cos\theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$$
(1.3)

Denoting the transformation of frame R relative to frame U (universe) as ${}^{U}T_{R}$, denoting the p relative to the frame R as ${}^{R}p=p_{noa}$, and denoting p relative to frame U as ${}^{U}p=p_{xyz}$ we have

$${}^{U}p = {}^{U}T_R \times {}^{R}p \tag{1.4}$$

- 1.2 Differential Motion
- 1.3 Dynamic Analysis
- 1.4 Trajectory Planning

Chapter 2

Controls

2.1 State Variable Representation

Powerful tools from matrix algebra can be used to solve sets of first-order differential equations. That is why it is sometimes helpful to transform a system described by *n*th-order differential equations into a system of first-order differential equations.

2.1.1 Systems Modeled by Linear Differential Equations

Consider a system represented by a nth-order, single-input linear constant coefficient differential equation

$$\sum_{i=0}^{n} a_i \frac{d^i y}{dt^i} = u \tag{2.1}$$

This equation can be replaced by n first-order differential equations

$$\begin{cases} \frac{dx_k}{dt} = x_{k+1}, \ 1 < k < n \\ \frac{dx_n}{dt} = \frac{1}{a_n} \left[\sum_{k=0}^{n-1} a_k x_{k+1} \right] + \frac{1}{a_n} u \end{cases}$$
 (2.2)

where $x_1 \equiv y$ and $i, k \in \mathbb{W}$. This can be written as a matrix equation

$$\begin{bmatrix}
\frac{dx_1}{dt} \\
\frac{dx_2}{dt} \\
\vdots \\
\frac{dx_n}{dt}
\end{bmatrix}
\begin{bmatrix}
0 & 1 & 0 & \cdots & 0 \\
0 & 0 & 1 & \cdots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
-\frac{a_0}{a_n} & -\frac{a_1}{a_n} & -\frac{a_2}{a_n} & \cdots & -\frac{a_{n-1}}{a_n}
\end{bmatrix}
\begin{bmatrix}
x_1 \\
x_2 \\
\vdots \\
x_n
\end{bmatrix} +
\begin{bmatrix}
0 \\
0 \\
\vdots \\
0 \\
\frac{1}{a_n}
\end{bmatrix} u$$
(2.3)

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$$\frac{d\mathbf{x}}{dt} = A\mathbf{x} + \mathbf{b}u\tag{2.4}$$

A multi-input-multi-output (MIMO) system can be represented by

$$\begin{bmatrix}
\frac{dx_1}{dt} \\
\frac{dx_2}{dt} \\
\vdots \\
\frac{dx_n}{dt}
\end{bmatrix}
\begin{bmatrix}
a_{11} & a_{12} & \cdots & a_{1n} \\
a_{21} & a_{22} & \cdots & a_{2n} \\
\vdots & \vdots & \ddots & \vdots \\
a_{n1} & a_{n2} & \cdots & a_{nn}
\end{bmatrix}
\begin{bmatrix}
x_1 \\
x_2 \\
\vdots \\
x_n
\end{bmatrix} +
\begin{bmatrix}
b_{11} & b_{12} & \cdots & b_{1r} \\
b_{21} & b_{22} & \cdots & b_{2r} \\
\vdots & \vdots & \ddots & \vdots \\
b_{n1} & b_{n2} & \cdots & b_{nr}
\end{bmatrix}
\begin{bmatrix}
u_1 \\
u_2 \\
\vdots \\
u_r
\end{bmatrix}$$
(2.5)

or

$$\frac{d\mathbf{x}}{dt} = A\mathbf{x} + B\mathbf{u} \tag{2.6}$$

where \mathbf{u} is an r-vector of input functions.

Let Φ be the $n \times n$ transition matrix of the differential equation given above which is described by the matrix equation

$$\frac{d\mathbf{\Phi}}{dt} = A\mathbf{\Phi} \tag{2.7}$$

If $\Phi(0) = I$ (initial condition) then $\Phi(t) = e^{At}$ where

$$e^{At} = \sum_{n=0}^{\infty} \frac{A^n t^n}{n!} \tag{2.8}$$

The solution to (2.6) on the interval $0 \le t < \infty$ is given by

$$\mathbf{x}(t) = e^{At}\mathbf{x}(0) + \int_0^t e^{A(t-\tau)}B\mathbf{u}(\tau) d\tau$$
 (2.9)

2.1.2 Systems Modeled by Constant Coefficient Linear Difference Equations

An *n*-th order (linear constant-coefficient) difference equation is given by

$$\sum_{i=0}^{n} a_i y(k+i) = \sum_{i=0}^{m} b_i u(k+i)$$
 (2.10)

or

- 2.2 Signal Flow Graphs
- 2.3 Nyquist Analysis and Design
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Chapter 3

Image Processing