CSE 483: Mobile Robotics

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Non-holonomic Trajectory Planning (Bernstein Basis method)

This document discusses the theory of non holonomic trajectory planning using Bernestein polynomial along with the brief description of the associated topics.

1 What is a Nonholonomic motion

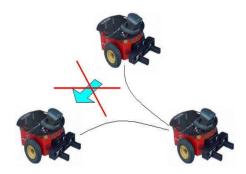


Figure 1: Non holonomic characteristic of wheeled robots.

Non-holonomic systems are characterized by constraint equations which involves the time derivatives of the system configuration variables. In a configuration space $Q \subset R^n$, the configuration of a mechanical system can be uniquely described by an n-dimensional vector of generalized coordinates.

$$q = (q_1, q_2, q_3, \dots, q_n)^T$$

The generalized velocity at a generic point of a trajectory $q(t) \subset Q$ is the tangent vector given by

$$\dot{q} = (\dot{q}_1, \dot{q}_2, \dot{q}_3, \dots, \dot{q}_n)^T$$

For a non holonomic systems, these equations are non integrable as they typically arise when the system has less controls than configuration variables (underactuated systems). As a result, a nonholonomic mechanical system cannot move in arbitrary directions in its configuration space. For instance, a unicycle has two controls (linear(v) and angular(w) velocities), while it moves in a 3-dimensional configuration space(x, y, θ). As a consequence, any path in the configuration space does not necessarily correspond to a feasible path for the system. In other words, for a non-holonomic systems, the line integrals depend not just on the start and end points but also the path taken.

The state transition matrix representation of a holonomic system is of the form

$$\dot{x} = f(x, u) \tag{1}$$

Since, equation 1 is non-integrable, we can approximate the integration using numerical integration methods, say Euler's method, which gives

$$x_{new} \approx x + f(x, u)\Delta t$$
,

This shows that the new state x_{new} is constrained due to the choice of f.

Figure 1 shows one of the feasible paths (represented with lines) of a non-holonomic mobile robot to move between 2 states.

2 Differential Drive Robots

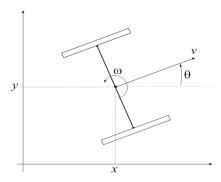


Figure 2: Differential drive wheeled robot

Consider a differential drive nonholonomic mobile robot in a two-dimensional, free-space environment, as shown in figure 2. It is assumed that the robot cannot slip in lateral direction.

generalized coordinates : $q = (x, y, \theta)^T$

Nonholonomic constraints : $\dot{x}\sin\theta - \dot{y}\cos\theta = 0 \implies \dot{y} = \dot{x}\tan\theta$

$$y = \int \dot{x} \tan \theta dt \tag{2}$$

With 2 control inputs as (v, w), the kinematics model of the system is given by:

$$\dot{q} = \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & 0 \\ \sin \theta & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ w \end{bmatrix} \tag{3}$$

3 Motion Planning with Bernstein Polynomials

Since, equation 2 is non integrable, we can approximate the functions \dot{x} and $\tan \theta$ with the bernestein polynomials and solve the integral.

NOTE: The Bernstein polynomials are advantageous over other approximation techniques like taylor series as the former and its derivatives polynomials uniformly approximates f and \dot{f} , respectively. It holds true for higher order derivatives as well. Moreover, they are the most numerically stable basis.

3.1 Bernstein basis polynomial

Let f(x) be a real-valued function defined and bounded on the interval [0,1], then $B_n(f)$ is the polynomial on [0,1].

$$\mathbf{B_n}(\mathbf{f}(\mathbf{x});\mathbf{t}) = \sum_{i=0}^{n} \mathbf{f}(\frac{\mathbf{i}}{\mathbf{n}})^{\mathbf{n}} \mathbf{C_i} \mathbf{t^i} (1-\mathbf{t})^{\mathbf{n}-\mathbf{i}}$$
(4)

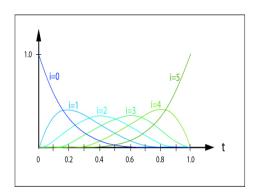


Figure 3: Bernstein basis polynomials for n=5.

If function f(x) is continuous on [0,1], then the Bernstein polynomial $B_n(f(x))$ tends uniformly to f as $n \to \infty$. Bernstein basis polynomials with n=5, are shown in figure 3.

For the given initial state $(x_{t_0}, y_{t_0}, \theta_{t_0})$ and final state $(x_{t_f}, y_{t_f}, \theta_{t_f})$, (more state constraints can be added to the system), with the start time t_0 and end time t_f of the trajectory of a nonholonomic system, y(t) and x(t) can be related as, $y(t) = x(t) \tan \theta(t)$,

where, the functions x(t) and $\tan \theta(t)$ can be approximated as a linear combination of Bernstein basis polynomials as following-

$$x(t) \approx B_n(x(t)) = B_x(\mu(t)) = \sum_{i=0}^{5} W_{x_i} B_i(\mu(t)),$$
 (5)

Similarly,

$$\tan \theta(t) = k(t) \approx B_n(k(t)) = B_k(\mu(t)) = \sum_{i=0}^{5} W_{k_i} B_i(\mu(t))$$
 (6)

where,

$$B_i(\mu(t)) = {n \choose t} C_i (1-\mu)^i (\mu)^{n-i}$$

$$\mu(t) = \frac{t-t_0}{t_f-t_0}$$

Differentiating equation 5 w.r.t time gives

$$\dot{x}(t) = \dot{B}_x(\mu(t)) = \sum_{i=0}^{5} W_{x_i} \dot{B}_i(\mu(t))$$
(7)

using above 2 equations, equation 2 can be rewritten as-

$$y(t) = y_0 + \int_{t_0}^{t} \left(\sum_{i=0}^{5} W_{x_i} \dot{B}_i(\mu(t))\right) \left(\sum_{i=0}^{5} W_{k_i} B_i(\mu(t)) dt\right)$$
(8)

The bernstein coefficients of the polynomials and their derivatives for n=5, at time $t = t_0$ and $t = t_f$ are shown in the tables below.

Bernstein coefficients	$t = t_0$	$t = t_f$
$B_0(\mu) = {}^5 C_0(1-\mu)^5 \mu^0$	1	0
$B_1(\mu) = {}^5 C_1(1-\mu)^4\mu$	0	0
$B_2(\mu) = {}^5 C_2(1-\mu)^3 \mu^2$	0	0
$B_3(\mu) = {}^5 C_3(1-\mu)^2 \mu^3$	0	0
$B_4(\mu) = {}^5 C_4(1-\mu)^1 \mu^4$	0	0
$B_5(\mu) = {}^5 C_5(1-\mu)^0 \mu^5$	0	1

Bernstein coefficients derivatives	$t = t_0, \mu = 0$	$t = t_f, \mu = 1$
$\dot{B}_0(\mu) = {}^5 C_0 \frac{-5(1-\mu)^4}{(t_f - t_0)}$	$\frac{-5}{(t_f - t_0)}$	0
$\dot{B}_1(\mu) = {}^5 C_1 \frac{-4\mu(1-\mu)^3 + (1-\mu)^4}{(t_f - t_0)}$	$\frac{5}{(t_f - t_0)}$	0
$\dot{B}_2(\mu) = {}^{5} C_2 \frac{{}^{-3\mu^2(1-\mu)^2 + 2(1-\mu)^3\mu}}{(t_f - t_0)}$	0	0
$\dot{B}_3(\mu) = {}^5 C_3 \frac{-2\mu^3 (1-\mu) + 3(1-\mu)^2 \mu^2}{(t_f - t_0)}$	0	0
$\dot{B}_4(\mu) = {}^5 C_4 \frac{-\mu^4 + 4(1-\mu)\mu^3}{(t_f - t_0)}$	0	$\frac{-5}{(t_f - t_0)}$
$\dot{B}_5(\mu) = {}^5 C_5 \frac{5\mu^4}{(t_f - t_0)}$	0	$\frac{5}{(t_f-t_0)}$

With the given state constraints (i.e. position, velocity, acceleration, etc.) of the robot at different instants, the unknown, time independent weight parameters $(W_{x_0}, W_{x_1}, W_{x_2},, W_{x_5})$ and $(W_{k_0}, W_{k_1}, W_{k_2},, W_{k_5})$ can be determined.

3.2 Finding $W_{x_0}, W_{x_1}, ..., W_{x_5}$

Given Constraints: $(x_{t_0}, y_{t_0}), (x_{t_c}, y_{t_c}), (x_{t_f}, y_{t_f}), (x_{t_0}, y_{t_0}), (x_{t_c}, y_{t_c}), (x_{t_f}, y_{t_f})$ Using equations 5, known parameters can be represented as a linear combination of bernstein basis polynomial as follows-

$$x(t_0) = W_{x_0}B_0(\mu(t_0)) + W_{x_1}B_1(\mu(t_0)) + W_{x_2}B_2(\mu(t_0)) + W_{x_3}B_3(\mu(t_0)) + W_{x_4}B_4(\mu(t_0)) + W_{x_5}B_5(\mu(t_0))$$

$$(9)$$

$$x(t_f) = W_{x_0} B_0(\mu(t_f)) + W_{x_1} B_1(\mu(t_f)) + W_{x_2} B_2(\mu(t_f)) + W_{x_3} B_3(\mu(t_f)) + W_{x_4} B_4(\mu(t_f)) + W_{x_5} B_5(\mu(t_f))$$
(10)

Putting values of bernstein polynomial coefficients (from tables) in the equations 9 and 10, gives

$$W_{x_0} = x(t_0) = x_{t_0} (11)$$

$$W_{x_5} = x(t_f) = x_{t_f} (12)$$

Using the remaining constraints, all weights $W_{x_1}, W_{x_2}, W_{x_3}, W_{x_4}$ can be evaluated.

$$\begin{bmatrix} x_{tc} - W_{x_0} B_0(\mu(t_c)) - W_{x_5} B_5(\mu(t_c)) \\ \dot{x}_{t_0} - W_{x_0} \dot{B}_0(\mu(t_0)) - W_{x_5} \dot{B}_5(\mu(t_0)) \\ \dot{x}_{t_f} - W_{x_0} \dot{B}_0(\mu(t_f)) - W_{x_5} \dot{B}_5(\mu(t_f)) \\ \dot{x}_{t_c} - W_{x_0} \dot{B}_0(\mu(t_c)) - W_{x_5} \dot{B}_5(\mu(t_c)) \end{bmatrix} = \begin{bmatrix} B_1(\mu(t_c)) & B_2(\mu(t_c)) & B_3(\mu(t_c)) & B_4(\mu(t_c)) \\ \dot{B}_1(\mu(t_0)) & \dot{B}_2(\mu(t_0)) & \dot{B}_3(\mu(t_0)) & \dot{B}_4(\mu(t_0)) \\ \dot{B}_1(\mu(t_f)) & \dot{B}_2(\mu(t_f)) & \dot{B}_3(\mu(t_f)) & \dot{B}_4(\mu(t_f)) \\ \dot{B}_1(\mu(t_c)) & \dot{B}_2(\mu(t_c)) & \dot{B}_3(\mu(t_c)) & \dot{B}_4(\mu(t_c)) \end{bmatrix} \begin{bmatrix} W_{x_1} \\ W_{x_2} \\ W_{x_3} \\ W_{x_4} \end{bmatrix}$$

$$(13)$$

$$A_x = B_x W_x \tag{14}$$

$$W_x = pinv(B_x)A_x \tag{15}$$

3.3 Finding $W_{k_0}, W_{k_1},, W_{k_5}$

Expanding equation 8,

$$y(t) = y_0 + \int_{t_0}^{t} (W_{k_0} \cdot f_0(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + (W_{k_1} \cdot f_1(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x_1}, W_{x_2}, ...W_{x_5}) + ... + (W_{k_5} \cdot f_5(t, t_0, t_f, W_{x$$

with the weight parameters $W_{x_1}...W_{x_5}$ calculated above, equation 16 further reduces to,

$$y(t) = y_0 + W_{k_0} F_0(t) + W_{k_1} F_1(t) + W_{k_2} F_2(t) + W_{k_3} F_3(t) + W_{k_4} F_4(t) + W_{k_5} F_5(t)$$
(17)

$$where F_i(t) = \int_{t_0}^t f_i(t, t_0, t_f, W_{x_1}, W_{x_2}, \dots W_{x_5}) dt$$
(18)

Our objective is to get weights, $W_{K_0}, W_{K_1}, W_{K_2}, W_{K_3}, W_{K_4}, W_{K_4}$

$$k(t_0) = k_0 = W_{k_0} B_0(\mu(t_0)) + W_{k_1} B_1(\mu(t_0)) + W_{k_2} B_2(\mu(t_0)) + W_{k_3} B_3(\mu(t_0)) + W_{k_4} B_4(\mu(t_0)) + W_{k_5} B_5(\mu(t_0))$$

$$(19)$$

which gives,

$$W_{k_0} = k_0 \tag{20}$$

$$k(t_f) = k_f = W_{k_0} B_0(\mu(t_f)) + W_{k_1} B_1(\mu(t_f)) + W_{k_2} B_2(\mu(t_f)) + W_{k_3} B_3(\mu(t_f)) + W_{k_4} B_4(\mu(t_f)) + W_{k_5} B_5(\mu(t_f))$$
 (21)

which gives,

$$W_{k_5} = k_f \tag{22}$$

Using any 4 out of remaining constraints on y and k we can form a full rank matrix for B_k as:

$$\begin{bmatrix} \dot{k}_{t_0} - W_{k_0} \dot{B}_0(\mu(t_0)) - W_{k_5} \dot{B}_5(\mu(t_0)) \\ \dot{k}_{t_f} - W_{k_0} \dot{B}_0(\mu(t_f)) - W_{k_5} \dot{B}_5(\mu(t_f)) \\ y_0 - W_{k_0} F_0 - W_{k_5} F_5 \\ y_f - W_{k_5} F_0 - W_{k_5} F_5 \end{bmatrix} = \begin{bmatrix} \dot{B}_{k_1}(\mu(t_0)) & \dot{B}_2(\mu(t_0)) & \dot{B}_3(\mu(t_0)) & \dot{B}_4(\mu(t_0)) \\ \dot{B}_1(\mu(t_f)) & \dot{B}_2(\mu(t_f)) & \dot{B}_3(\mu(t_f)) & \dot{B}_4(\mu(t_f)) \\ F_1(t_0) & F_2(t_0) & F_3(t_0) & F_4(t_0) \\ F_1(t_f) & F_2(t_f) & F_3(t_f) & F_4(t_f) \end{bmatrix} \begin{bmatrix} W_{k_1} \\ W_{k_2} \\ W_{k_3} \\ W_{k_4} \end{bmatrix}$$

$$(23)$$

$$A_k = B_k W_k \tag{24}$$

$$W_k = pinv(B_k)A_k \tag{25}$$

Depending on the number of given constraints i.e., rank of the matrices B_x $(p \times q)$ and B_k $(m \times n)$, the system can be categorized as under constrained, critically constrained and over constrained.

1. Critically constrained:

p=q, rank
$$(B_x)$$
= $p \implies pinv(B_x) = B_x^{-1}$
m=n, rank (B_k) = $m \implies pinv(B_k) = B_k^{-1}$

2. Under constrained:

$$p < q$$
, rank $(B_x) = p \implies pinv(B_x) = (B_x^T B_x)^{-1} B_x^T$
 $m < n$, rank $(B_k) = m \implies pinv(B_k) = (B_k^T B_k)^{-1} B_k^T$

3. Over constrained:

$$p > q$$
, rank $(B_x) = q \implies pinv(B_x) = B_x^T (B_x B_x^T)^{-1}$
 $m > n$, rank $(B_k) = n \implies pinv(B_k) = B_k^T (B_k B_k^T)^{-1}$

Once, the weight parameters W_k and W_x are determined, the trajectory in 2D plane can be estimated as following,

$$x(t) = W_{x_0}B_0(\mu(t)) + W_{x_1}B_1(\mu(t)) + W_{x_2}B_2(\mu(t)) + W_{x_3}B_3(\mu(t)) + W_{x_4}B_4(\mu(t)) + W_{x_5}B_5(\mu(t))$$
(26)

$$y(t) = y_0 + W_{k_0}F_0(t) + W_{k_1}F_1(t) + W_{k_2}F_2(t) + W_{k_3}F_3(t) + W_{k_4}F_4(t) + W_{k_5}F_5(t)$$
 (27)

and the orientation of the robot can be defined by-

$$\theta(t) = \arctan((W_{k_0}B_0(\mu(t) + W_{k_1}B_1(\mu(t)) + W_{k_2}B_2(\mu(t)) + W_{k_3}B_3(\mu(t)) + W_{k_4}B_4(\mu(t)) + W_{k_5}B_5(\mu(t)))$$
(28)