Robotic Motion Planning: curve tracing

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Move to Goal

- Distance d(a,b) = $((a_x b_x)^2 + (a_y b_y)^2)^{\frac{1}{2}}$
- Gradient descent of d(a,b), i.e., decrease distance to the goal



⊖ goal

Circumnavigating Obstacles: Curve Tracing



Predict: Tangent

Correct: Something else

Normal (and hence Tangent) to Obstacle



Circumnavigate Obstacles: Boundary Following



 $D(x) = \min d(x,c)$

Normal is parallel to $\nabla D(x)$

Increase/Decrease/Same

Safety distance W*

Tangent is orthogonal to both

$$\dot{c}(t) = v \quad v \text{ is in } (n(c(t))^{\perp})$$

Raw Distance Function



$$\rho(x,\theta) = \min_{\lambda \in [0,\infty]} d(x, x + \lambda [\cos \theta, \sin \theta]^T),$$

such that $x + \lambda [\cos \theta, \sin \theta]^T \in [$

Saturated raw distance function

$$\rho_R(x,\theta) = \begin{cases} \rho(x,\theta), & \text{if } \rho(x,\theta) < R\\ \infty, & \text{otherwise.} \end{cases}$$

Implicit Function Theorem

 $G(x) = D(x) - W^*$

Roots of G(x) trace the offset curve

DG(x) = DD(x), which is like a gradient in Euclidean spaces

Null of DG(x) is tangent, hence perp of DD(x) is too

THEOREM D.1.1 (Implicit Function Theorem) Let $f : \mathbb{R}^m \times \mathbb{R}^n \to \mathbb{R}^n$ be a smooth vector-valued function, f(x, y). Assume that $D_y f(x_0, y_0)$ is invertible for some $x_0 \in \mathbb{R}^m$, $y_0 \in \mathbb{R}^n$. Then there exist neighborhoods X_0 of x_0 and Z_0 of $f(x_0, y_0)$ and a unique, smooth map $g : X_0 \times Z_0 \to \mathbb{R}^n$ such that

f(x,g(x,z))=z

for all $x \in X_0$, $z \in Z_0$.

Correction

THEOREM D.2.1 (Newton-Raphson Convergence Theorem) Let $f : \mathbb{R}^n \to \mathbb{R}^n$ and $f(y^*) = 0$. For some $\rho > 0$, let f satisfy

• $Df(y^*)$ is nonsingular with bounded inverse, i.e., $||(Df(y^*))^{-1}|| \leq \beta$

$$||Df(x) - Df(y)|| \le \gamma ||x - y|| \text{ for all } x, y \in B_{\rho}(y^*), \text{ where } \gamma \le \frac{2}{\rho \beta}$$

Now consider the sequence $\{y^h\}$ defined by

$$y^{h+1} = y^h - (Df(y^h))^{-1}f(y^h),$$

for any $y^0 \in B_{\rho}(y^*)$. Then $y^h \in B_{\rho}(y^*)$ for all h > 0, and the sequence $\{y^h\}$ quadratically converges onto y^* , i.e.,

 $||y^{h+1} - y^*|| \le a ||y^h - y^*||^2$ where $a = \frac{\beta \gamma}{2(1 - \rho \beta \gamma)} < \frac{1}{\rho}$.