

第六章作业分享





## 纲要

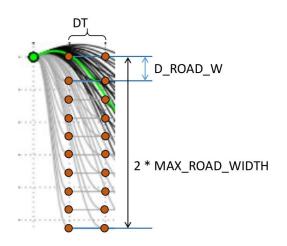


- Lateral sampling
- Longitudinal sampling
- ➤ Trajectory cost function
- ▶Path selection

## Lateral sampling



- 首先对横向位移(采样距离应限制在车道内)进行采样;
- 2. 时间进行采样,从而得到末端横向距离固定的不同曲线;
- 3. 首末端共**6个固定量**,用**五次多项式**拟合成一条用于描述横向采样性质的曲线。



```
for (float di = 0; di <= 5; di += D_ROAD_W) {
    //different curves of end point d
    for (float ti = MINT; ti <= MAXT; ti += DT) {
        // Different driving time under the same d
        FrenetPath fp_without_s;

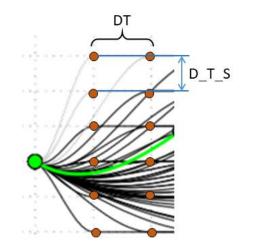
        QuinticPolynomial lateral_fp(c_d, c_d_d, c_d_dd, di, 0.0, 0.0, ti);

        for (float _t = 0.0; _t <= ti; _t += DT) {
            fp_without_s.t.push_back(_t);
            fp_without_s.d.push_back(lateral_fp.calc_point(_t));
            fp_without_s.d_d.push_back(lateral_fp.calc_first_derivative(_t));
            fp_without_s.d_ddd.push_back(lateral_fp.calc_second_derivative(_t));
            fp_without_s.d_dddd.push_back(lateral_fp.calc_third_derivative(_t));
        }
}</pre>
```

## Longitudinal sampling



- 1. 纵向对车速的采样,纵坐标为速度,横坐标为时间(此处的时间t值为d-t曲线中的离散时间值,保证s-d——对应);
- 2. 纵向速度确定然纵向位移不定,首末端共**5个固定量**,用**四次多项式**拟合成一条用于描述 横向采样性质的曲线



```
// 当前遍历下,每一条s-t曲线复用上面刚刚生成的d-t曲线(只有一根),for循环每运行一次,生成一条完整的备选轨迹
for (float vi = TARGET_SPEED - D_T_S * N_S_SAMPLE; vi < TARGET_SPEED + D_T_S * N_S_SAMPLE; vi += D_T_S) {

FrenetPath fp_with_s = fp_without_s;
QuarticPolynomial longitudinal_qp(s0, c_speed, 0.0, vi, 0.0, ti);

for (float _t : fp_without_s.t) {
    fp_with_s.s.push_back(longitudinal_qp.calc_point(_t));
    fp_with_s.s_d.push_back(longitudinal_qp.calc_first_derivative(_t));
    fp_with_s.s_dd.push_back(longitudinal_qp.calc_second_derivative(_t));
    fp_with_s.s_ddd.push_back(longitudinal_qp.calc_third_derivative(_t));
}
```

## Trajectory cost function



 参考论文 Optimal Trajectory Generation for Dynamic Street Scenarios in a Frenet Frame 给出的评价函数,算出每条采样轨迹的代价;

# // The lateral cost of each alternate trajectory, fp\_with\_s.cd = KJ \* Jp + KT \* ti + KD \* pow(fp\_with\_s.d.back(), 2); // The longitudinal cost of each alternate track, float ds = pow(TARGET\_SPEED - fp\_with\_s.s\_d.back(), 2); fp\_with\_s.cv = KJ \* Js + KT \* ti + KD \* ds; // The weighted sum of the above costs -- the total cost fp\_with\_s.cf = KLAT \* fp\_with\_s.cd + KLON \* fp\_with\_s.cv; fp\_list.push\_back(fp\_with\_s);

#### ◆横向代价

横向加速度; 时间; 末尾处横向位移

$$C_d = k_j J_p + k_t t + k_d d^2$$

#### ◆纵向代价

纵向加速度; 时间; 末尾处速度偏差

$$C_{v} = k_{j}J_{s}\left(s\left(t\right)\right) + k_{t}t + k_{\dot{s}}\left[\dot{s}_{1} - \dot{s}_{2}\right]^{2}$$

### Path selection



- 碰撞检测
- 速度阈值检测
- 加速度阈值检测
- 曲线行驶曲率检测
- Jerk检测
- • • •

```
Vec_Path FrenetOptimalTrajectory::check_paths(Vec_Path path_list, const Vec_Poi ob) {
    Vec Path output fp list;
    for (FrenetPath fpp : path list)
        bool collision free = false;
        collision free = check collision( fpp, ob);
        if(collision free){
            if(fabs(_fpp.max_speed) < MAX_SPEED){</pre>
                if(fabs( fpp.max accel) < MAX_ACCEL){</pre>
                    if(fabs(_fpp.max_curvature) < MAX_CURVATURE){</pre>
                         output_fp_list.push_back(_fpp);
    return output fp list;
```

## Path selection



• 可行路径中,根据轨迹的代价确定最优路径

```
FrenetPath FrenetOptimalTrajectory::frenet_optimal_planning(Spline2D csp, float c_speed, float c_d, float c_d_d, float c_d_d, Vec_Poi ob) {
    Vec_Path fp_list = calc_frenet_paths(c_speed,c_d,c_d_d,c_d_dd,s0);
    calc_global_paths(fp_list, csp);
    Vec_Path save_paths = check_paths(fp_list, ob);
    float min_cost = std::numeric_limits<float>::max();
    FrenetPath final_path;
    for (auto path:save_paths){
        if(min_cost >= path.cf){
            min_cost = path.cf;
            final_path = path;
        }
    }
    return final_path;
}
```

## 在线问答







## 感谢各位聆听 / Thanks for Listening

