

AAE 590: Lie Group Methods for Control and Estimation

Motivation: Why Lie Groups?

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Lie Group Methods for Control and Estimation

What you'll learn:

- How to represent rotations and rigid body poses correctly
- Why traditional methods fail (and when)
- Modern geometric approaches used in aerospace research
- Practical algorithms: IEKF, Equivariant Filter, geometric control

Prerequisites assumed:

- Calculus (derivatives, Taylor series)
- Basic linear algebra (matrices, eigenvalues)
- Programming experience (Python for homework)

The Central Problem

Central Question

How do we represent the orientation of a rigid body?

Seems simple...

- We have 3 degrees of freedom (roll, pitch, yaw)
- Just use 3 numbers, right?

The Catch

- Rotations don't behave like vectors in \mathbb{R}^3
- You can't just add/subtract orientations
- The space of rotations is **curved**

Why Should You Care?

These problems appear everywhere in aerospace:

Spacecraft:

- Attitude determination
- Slew maneuvers
- Pointing control

Aircraft:

- Flight control systems
- Upset recovery
- Aerobatic maneuvers

Robotics:

- SLAM and localization
- Manipulation
- Path planning

Autonomous vehicles:

- Sensor fusion (IMU + GPS)
- State estimation
- Motion planning

A Motivating Failure

Scenario: You're designing an attitude estimator using Euler angles.

Your approach:

- ① Propagate: $\phi_{k+1} = \phi_k + \dot{\phi} \cdot \Delta t$
- ② Update with sensor measurements
- ③ Works great in simulation!

Flight test: Aircraft pitches to 85° ...

- Euler rates explode: $\dot{\phi}, \dot{\psi} \rightarrow \infty$
- Estimator diverges
- Control system fails

What happened?

Gimbal lock! Your 3-parameter representation has a **singularity**.

Let's Start Simple: 2D Rotations

Rotation in 2D: Rotate point (x, y) by angle θ

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

This is a rotation matrix: $R(\theta) \in \text{SO}(2)$

Key properties:

- $R^T R = I$ (orthogonal)
- $\det(R) = \cos^2 \theta + \sin^2 \theta = 1$
- $R(\theta_1)R(\theta_2) = R(\theta_1 + \theta_2)$ (composition = addition!)

2D is Special

In 2D, angles just add. In 3D, life gets complicated!

The 3D Rotation Problem

In 3D, we need to specify:

- Which axis to rotate around
- How much to rotate
- The **order** of multiple rotations

Degrees of freedom: 3 (same as 2D + 1)

But now:

- Rotations don't commute: $R_x R_y \neq R_y R_x$
- Composition is matrix multiplication, not addition
- The space has non-trivial topology

The Challenge

How do we parameterize this 3-DOF space without problems?

Rotation Representations: A Comparison

Representation	Parameters	Singularities	Redundancy
Euler angles	3	Yes (gimbal lock)	No
Axis-angle	3	Yes ($\theta = 0$)	No
Rodrigues vector	3	Yes ($\theta = \pi$)	No
MRP	3	Yes ($\theta = 2\pi$)	No
Quaternions	4	No	Yes ($q = -q$)
Rotation matrix	9	No	Yes (6 constraints)

A Fundamental Trade-off

Minimal parameters \Rightarrow Singularities

No singularities \Rightarrow Redundancy

Euler Angles: Definition

Idea: Compose three rotations about coordinate axes

3-2-1 (Aerospace) convention:

$$R(\phi, \theta, \psi) = R_z(\psi) \cdot R_y(\theta) \cdot R_x(\phi)$$

- ϕ = roll (rotation about x -axis)
- θ = pitch (rotation about y -axis)
- ψ = yaw (rotation about z -axis)

Other conventions exist:

- 3-1-3 (classical Euler angles)
- 1-2-3 (roll-pitch-yaw)
- 12 total possible sequences

Euler Angles: The Rotation Matrices

Elementary rotations:

$$R_x(\phi) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi & \cos \phi \end{bmatrix}$$

$$R_y(\theta) = \begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix}$$

$$R_z(\psi) = \begin{bmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Euler Angle Kinematics

Problem: Given body angular velocity $\omega = [p, q, r]^T$, find Euler rates.

The relationship:

$$\begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix} = \begin{bmatrix} 1 & \sin \phi \tan \theta & \cos \phi \tan \theta \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi \sec \theta & \cos \phi \sec \theta \end{bmatrix} \begin{bmatrix} p \\ q \\ r \end{bmatrix}$$

Notice the problem terms:

- $\tan \theta$ appears in row 1
- $\sec \theta$ appears in row 3

When $\theta = \pm 90^\circ$

$$\tan(\pm 90^\circ) = \pm \infty \text{ and } \sec(\pm 90^\circ) = \pm \infty$$

The kinematic equations become **singular!**

What is Gimbal Lock?

Physical picture: Three nested gimbals

Normal operation:

- Outer gimbal: yaw
- Middle gimbal: pitch
- Inner gimbal: roll
- 3 independent axes

At $\theta = 90^\circ$:

- Outer and inner axes align!
- Yaw and roll do the same thing
- Only 2 independent axes
- Lost 1 degree of freedom

Key Insight

This is a **coordinate singularity**, not a physical limitation.

The vehicle can still rotate in all directions!

Gimbal Lock: Mathematical View

At $\theta = 90^\circ$:

The Euler angle to rotation matrix becomes:

$$R = \begin{bmatrix} 0 & -\sin(\phi - \psi) & \cos(\phi - \psi) \\ 0 & \cos(\phi - \psi) & \sin(\phi - \psi) \\ -1 & 0 & 0 \end{bmatrix}$$

Notice: Only $(\phi - \psi)$ appears, not ϕ and ψ separately!

Consequences:

- Infinitely many Euler angle combinations give the same rotation
- Cannot recover ϕ and ψ individually from R
- Jacobian of $(\phi, \theta, \psi) \mapsto R$ loses rank

Gimbal Lock: A Historical Note

Apollo Program

The Apollo spacecraft IMU used physical gimbals to track orientation.

Engineering reality:

- Gimbal lock was a **known constraint**, not a failure
- Astronauts trained to avoid certain orientations
- Mission procedures kept spacecraft away from singular configurations
- Apollo successfully managed this through operational discipline

The lesson:

- Gimbal lock is a **mathematical** limitation of 3-parameter representations
- Can be managed with careful procedures, but adds operational burden
- Modern systems use quaternions or rotation matrices to avoid the issue entirely

Axis-Angle Representation

Idea: Any rotation is a rotation by angle θ about some axis \hat{a}

Representation: (\hat{a}, θ) where $\|\hat{a}\| = 1$

Or as a vector: $\omega = \theta \hat{a}$ (angle encoded in magnitude)

Rodrigues' rotation formula:

$$R = I + \sin \theta \hat{a}^\wedge + (1 - \cos \theta)(\hat{a}^\wedge)^2$$

where \hat{a}^\wedge is the skew-symmetric matrix of \hat{a} .

Singularity at $\theta = 0$

When $\theta = 0$, the axis \hat{a} is **undefined!**

Any axis gives the identity rotation.

Rodrigues Vector (Classical)

Definition:

$$r = \hat{a} \tan(\theta/2)$$

Conversion to rotation matrix:

$$R = I + \frac{2}{1 + \|r\|^2} (r^\wedge + (r^\wedge)^2)$$

Advantages:

- Only 3 parameters
- Rational functions (no trig in formulas)

Singularity at $\theta = \pi$

$$\tan(\pi/2) = \infty$$

Cannot represent 180° rotations!

Modified Rodrigues Parameters (MRP)

Definition:

$$\sigma = \hat{a} \tan(\theta/4)$$

Advantages over classical Rodrigues:

- Singularity moved from 180° to 360°
- Well-behaved for most practical rotations
- Used in some spacecraft attitude systems

Singularity at $\theta = 2\pi$

$$\|\sigma\| = \tan(\theta/4) \rightarrow \tan(\pi/2) = \infty$$

Still blows up at one full rotation.

Shadow MRP: $\sigma_s = -\sigma/\|\sigma\|^2$ can extend range, but adds complexity.

Understanding “Curved” Spaces: Intuition

Key question: What makes rotation space different from translation space?

Translation (\mathbb{R}^3):

- Move in any direction as far as you want
- You never come back to where you started
- Space is “flat” — like an infinite plane

Rotation ($\text{SO}(3)$):

- Rotate around any axis
- After 360° , you’re back where you started!
- Space “wraps around” — it’s **compact**

The Problem

You can’t flatten a compact space into \mathbb{R}^n without tearing or overlapping.

Key Topology Terms (For Engineers)

Compact:

- A space is **compact** if it is “bounded and closed”
- Intuition: you can’t go to infinity; everything fits in a finite region
- Example: a sphere is compact; a plane is not

Homeomorphic:

- Two spaces are **homeomorphic** if you can stretch/bend one into the other (no tearing or gluing)
- Intuition: same “shape” from a topology perspective
- Example: a coffee cup and donut are homeomorphic (both have one hole)

Diffeomorphic:

- Like homeomorphic, but the stretching must be smooth (differentiable)
- Intuition: same smooth structure, so calculus works the same way
- For this course: “diffeomorphic” \approx “smoothly equivalent”

Topology Example: The Cylinder

Consider: A cylinder $S^1 \times \mathbb{R}$

One axis is “flat”:

- Move up/down forever
- Never return to start
- Like \mathbb{R}

One axis is “curved”:

- Go around the cylinder
- After one loop, back to start
- Like S^1

Key insight: You can parameterize the flat direction with a single coordinate that works everywhere. But the curved direction needs angle wrapping or multiple charts!

Why Minimal Representations Fail: Topology

Theorem (Topological Obstruction)

$\text{SO}(3)$ is diffeomorphic to \mathbb{RP}^3 (real projective 3-space), not \mathbb{R}^3 .
Any 3-parameter global chart must have at least one singularity.

Why?

- $\text{SO}(3)$ is **compact**: all rotations fit in a bounded region
- \mathbb{R}^3 is **not compact**: extends to infinity
- No continuous bijection between them can exist!

This Is Fundamental

Gimbal lock, MRP singularities, etc. are **unavoidable** for any 3-parameter representation. It's topology, not a failure of cleverness!

Quaternions: Definition

Unit quaternion:

$$q = q_0 + q_1\mathbf{i} + q_2\mathbf{j} + q_3\mathbf{k}$$

with $\|q\|^2 = q_0^2 + q_1^2 + q_2^2 + q_3^2 = 1$

Quaternion multiplication rules:

$$\mathbf{i}^2 = \mathbf{j}^2 = \mathbf{k}^2 = \mathbf{ijk} = -1$$

From axis-angle:

$$q = \cos(\theta/2) + \sin(\theta/2)(a_1\mathbf{i} + a_2\mathbf{j} + a_3\mathbf{k})$$

Compact notation: $q = [q_0, \vec{q}]$ where $\vec{q} = [q_1, q_2, q_3]^T$

Quaternion Operations

Quaternion multiplication:

$$p \otimes q = \begin{bmatrix} p_0 q_0 - \vec{p} \cdot \vec{q} \\ p_0 \vec{q} + q_0 \vec{p} + \vec{p} \times \vec{q} \end{bmatrix}$$

Conjugate: $q^* = [q_0, -\vec{q}]$

Inverse: $q^{-1} = q^*/\|q\|^2 = q^*$ for unit quaternions

Rotating a vector:

$$v' = q \otimes v \otimes q^*$$

where $v = [0, \vec{v}]$ is a “pure” quaternion.

Composition: $(q_1 \otimes q_2)$ represents $R_1 R_2$

Quaternions: Advantages

Why quaternions are popular:

- ① **No gimbal lock:** Smooth everywhere
- ② **Compact:** Only 4 parameters (vs 9 for matrices)
- ③ **Efficient:** Quaternion multiply is cheaper than matrix multiply
- ④ **Interpolation:** SLERP gives smooth rotations
- ⑤ **Numerical stability:** Easy to re-normalize

Industry adoption:

- Computer graphics (game engines, animation)
- Aerospace (spacecraft attitude)
- Robotics (motion planning)

The Quaternion Double Cover

The Problem

q and $-q$ represent the **same rotation!**

Mathematically:

$$(-q) \otimes v \otimes (-q)^* = (-1)^2 \cdot q \otimes v \otimes q^* = q \otimes v \otimes q^*$$

Why does this happen?

- Unit quaternions live on S^3 (the 3-sphere in 4D)
- $\text{SO}(3)$ is topologically \mathbb{RP}^3 (projective space)
- The map $S^3 \rightarrow \mathbb{RP}^3$ is 2-to-1

Not a Bug, a Feature (of Topology)

This is exactly how S^3 and $\text{SO}(3)$ are related.

Quaternion Unwinding Problem

Scenario: Attitude control using quaternion feedback

Naive error: $q_e = q_d^* \otimes q$

Problem: Current q is near $-q_d$ (same physical orientation as q_d)

- Error quaternion $q_e \approx [-1, 0, 0, 0]$
- Controller interprets this as “rotate 360°”!
- Vehicle spins unnecessarily (“unwinding”)

Standard workaround:

- ① Check if $q_0 < 0$
- ② If so, use $-q$ instead of q

This Creates Discontinuities

Switching signs introduces jumps in the quaternion trajectory.

Rotation Matrices: Definition

Rotation matrix:

$$R \in \text{SO}(3) = \{R \in \mathbb{R}^{3 \times 3} : R^T R = I, \det(R) = 1\}$$

Properties:

- Columns are orthonormal basis vectors
- $R^{-1} = R^T$ (inverse equals transpose)
- $\det(R) = 1$ (preserves orientation)
- $\|Rv\| = \|v\|$ (preserves lengths)

Action on vectors:

$$v' = Rv$$

Direct matrix-vector multiplication!

Rotation Matrices: Advantages and Disadvantages

Advantages:

- No singularities (globally defined)
- No ambiguity (unique representation)
- Direct action: $v' = Rv$
- Composition: $R_{total} = R_1 R_2$

Disadvantages:

- 9 parameters for 3 DOF
- Must maintain 6 constraints: $R^T R = I$, $\det R = 1$
- Numerical drift: R may become non-orthogonal
- How do we do calculus with matrix constraints?

The Key Question

How do we differentiate, integrate, and optimize over $SO(3)$?

The Lie Group Solution

Instead of fighting the constraints...

Key Insight

Work with the natural mathematical structure!

The Lie group approach:

- ① Recognize $\text{SO}(3)$ as a **Lie group**
- ② Do calculus in the **Lie algebra** $\mathfrak{so}(3) \cong \mathbb{R}^3$
- ③ Map between them via the **exponential map**

Benefits:

- Work directly with rotation matrices (no singularities)
- All calculus happens in a vector space (linear!)
- Constraints are automatically satisfied

Preview: The Lie Algebra

The tangent space at identity: $\mathfrak{so}(3) = T_I \mathrm{SO}(3)$

What is it?

$$\mathfrak{so}(3) = \{\Omega \in \mathbb{R}^{3 \times 3} : \Omega^T = -\Omega\}$$

The space of 3×3 skew-symmetric matrices!

Isomorphism with \mathbb{R}^3 :

$$\omega^\wedge = \begin{bmatrix} 0 & -\omega_3 & \omega_2 \\ \omega_3 & 0 & -\omega_1 \\ -\omega_2 & \omega_1 & 0 \end{bmatrix} \quad \leftrightarrow \quad \omega = \begin{bmatrix} \omega_1 \\ \omega_2 \\ \omega_3 \end{bmatrix}$$

Physical meaning: ω is angular velocity!

Preview: The Exponential Map

From Lie algebra to Lie group:

$$\exp : \mathfrak{so}(3) \rightarrow \text{SO}(3)$$

Definition (matrix exponential):

$$\exp(\omega^\wedge) = I + \omega^\wedge + \frac{(\omega^\wedge)^2}{2!} + \frac{(\omega^\wedge)^3}{3!} + \dots$$

Closed form (Rodrigues):

$$\exp(\omega^\wedge) = I + \frac{\sin \|\omega\|}{\|\omega\|} \omega^\wedge + \frac{1 - \cos \|\omega\|}{\|\omega\|^2} (\omega^\wedge)^2$$

The Bridge

This maps angular velocities to rotation matrices!

Research Example: UAM Safety Verification

Application: Urban Air Mobility (eVTOL aircraft)

The Problem

Verify safety of control systems for vehicles with **coupled rotation and translation**.

Lie Group Approach on SE(2):

- State: pose $X \in \text{SE}(2)$ (position + heading)
- Dynamics: $\dot{X} = X\xi^\wedge$ (left-invariant)
- **Flow pipes:** Over-approximate reachable sets on the manifold
- Provable collision avoidance guarantees

Why Lie Groups?

Traditional Euclidean methods fail to capture the geometric structure.

Lie group formulation enables **global safety certificates**.

Research Example: Inertial Navigation

Application: Strapdown Inertial Navigation Systems (SINS)

The Problem

Integrate high-rate IMU data efficiently while maintaining accuracy.

Lie Group Preintegration on $\text{SE}_n(3)$:

- Extended pose group: rotation, velocity, position in one matrix
- Closed-form preintegration (no iterative updates)
- **45% computational savings** over iterative methods
- Exact propagation of uncertainty

Key Insight

When dynamics are **group affine**, the Lie group structure enables analytical solutions that would be impossible in Euclidean coordinates.

Research Example: Invariant Kalman Filtering

Application: Robot localization and SLAM

Traditional EKF Problems

- Linearization errors depend on state estimate
- Inconsistency: covariance doesn't match actual errors
- Poor performance with large initial uncertainty

Invariant EKF (IEKF):

- Error defined using group structure
- Linearization errors are **state-independent**
- Guaranteed consistency properties
- Works well even with poor initialization

What You'll Learn in This Course

Foundations (Weeks 1-4):

- Groups, manifolds, Lie groups
- $\text{SO}(2)$ as the simplest example
- Exponential/logarithm maps
- Wedge/vee, Ad/ad operators

Core Groups (Weeks 5-12):

- $\text{SE}(2)$: planar motion
- $\text{SO}(3)$: 3D rotations
- $\text{SE}(3)$: 3D rigid body poses

Applications (Weeks 13-17):

- Invariant Extended Kalman Filter
- Equivariant Filter
- Geometric control, dynamics, your projects

Lecture 1 Summary

The Problem:

- Rotations live on a curved 3D space, not \mathbb{R}^3
- Any 3-parameter representation has singularities
- Quaternions avoid singularities but have double cover
- Rotation matrices have no singularities but are redundant

The Solution:

- Work with the natural Lie group structure
- Do calculus in the Lie algebra (a vector space!)
- Use exponential/logarithm maps to go between them

Next: Mathematical foundations — groups and manifolds