

Mobile Robotics

Robot Motion:

Path Smoothing and PID Control

Programming Assignments and Projects

PA5A - Generating Smooth Path

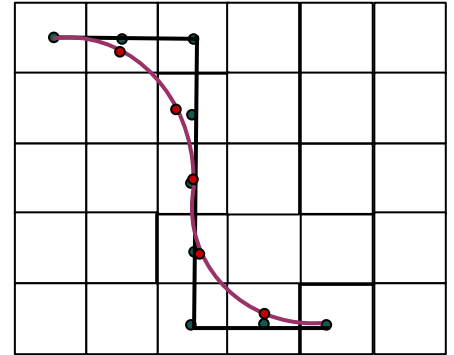
- Blue path is generated by a path planning algorithm
- Red curve is smoothing and easy to follow by a robot
- Smoothing algorithm

- Old path: $x_0, x_1, \dots, x_i, \dots, x_{n-1}$
- New path: $y_0, y_1, \dots, y_i, \dots, y_{n-1}$
- Initially, $y_i = x_i$ for $i = 0, 1, \dots, n - 1$
- Want to minimize

$$\alpha(x_i - y_i)^2 + \beta[(y_{i-1} - y_i)^2 + (y_{i+1} - y_i)^2]$$

- Use the following gradient descent equations to update y_i iteratively for $i = 1, \dots, n - 2$ until the change of values of y_i 's less than a threshold value

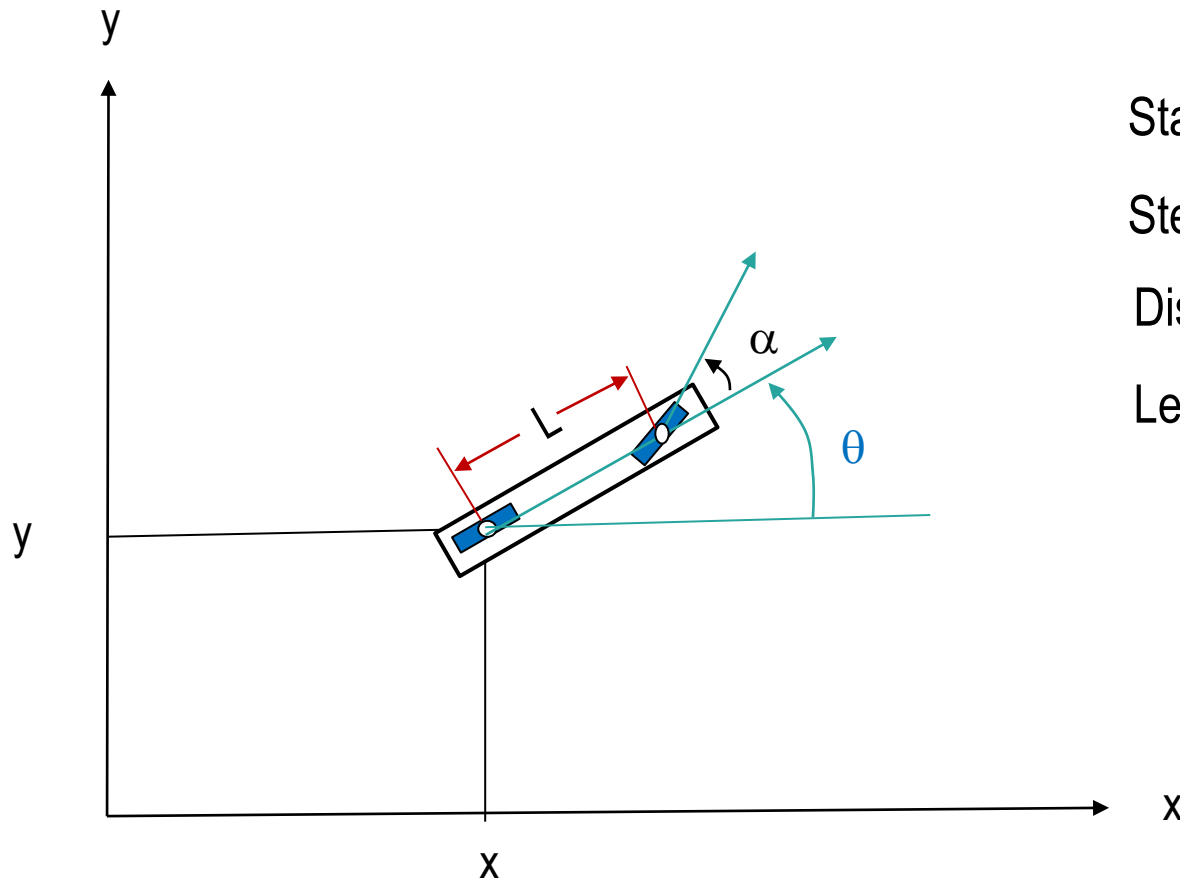
$$y_i = y_i + \alpha(x_i - y_i) + \beta(y_{i-1} + y_{i+1} - 2y_i) \quad \alpha = 0.5 \quad \beta = 0.1$$



- What will happen? When $\alpha = 0$ or $\beta = 0$

Recall: Bicycle Model

- The bicycle robot shown below will be used in this project



State: (x, y, θ)

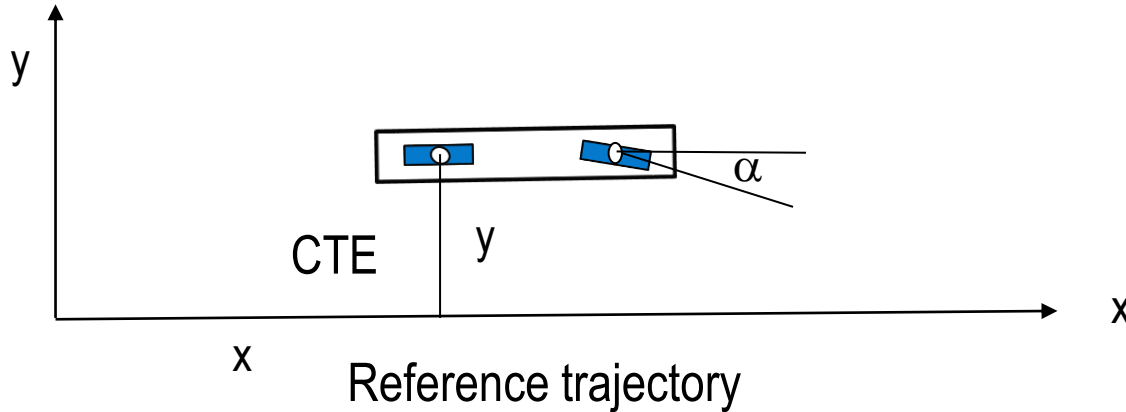
Steering angle : α

Distance: d

Length: L

PA5B : P (Proportional) Control

- Reference trajectory and Cross track error (CTE)

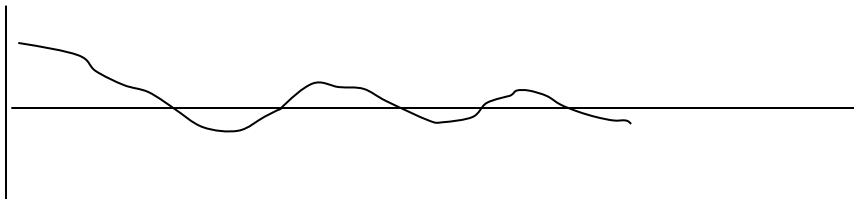


State: (x, y, θ)

Steering angle : α

Distance: d

- To steer in proportional to CTE: $\alpha = -\tau * CTE$
- Problem: Overshoot and oscillation



Marginally stable

- What will happen? When
 - τ is bigger

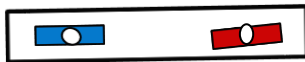
PA5C : PD (Proportional and Differential) Control

- To steer in proportional to CTE and difference of CTE:

$$\alpha = -\tau_P * CTE - \tau_D * \frac{d}{dt} CTE$$

$$\frac{d}{dt} CTE = \frac{CTE_t - CTE_{t-1}}{\Delta t} \quad \Delta t=1$$

- Problem
 - Systematic Bias causes big CTE



- P-term and D-term cannot solve this problem

PA5D : PID Control

- PID means Proportional, Integral, and Differential
- To steer in proportional to CTE and integral and difference of CTE:

$$\alpha = -\tau_P * CTE - \tau_D * \frac{d}{dt} CTE - \tau_I * \int CTE$$

$$\frac{d}{dt} CTE = \frac{CTE_t - CTE_{t-1}}{\Delta t} \quad \Delta t=1$$

$$\int CTE = \sum CTE_t * \Delta t \quad \Delta t=1$$

- Question: how can we find good control gains τ_P, τ_D, τ_I ?
- The answer is called Twiddle or Coordinate ascent

Twiddle or Coordinate Ascent Algorithm

- **"Twiddle"** refers to an optimization algorithm that is used to fine-tune the parameters of a system or a model. It is a heuristic algorithm that is often used in control theory and machine learning. The basic idea of twiddle is to iteratively adjust the parameters of the system or model until a satisfactory result is achieved. At each iteration, the algorithm tweaks the parameters and measures the effect on the system's output. If the output is improved, the algorithm keeps the new parameter values; otherwise, it reverts to the old parameter values and tries a different set of adjustments.
- **"Coordinate ascent algorithm"** is a type of optimization algorithm that is often used in machine learning, particularly in the context of reinforcement learning. The basic idea of coordinate ascent is to optimize a function by successively maximizing it with respect to each of its variables, while holding the other variables fixed. The algorithm iteratively updates the value of one variable, then fixes it and updates the value of another variable, and so on, until convergence is achieved.
- **Note: the above two points are stolen from ChatGPT.**

Twiddle Algorithm for PID Controller

- Initialize the PID controller with some initial parameter values
- Set the twiddle factor and calculate initial error as the best error so far
- Loop until the sum of the twiddle values is less than the tolerance
 - Try each parameter in turn
 - Add the twiddle factor to the parameter value and calculate the new error.
 - If the new error is less than the best error so far, update the best error and increase the twiddle factor for that parameter by 10%.
 - else subtract twice the twiddle factor from the parameter value, and calculate the new error
 - If the new error is less than the best error so far, update the best error and increase the twiddle factor for that parameter by 10%.
 - else reset the parameter value to its original value and decrease the twiddle factor for that parameter by 10%.

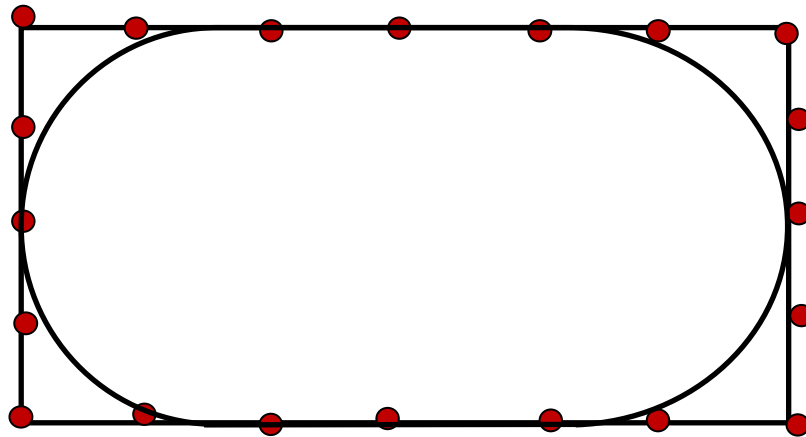
PA5E : Implement Twiddle Function

- The run(p) function in PA5D has three parameters.
- Twiddle algorithm can generate a set of optimized parameters p so that the function run(p) will minimize the average CTE.

```
p = [0, 0, 0]
dp = [1, 1, 1]
best_err = run(p)
while sum(dp) > tol:
    for i in range(len(p)):
        p[i] += dp[i]
        err = run(p)
        if err < best_err:
            best_err = err
            dp[i] *= 1.1
    else:
        p[i] -= 2*dp[i]
        err = run(p)
        if err < best_err:
            best_err = err
            dp[i] *= 1.1
        else:
            p[i] += dp[i]
            dp[i] *= 0.9
return p
```

PP5A: Cyclic Path Smoothing

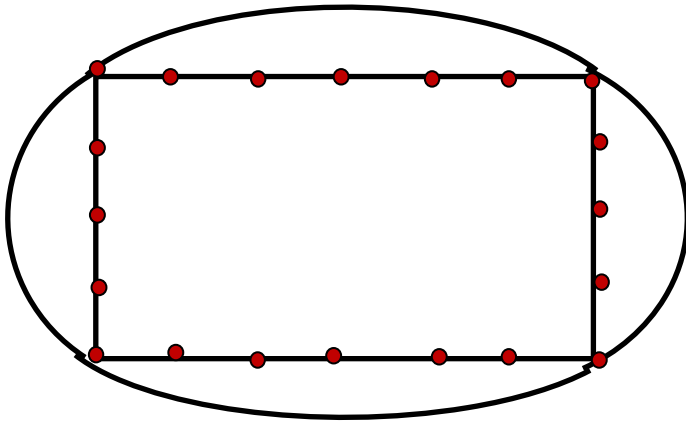
- Very similar to PA5A.
 - Use gradient descent equations
 - Not fix the end points
 - Example: Unsmoothed path is a rectangular
 - the smoothed path is inside unsmoothed path



```
path=[[0, 0],  
      [1, 0],  
      [2, 0],  
      [3, 0],  
      [4, 0],  
      [5, 0],  
      [6, 0],  
      [6, 1],  
      [6, 2],  
      [6, 3],  
      [5, 3],  
      [4, 3],  
      [3, 3],  
      [2, 3],  
      [1, 3],  
      [0, 3],  
      [0, 2],  
      [0, 1]]
```

PP5B: Constrained Smoothing (1)

- Given a path with fixed points.
 - During smoothing the given path, fixed points cannot be moved.
 - Example: Unsmoothed path is a rectangular and its 4 corners are fixed.
 - the smoothed path is outside unsmoothed path



```
path=[[0, 0], #fix
      [1, 0],
      [2, 0],
      [3, 0],
      [4, 0],
      [5, 0],
      [6, 0], #fix
      [6, 1],
      [6, 2],
      [6, 3], #fix
      [5, 3],
      [4, 3],
      [3, 3],
      [2, 3],
      [1, 3],
      [0, 3], #fix
      [0, 2],
      [0, 1]]
```

```
fix = [1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0]
```

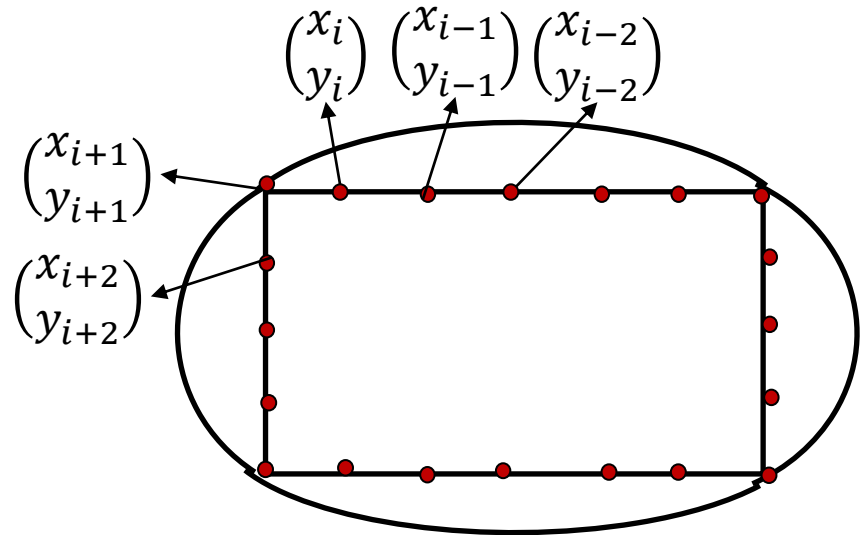
PP5B: Constrained Smoothing (2)

- Given a path with fixed points.
 - During smoothing the given path, fixed points cannot be moved.
 - Also want to minimize

$$\gamma[(A - B)^2 + (C - D)^2]$$

– here

$$A = \begin{pmatrix} x_{i-1} \\ y_{i-1} \end{pmatrix} - \begin{pmatrix} x_i \\ y_i \end{pmatrix}$$
$$B = \begin{pmatrix} x_{i-2} \\ y_{i-2} \end{pmatrix} - \begin{pmatrix} x_{i-1} \\ y_{i-1} \end{pmatrix}$$
$$C = \begin{pmatrix} x_{i+1} \\ y_{i+1} \end{pmatrix} - \begin{pmatrix} x_i \\ y_i \end{pmatrix}$$
$$D = \begin{pmatrix} x_{i+2} \\ y_{i+2} \end{pmatrix} - \begin{pmatrix} x_{i+1} \\ y_{i+1} \end{pmatrix}$$



- So, we have the following additional gradient descent equation

$$\begin{pmatrix} x_i \\ y_i \end{pmatrix} = \begin{pmatrix} x_i \\ y_i \end{pmatrix} + \gamma \left[2 \begin{pmatrix} x_{i-1} \\ y_{i-1} \end{pmatrix} - \begin{pmatrix} x_{i-2} \\ y_{i-2} \end{pmatrix} - \begin{pmatrix} x_i \\ y_i \end{pmatrix} \right] + \gamma \left[2 \begin{pmatrix} x_{i+1} \\ y_{i+1} \end{pmatrix} - \begin{pmatrix} x_{i+2} \\ y_{i+2} \end{pmatrix} - \begin{pmatrix} x_i \\ y_i \end{pmatrix} \right]$$

$$\gamma = 0.5 * \text{weight_smooth}$$

PP5C: Racetrack Control

- Drive a robot along the following racetrack
- The robot starts at $(0, r)$ and faces north ($\pi/2$)
- A **run** function is provided to drive the robot
- You need to define the **cte** function which is called inside the **run** function to determine the steering angle.
- Four cases should be considered when compute **cte**
 - $x \leq r$
 - $r < x \leq 3r$ and $y < r$
 - $r < x \leq 3r$ and $y > r$
 - $r > 3r$

