

Last Time:

- Walking
- MPC control for quadrupeds

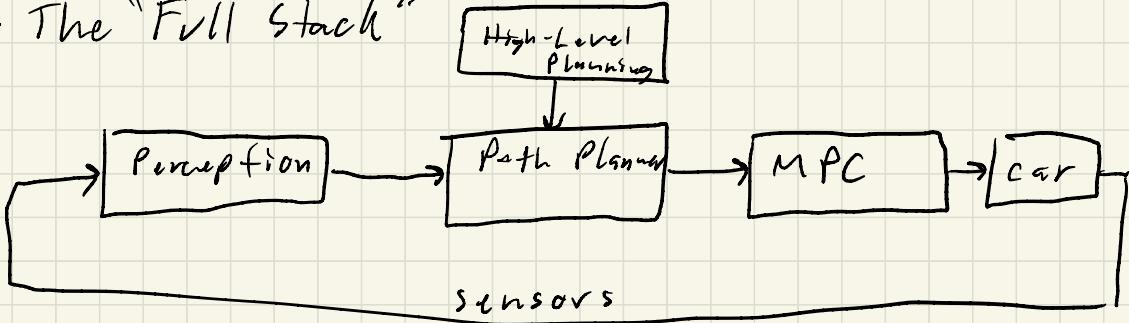
Today:

- How to drive a car
- Game-theoretic MPC

» History

- Primitive demos going back to 1920s-30s
- First serious modern work in 1980s at CMU and Mercedes-Benz
- Lots of demos in 1990s with 75%+ autonomy on long highway trips
- DARPA Grand Challenge in early 2000s greatly accelerated progress
- Most current efforts trace back to ~2010.

* The "Full Stack"



- Sensors: GPS, IMU, Cameras, LIDAR, RADAR
- Perception is the hard part

- High-level Planner: Route planning with graph search (generates waypoints)
- Path Planner: Generate smooth spline curve that is collision free (no dynamics)
- MPC Controller: Tracks spline curve while reasoning about vehicle dynamics + constraints.

* Vehicle Dynamics:

- Lots of options with different levels of fidelity
- Most common is the "bicycle" or "single track" model
- Kinematic Bicycle Model:



- Assume inputs v, α (or $\dot{\alpha}$)
- Assume tires don't slip

$$\begin{aligned}\dot{x} &= v \cos(\theta) \\ \dot{y} &= v \sin(\theta) \\ \dot{\theta} &= \frac{v}{l} \tan(\alpha) \\ \ddot{\alpha} &= u_2\end{aligned}$$

$$x = \begin{bmatrix} x \\ y \\ \theta \end{bmatrix}$$

$$U = \begin{bmatrix} v \\ \alpha \\ u_2 \end{bmatrix}$$

- Works well for "normal" driving
- Breaks down in more aggressive settings (high acceleration).
- More Complex Models:
 - o "Dynamic" Bicycle Model: Model car as a rigid body. $F = ma$, $\tau = J\dot{\omega}$. Reason engine torque and tire forces explicitly. Can handle more dynamic behaviors.
 - o "Double-Track" Model: Model 2 tires, full 3D rigid body dynamics, suspension, etc. Reason about body roll + weight transfer in aggressive cornering. Needed for e.g. off-road, racing, drifting.
- Tire models can go from simple (no-slip) to medium complexity (Coulumb), to very complicated (contact patch, deformation, nonlinear stiffness, etc.)

* State-of-the-Art Nonlinear MPC

- Dynamic bicycle model with good tire models gets you really far.
- Online MPC with IPOPT at ~ 80 Hz

* The Frozen Robot Problem:

- We want the MPC controller to reason about coupled behavior with other drivers. "If I do this, the other car will probably do that."
- Current systems assume other cars will continue at constant velocity over MPC horizon.
- Works well for highway driving
- Breaks down in scenarios with tighter coupling between cars. e.g. ramp merging
- If there's not a big enough gap between cars, MPC controller will just stop.

* Game-Theoretic Trajectory Optimization

- High-level idea: Assume other cars are also solving an optimal control problem like me.
- Solve a joint trajectory optimization problem for all cars simultaneously
- One version of this idea: "Nash equilibrium"

$$\bar{X} = \begin{bmatrix} x^1 \\ x^2 \\ \vdots \\ x^n \end{bmatrix}, \quad \bar{U} = \begin{bmatrix} u^1 \\ u^2 \\ \vdots \\ u^n \end{bmatrix}$$

← Strategies + inputs for all cars searched

$$\min_{\bar{X}, \bar{u}^i} J^i(\bar{X}, \bar{u}^i) \leftarrow \text{cost for car } i$$

$$\text{s.t. } \begin{aligned} D(\bar{X}, \bar{u}) &= 0 \\ C(\bar{X}, \bar{u}) &\leq 0 \end{aligned} \quad \left. \begin{array}{l} \text{dynamics} \\ \text{collision constraints} \end{array} \right\} \text{for all cars}$$

- We get n (number of cars) of these problems that we must solve simultaneously.
- Interpretation: No player can unilaterally improve their cost
- Good model of non-cooperative driving behavior
- Cost functions can capture driver aggressiveness etc.

* Solution Strategy (ALGAMES)

- Form Augmented Lagrangian for each player's 1st-order necessary conditions:

$$L^i(\bar{X}, \bar{u}, \lambda^i, \mu^i) = J^i(\bar{X}, \bar{u}^i) + \frac{\rho}{2} \|D(\bar{X}, \bar{u})\|_2^2 + \lambda^{i\top} D(\bar{X}, \bar{u}) + \frac{\rho}{2} \|C(\bar{X}, \bar{u})^i\|_2^2 + \mu^{i\top} C(\bar{X}, \bar{u})$$

$$\Rightarrow \nabla_{u^i} L^i = 0$$

- Stack FON conditions for all players:

$$\begin{bmatrix} \nabla_{\mu} L' \\ \nabla_{\mu} L^2 \\ \vdots \\ \nabla_{\mu} L^n \end{bmatrix} = 0$$

- Solve with Newton's Method
- Apply standard AL update to estimate λ, μ
- Generates human-like interaction strategies
- By taking advantage of sparsity in Newton solver, can solve at 30ms/Hz for 4~5 cars