

Mathematical Modeling of Drone Kinematics:

From Single-Drone Point-to-Point Motion to Scaling Multi-Drone Formations

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1 Overview

This document develops the kinematic model of drone motion in two stages:

1. **Part I** — a single drone commanded to move from an initial position to a goal position (the point-to-point, or *go-to-goal*, problem);
2. **Part II** — n drones holding a geometric formation defined by a centroid, orientation, and *scale*, with the formation Jacobian and inverse kinematics that let the group translate, rotate, and expand or contract as a single virtual body.

Throughout, kinematics means we model *motion* (positions, velocities) and treat velocity as the control input; forces, thrust, and aerodynamics are abstracted into a lower-level controller.

2 Part I: Single Drone Moving to One Position

2.1 Kinematic model

Let the drone's position in the world frame be $\mathbf{p}(t) \in \mathbb{R}^3$. The simplest kinematic model is the *single integrator*:

$$\dot{\mathbf{p}}(t) = \mathbf{u}(t), \quad \mathbf{u}(t) \in \mathbb{R}^3, \quad (1)$$

where $\mathbf{u}$ is the commanded velocity. This abstraction is justified for multirotors because the onboard autopilot tracks velocity commands quickly relative to the timescale of trajectory-level motion.

A more faithful model retains the attitude. With rotation matrix $R \in SO(3)$ (body to world) and body angular velocity $\boldsymbol{\omega}_b$:

$$\dot{\mathbf{p}} = \mathbf{v}, \quad \dot{R} = R \hat{\boldsymbol{\omega}}_b, \quad (2)$$

where $\hat{\cdot}$ maps a vector to its skew-symmetric matrix, $\hat{\boldsymbol{\omega}} \mathbf{a} = \boldsymbol{\omega} \times \mathbf{a}$. A quadrotor is *underactuated*: it produces thrust only along its body z -axis, so lateral acceleration requires tilting. In this document the formation layer outputs velocity references $\mathbf{u}$, and (2) is handled by the inner attitude loop.

2.2 Go-to-goal control law

Let $\mathbf{p}^* \in \mathbb{R}^3$ be the desired position. Define the position error

$$\mathbf{e}(t) = \mathbf{p}^* - \mathbf{p}(t). \quad (3)$$

The proportional go-to-goal law is

$$\mathbf{u}(t) = k \mathbf{e}(t) = k(\mathbf{p}^* - \mathbf{p}(t)), \quad k > 0. \quad (4)$$

2.3 Closed-form solution and convergence

Substituting (4) into (1) gives the linear ODE

$$\dot{\mathbf{p}} = k(\mathbf{p}^* - \mathbf{p}) \implies \mathbf{p}(t) = \mathbf{p}^* + (\mathbf{p}(0) - \mathbf{p}^*)e^{-kt}. \quad (5)$$

The error decays exponentially, $\|\mathbf{e}(t)\| = \|\mathbf{e}(0)\| e^{-kt}$, so the drone converges to the goal from any initial condition. Formally, the Lyapunov function $V = \frac{1}{2}\|\mathbf{e}\|^2$ satisfies

$$\dot{V} = \mathbf{e}^\top \dot{\mathbf{e}} = -k \|\mathbf{e}\|^2 \leq 0, \quad (6)$$

proving global exponential stability of $\mathbf{p} = \mathbf{p}^*$.

2.4 Velocity saturation

Real drones have a maximum speed $v_{\max}$. A common modification caps the command while preserving its direction:

$$\mathbf{u} = \begin{cases} k \mathbf{e}, & k\|\mathbf{e}\| \leq v_{\max}, \\ v_{\max} \frac{\mathbf{e}}{\|\mathbf{e}\|}, & \text{otherwise.} \end{cases} \quad (7)$$

Far from the goal the drone flies at $v_{\max}$ straight toward $\mathbf{p}^*$; near the goal it slows exponentially, arriving without overshoot.

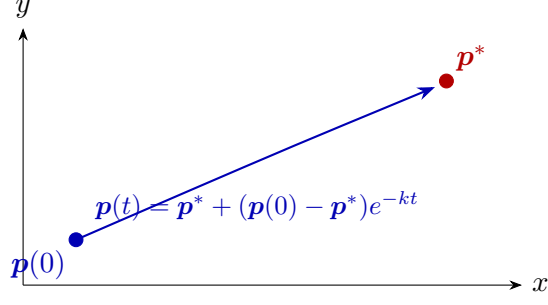


Figure 1: Exponential point-to-point trajectory of a single drone under the proportional law.

3 Part II: Multiple Drones in Formation with Scaling

3.1 Stacked multi-drone kinematics

Consider n drones with positions $\mathbf{p}_i \in \mathbb{R}^3$, each obeying $\dot{\mathbf{p}}_i = \mathbf{u}_i$. Stack them:

$$\mathbf{p} = \begin{bmatrix} \mathbf{p}_1 \\ \vdots \\ \mathbf{p}_n \end{bmatrix} \in \mathbb{R}^{3n}, \quad \dot{\mathbf{p}} = \mathbf{u} \in \mathbb{R}^{3n}. \quad (8)$$

3.2 The formation as a virtual body

Define a *template* of fixed offsets $\mathbf{r}_1, \dots, \mathbf{r}_n \in \mathbb{R}^3$ describing the desired shape (e.g., the vertices of a triangle), chosen so $\sum_i \mathbf{r}_i = \mathbf{0}$ (offsets measured from the shape's centroid). The formation state is

$$\boldsymbol{\xi} = (\mathbf{c}, R, s), \quad \mathbf{c} \in \mathbb{R}^3 \text{ (centroid)}, \quad R \in SO(3) \text{ (orientation)}, \quad s > 0 \text{ (scale)}. \quad (9)$$

3.3 Forward kinematics

Given the formation state, each drone's position is

$$\boxed{\mathbf{p}_i = \mathbf{c} + s R \mathbf{r}_i} \quad (10)$$

This is the forward map $\mathbf{p} = f(\boldsymbol{\xi})$: the analog of joint angles $\rightarrow$ end-effector pose for a robot arm. The scale s acts uniformly: doubling s doubles every drone's distance from the centroid while preserving the shape and orientation.

3.4 Differential kinematics and the formation Jacobian

Differentiate (10) using $\dot{R} = \hat{\boldsymbol{\omega}} R$:

$$\dot{\mathbf{p}}_i = \dot{\mathbf{c}} + \dot{s} R \mathbf{r}_i + s \hat{\boldsymbol{\omega}} R \mathbf{r}_i = \dot{\mathbf{c}} + \dot{s} R \mathbf{r}_i + \boldsymbol{\omega} \times (s R \mathbf{r}_i). \quad (11)$$

Writing $\mathbf{q}_i = R \mathbf{r}_i$, each drone's velocity is linear in the formation velocity $\dot{\boldsymbol{\xi}} = (\dot{\mathbf{c}}, \boldsymbol{\omega}, \dot{s})$:

$$\dot{\mathbf{p}}_i = \underbrace{\begin{bmatrix} I_3 & -s \hat{\mathbf{q}}_i & \mathbf{q}_i \end{bmatrix}}_{J_i(\boldsymbol{\xi}) \in \mathbb{R}^{3 \times 7}} \begin{bmatrix} \dot{\mathbf{c}} \\ \boldsymbol{\omega} \\ \dot{s} \end{bmatrix}, \quad \text{so} \quad \dot{\mathbf{p}} = J(\boldsymbol{\xi}) \dot{\boldsymbol{\xi}}, \quad J = \begin{bmatrix} J_1 \\ \vdots \\ J_n \end{bmatrix} \in \mathbb{R}^{3n \times 7}. \quad (12)$$

J is the *formation Jacobian*. Note the three blocks: translation passes straight through (I_3), rotation contributes $\boldsymbol{\omega} \times$ the drone's offset (the $-s \hat{\mathbf{q}}_i$ term), and scaling pushes each drone radially along its own offset direction ($\mathbf{q}_i$).

3.5 Inverse kinematics of the formation

Position-level. Given a desired formation state $\xi_d = (c_d, R_d, s_d)$, the desired drone positions follow directly from the forward map:

$$\mathbf{p}_i^* = \mathbf{c}_d + s_d R_d \mathbf{r}_i, \quad i = 1, \dots, n. \quad (13)$$

Velocity-level. Given a desired formation trajectory $\xi_d(t)$ with velocity $\dot{\xi}_d = (\dot{\mathbf{c}}_d, \boldsymbol{\omega}_d, \dot{s}_d)$, the feedforward-plus-feedback control for drone i is

$$\mathbf{u}_i = \underbrace{\dot{\mathbf{c}}_d + \dot{s}_d R_d \mathbf{r}_i + \boldsymbol{\omega}_d \times (s_d R_d \mathbf{r}_i)}_{\text{feedforward } J_i \dot{\xi}_d} + \underbrace{k(\mathbf{p}_i^*(t) - \mathbf{p}_i)}_{\text{feedback correction}} \quad (14)$$

Each drone's tracking error $\mathbf{e}_i = \mathbf{p}_i^* - \mathbf{p}_i$ again obeys $\dot{\mathbf{e}}_i = -k\mathbf{e}_i$, so *Part I's convergence result applies drone-by-drone*: the single-drone go-to-goal law is the elementary building block of formation tracking.

Redundancy. The drones have $3n$ degrees of freedom but the formation has only 7. When $3n > 7$ the map is redundant; secondary objectives (collision avoidance, energy) can be inserted in the null space of the Jacobian:

$$\mathbf{u} = J\dot{\xi}_d + (I_{3n} - J^\dagger J) \mathbf{u}_{\text{sec}}, \quad (15)$$

where $J^\dagger = (J^\top J)^{-1} J^\top$ is the Moore–Penrose pseudoinverse. The projected term moves drones without changing the formation-level motion — the same trick used for redundant robot manipulators.

3.6 Pure scaling maneuver

To expand or contract the formation in place, set $\dot{\mathbf{c}}_d = \mathbf{0}$, $\boldsymbol{\omega}_d = \mathbf{0}$, and prescribe a scale trajectory $s_d(t)$, e.g., exponential convergence to a target scale s^* :

$$\dot{s}_d = \lambda(s^* - s_d), \quad \lambda > 0 \implies s_d(t) = s^* + (s_d(0) - s^*)e^{-\lambda t}. \quad (16)$$

Then (14) reduces to a purely radial command

$$\mathbf{u}_i = \dot{s}_d R_d \mathbf{r}_i + k(\mathbf{p}_i^* - \mathbf{p}_i), \quad (17)$$

so every drone moves along its own offset ray through the centroid, and each drone's speed is proportional to $\|\mathbf{r}_i\|$: outer drones move faster, preserving the shape exactly during the expansion.

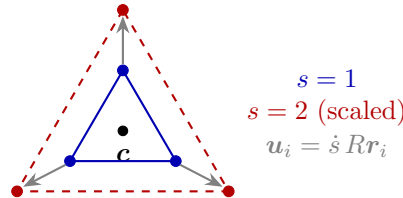


Figure 2: Pure scaling of a triangle formation about a fixed centroid. Each drone moves radially along its offset $R\mathbf{r}_i$; shape and orientation are preserved.

3.7 Worked example: three drones in a scaling triangle

Take $n = 3$ drones in the horizontal plane (z fixed), an equilateral-triangle template of unit circumradius,

$$\mathbf{r}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad \mathbf{r}_2 = \begin{bmatrix} -\frac{1}{2} \\ \frac{\sqrt{3}}{2} \end{bmatrix}, \quad \mathbf{r}_3 = \begin{bmatrix} -\frac{1}{2} \\ -\frac{\sqrt{3}}{2} \end{bmatrix}, \quad \sum_i \mathbf{r}_i = \mathbf{0}, \quad (18)$$

with planar rotation $R(\theta) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$. Command the formation to hold $\mathbf{c}_d = (5, 5)$, $\theta_d = 0$, and expand from $s_d(0) = 1$ to $s^* = 2$ with $\dot{s}_d = \lambda(s^* - s_d)$. The desired positions at any instant are

$$\mathbf{p}_1^* = \begin{bmatrix} 5 + s_d \\ 5 \end{bmatrix}, \quad \mathbf{p}_2^* = \begin{bmatrix} 5 - \frac{s_d}{2} \\ 5 + \frac{\sqrt{3}}{2}s_d \end{bmatrix}, \quad \mathbf{p}_3^* = \begin{bmatrix} 5 - \frac{s_d}{2} \\ 5 - \frac{\sqrt{3}}{2}s_d \end{bmatrix}, \quad (19)$$

and the commands from (14) are $\mathbf{u}_i = \dot{s}_d \mathbf{r}_i + k(\mathbf{p}_i^* - \mathbf{p}_i)$. As $t \rightarrow \infty$, $s_d \rightarrow 2$: the triangle doubles in size about the fixed centroid $(5, 5)$, and the drone-level errors decay as e^{-kt} exactly as in Part I.

3.8 Decentralized formulation (scale-consensus sketch)

The controller (14) assumes each drone knows the global formation state. To decentralize, define the formation on a communication graph G with neighbor sets $\mathcal{N}_i$ and use displacement constraints with a shared scale estimate $\hat{s}_i$ per drone:

$$\mathbf{u}_i = - \sum_{j \in \mathcal{N}_i} [(\mathbf{p}_i - \mathbf{p}_j) - \hat{s}_i R(\mathbf{r}_i - \mathbf{r}_j)], \quad \dot{\hat{s}}_i = -\gamma \sum_{j \in \mathcal{N}_i} (\hat{s}_i - \hat{s}_j) + (\text{leader input}). \quad (20)$$

Each drone uses only relative positions of its neighbors; the scale estimates reach consensus over the connected graph, so a single leader adjusting its $\hat{s}$ makes the entire swarm expand or contract cooperatively.

4 Summary of the Model Hierarchy

Level	Model	Key equation
Single drone	Single integrator, go-to-goal	$\mathbf{u} = k(\mathbf{p}^* - \mathbf{p})$
Formation (forward)	Virtual body: centroid, rotation, scale	$\mathbf{p}_i = \mathbf{c} + sR\mathbf{r}_i$
Formation (differential)	Formation Jacobian	$\dot{\mathbf{p}} = J(\boldsymbol{\xi})\dot{\boldsymbol{\xi}}$
Formation (inverse)	Feedforward + feedback tracking	Eq. (14)
Scaling	Radial motion, shape-preserving	$\mathbf{u}_i = \dot{s} R\mathbf{r}_i + k\mathbf{e}_i$
Decentralized	Graph Laplacian + scale consensus	$\mathbf{u}_i = - \sum_j [(\mathbf{p}_i - \mathbf{p}_j) - \hat{s}_i R(\mathbf{r}_i - \mathbf{r}_j)]$

The single-drone point-to-point law is not merely a warm-up: it is the exact feedback term reused inside the formation controller, with the formation's forward kinematics supplying each drone's moving goal $\mathbf{p}_i^*(t)$ and the Jacobian supplying the feedforward velocity. Scaling enters as one additional degree of freedom of the virtual body, commanded through $\dot{s}$ and realized as shape-preserving radial motion.