

• Topic and Motivation

- Minimum-time flight for VTOL-Drone
- Push platform to its limits → to explore full potential

• Use cases:

- Rescue
- Delivery
- Transportation

Image

• Our Approach

- 2D Electra VTOLs,
 - more dim in observation space
- 3D Quadcopters
 - other action space + double the hidden units

difference to primary source paper

- We produce as output PWM (pulse duration modulation) for each propeller individually, → low level controllers

- Paper produce collective Thrust + Body rate commands
 - low-level PID-controller then produces speed command for each rotor.

• What has been done so far

- Problem reduced to 2D
 - Stabilize slow/fast
 - Back to 3D case
- **lurge:** Waypoints connected with lines

• Training Strategy

Reward Scaling

pretraining slow model

Reward Scaling

fast model

lot of our time spent on Modeling of Reward

Reward

- Fly along trajectory → reward progress in each time step.
- Reward for reaching waypoint within certain distance

Penalties

- high body rates
- falling

differences in phases

- slow phase: encourage velocity range and flying close to trajectory

fast phase:

- encourage faster velocities
- + scale down progress reward such that reaching points is rewarded more in comparison → result more stable fast flight

- + penalty for body rate scaled down
→ allowing more rapid maneuvers

• Plots 2D

Success rate / velocity per step for slow + fast mode

- Success rate: how much drones reach every waypoint \rightarrow most 100% after 400 steps
- Success full drones: completion time goes down and converges
 \rightarrow drone goes faster

• Videos 2D

Old: Overshooting was a problem

Now: Drone knows how to corrigate it.

Plots 3D

- Reward: goes up in both cases
 - ? Slow learning: learned to takeoff at $\sim 10^4$ steps.

\Rightarrow Success rate goes suddenly up.

\Rightarrow Fast learning: velocity at 4 meter per seconds.

• Isn't time optimal yet

\Rightarrow pushing this further requires slowly increasing encouraged Velocity and more training. \rightarrow This takes some time!

More recent:

Office Test

Retraining with horizontal trajectories.

PID : 12.45

Ours: 3.55 → $7 \frac{m}{s}$ (Success rate 100%)

⇒ Further training
actually will increase
velocity.

Possible further Research

- Curriculum Learning: to slowly increase velocities in fast training
- Training with D Rand: Drone gets different weights during training so it's robust to model parameters
⇒ for sim to real
- Using phy. Model with latency: latency between sensor reading and action ~2ms

SAC / DDPG

Off policy \rightarrow more sample efficient

But we couldn't achieve convergence yet.