# Model Predictive Control (MPC) Project

#### The Model

The state contains the x position, y position, vehicle orientation, velocity, cross track error or **cte**, which express the error between the center of the road and the vehicle's position, and the orientation error or **epsi**.

It also has two actuators: the steering wheel, and the throttle and brake pedals considered as a single actuator (negative values signifies braking, and positive values signifies accelerating). Therefore, we have two control inputs, the steering angle  $\delta$  and the acceleration  $\mathbf{a}$ .

The model we have chosen in the one explained in the course, with the following update ecuations:

```
x_{t+1} = x[t] + v[t] * cos(psi[t]) * dt

y_{t+1} = y[t] + v[t] * sin(psi[t]) * dt

psi_{t+1} = psi[t] + v[t] / Lf * delta[t] * dt

v_{t+1} = v[t] + a[t] * dt

cte[t+1] = f(x[t]) - y[t] + v[t] * sin(epsi[t]) * dt

epsi[t+1] = psi[t] - psides[t] + v[t] * delta[t] / Lf * dt
```

Please notice that **Lf** measures the distance between the front of the vehicle and its center of gravity.

Our constrains are  $[\delta, a]$ , with steering angle  $\delta \in [-25,25]$  degrees (or  $\delta \in [-0.436332, 0.436332]$  in radians) and acceleration  $a \in [-1,1]$ .

In my model, I am trying to minimize the predicted distance of the vehicle from the trajectory (cte). Also, I try to minimize the predicted difference between the vehicle orientation and trajectory orientation (epsi).

## <u>Timestep Length and Elapsed Duration (N & dt)</u>

The duration over which future predictions are made is the prediction horizon or **T**.

 ${\bf N}$  is the number of timesteps in the horizon.  ${\it dt}$  is how much time elapses between actuations.

```
T = N \times dt.
```

In my final model, I have chosen N=10 and dt=0.1, which gives a prediction horizon equals 1 second. I expect that in 1 second, the environment does not change too much.

I have tried other combination of values, such as  $\{N=10, dt=0.5\}$  and  $\{N=20, dt=0.1\}$ , but the one I have noticed better results is the one I mentioned at the beginning  $\{N=10, dt=0.1\}$ .

## **Polynomial Fitting and MPC Preprocessing**

I calculate waypoints by using the following formula:

```
waypoints_x[i] = cos_psi * (ptsx[i] - px) + sin_psi * (ptsy[i] - py);
waypoints_y[i] = -sin_psi * (ptsx[i] - px) + cos_psi * (ptsy[i] - py);
```

I use a helper (**polyfit**) to fit a third order polynomial to waypoints, and another helper function (**polyeval**) to calculate the cross track error (**cte**) and orientation error (**epsi**).

The initial state is set with  $\mathbf{x}$ ,  $\mathbf{y}$  and  $\mathbf{psi}$  equals to 0. The initial state values for  $\mathbf{cte}$  and  $\mathbf{epsi}$  are the ones obtained the former paragraph. The initial state value for  $\mathbf{v}$  is calculated as described in the next section (Model Predictive Control with Latency).

```
// set the initial state, with x, y and psi equals 0
VectorXd initial_state(6);
initial_state << 0, 0, 0, v, cte, epsi;</pre>
```

Steering angle and throttle are calculated invoking mpc.Solve() method as follows:

```
// obtain throttle and steer via mpc
MPC_Results res = mpc.Solve(initial_state, coeffs);
double steer_value = -res.steering;
double throttle_value = res.throttle;
```

## **Model Predictive Control with Latency**

In real cars, an actuation command doesn't execute instantly. There is a delay as the command propagates through the system. This is known as "latency".

To deal with latency, we implement the following equations:

```
// Deal with latency
px = px + v * cos_psi * latency;
py = py + v * sin_psi * latency;
psi = psi - v * steering / 2.67 * latency;
v = v + throttle * latency;
```

where 2.67 is the Lf. That value was obtained by measuring the radius formed by running the vehicle in the simulator around in a circle with a constant steering angle and velocity on a flat terrain.