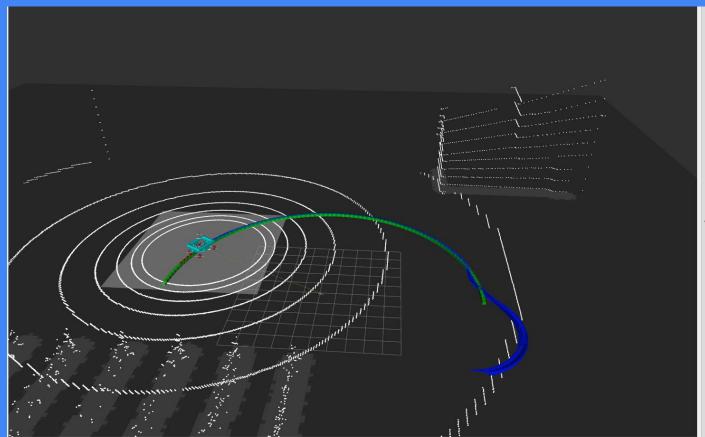
## Trajectory Control of Ackermann Steering Robot With Nonlinear MPC



#### Deriving the kinematics

#### of simplified bicycle model

```
m = \text{mass of the vehicle}
```

 $l_f, l_r = \text{Distance from CoG to Front/Rear Axels}$ 

 $I_z$  = Moment of Inertia of vehicle about z axis

x = Longitudinal position [global frame]

y = Lateral position [global frame]

 $v_x = \text{Longitudinal velocity [body frame]}$ 

 $v_y = \text{Lateral velocity [body frame]}$ 

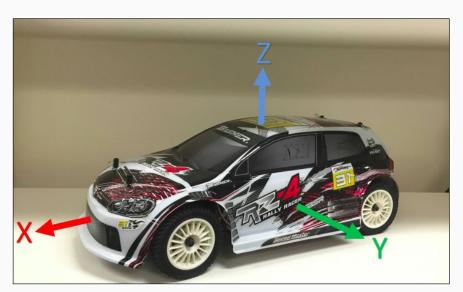
 $\dot{\psi} = r = \text{Yaw rate}$ 

 $\delta =$  Front steering angle

 $F_{xR} = \text{Input rear force}$ 

 $F_{xF}, F_{yF} = \text{Longitudinal/Lateral force on front tire}$ 

#### Robot's coordinate frames



# Deriving the kinematics

#### of simplified bicycle model

State dynamics then become;

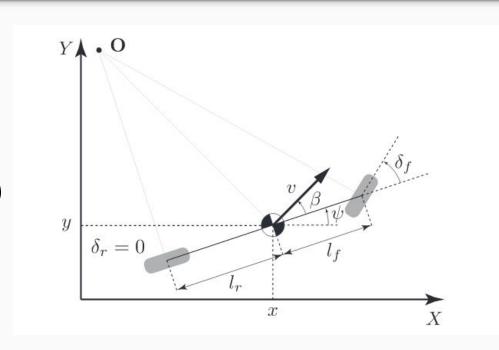
$$\dot{x} = v \cos(\psi + \beta)$$

$$\dot{y} = v \sin(\psi + \beta)$$

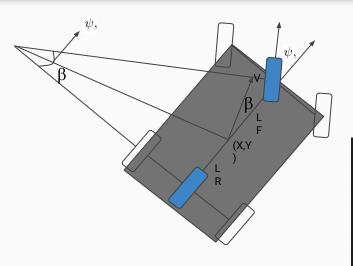
$$\dot{\psi} = \frac{v \cos(\beta)}{l_f + l_r} \tan(\delta_f) = \frac{v}{l_r} \sin(\beta)$$

$$\dot{v} = a$$

$$\beta = \tan^{-1}(\frac{l_r}{l_f + l_r} \tan(\delta_f))$$



#### Discretized state dynamics in C++



```
auto beta = casadi::MX::atan(L_R / (L_F + L_R) * casadi::MX::tan(df_dv_(i)));

opti_->subject_to(
    x_dv_(i + 1) == x_dv_(i) + DT * (v_dv_(i) * casadi::MX::cos(psi_dv_(i) + beta)));

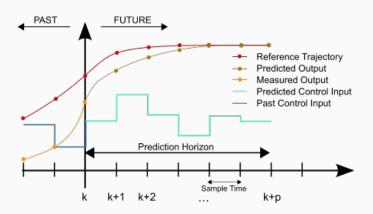
opti_->subject_to(
    y_dv_(i + 1) == y_dv_(i) + DT * (v_dv_(i) * casadi::MX::sin(psi_dv_(i) + beta)));

opti_->subject_to(
    psi_dv_(i + 1) == psi_dv_(i) + DT * (v_dv_(i) / L_R * casadi::MX::sin(beta)));

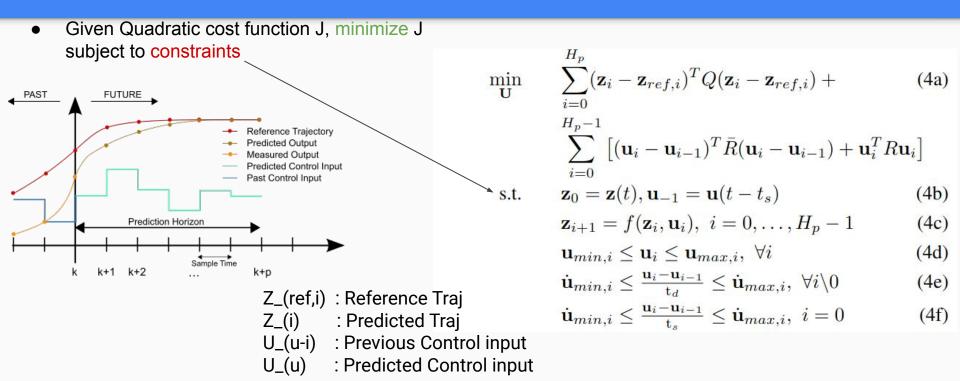
opti_->subject_to(
    v_dv_(i + 1) == v_dv_(i) + DT * acc_dv_(i));
```

# Model Predictive Control(MPC)

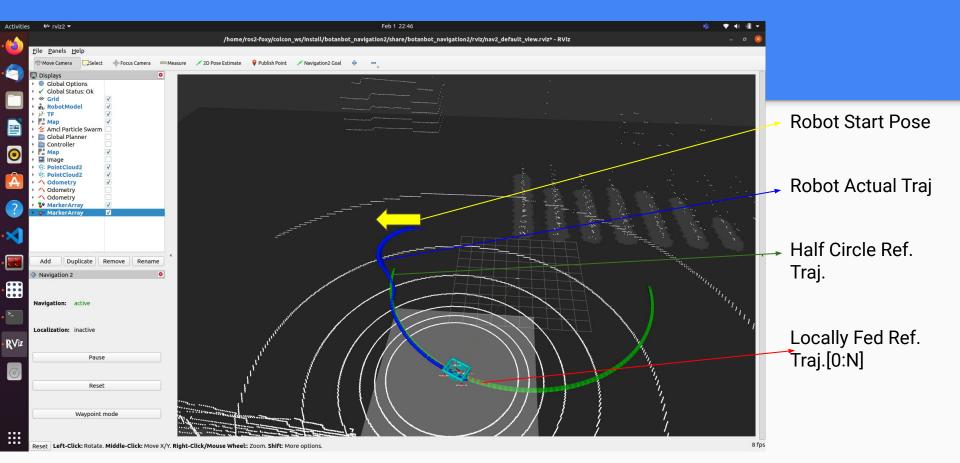
- Advanced method to control a process while satisfying some constraints
- Rely on dynamic models of the process
- MPC is based on iterative, finite-horizon optimization of a plant model



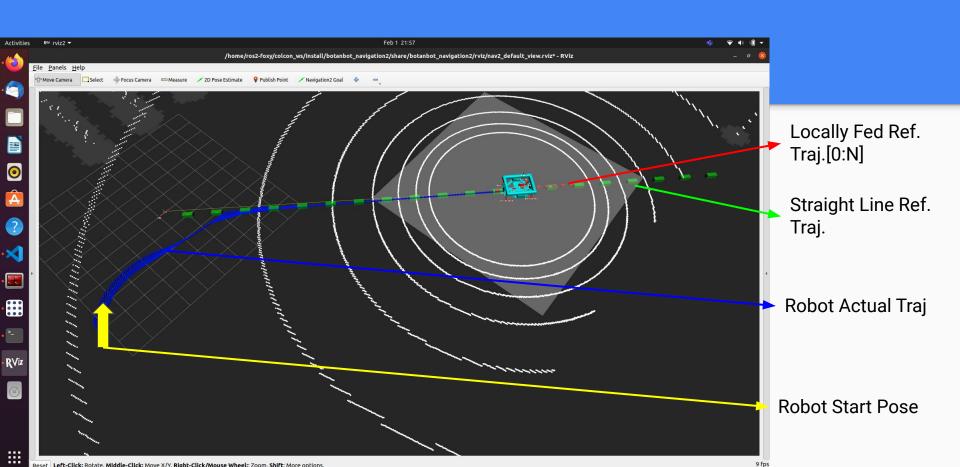
# Model Predictive Control(MPC)



## Results



## Results



# Video

```
* @brief all parameters usd by MPC class,
* user needs to create and reconfigure this
*/
struct Parameters
 // timesteps in MPC Horizon
 int N;
 // discretization time between timesteps(s)
 double DT:
 // distance from CoG to front axle(m)
 double L F;
 // distance from CoG to rear axle(m)
 double L R;
 // min / max velocity constraint(m / s)
 double V_MIN;
  double V MAX;
 // min / max acceleration constraint(m / s ^ 2)
  double A_MIN;
 double A_MAX;
 // min / max front steer angle constraint(rad)
 double DF MIN;
  double DF MAX;
 // min / max jerk constraint(m / s ^ 3)
 double A DOT MIN;
  double A DOT MAX;
 // min / max front steer angle rate constraint(rad / s)
  double DF_DOT_MIN;
 double DF_DOT_MAX;
 // weights on x, y, psi, and v.
 std::vector<double> 0;
 // weights on jerk and slew rate(steering angle derivative)
 std::vector<double> R;
 // enable/disable debug messages
 bool debug mode;
 // set this true only if user figured the configration
 bool params_configured;
```

# Next

std\_msgs/Empty result #feedback float32 distance\_to\_goal float32 speed Add Collisions as constraints

#goal definition

nav\_msgs/Path path string controller\_id

#result definition

- Expose MPC controller as an FollowPath: action
- Tune MPC Controllers Parameters

References: Kinematic and dynamic vehicle models for autonomous driving control

design Engineering, Computer Science

2015 IEEE Intelligent Vehicles Symposium (IV)