

KendraGraph Learning Guide

Space Science + Python for Orbital Risk Intelligence

Author's Note: This guide teaches you exactly what you need to build and understand orbital collision risk systems. No PhD required, no wasted theory. Learn by doing.

Timeline: 8 weeks to solid foundation

Daily Commitment: 1-2 hours

Philosophy: Learn what you need, when you need it

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PART 1: Space & Orbit Foundations

Chapter 1: Orbital Mechanics Basics

The Core Concept

Satellites don't fight gravity - they fall around Earth.

Imagine throwing a ball:

- Throw softly → falls to ground nearby

- Throw harder → lands farther away
- Throw at 7.8 km/s → curves around Earth before landing
- Throw at 7.8 km/s at right angle to gravity → never lands (orbit!)

Why This Matters for Collision Risk

Lower orbit = Faster speed = More satellites = More risk

Upper orbit = Slower speed = Fewer satellites = Less risk

LEO (Low Earth Orbit):

- Altitude: 200-2,000 km
- Speed: 7-8 km/s
- Period: 90-120 minutes
- Crowded! (Starlink, ISS, most satellites)

MEO (Medium Earth Orbit):

- Altitude: 2,000-35,786 km
- Speed: 3-7 km/s
- GPS satellites live here

GEO (Geostationary):

- Altitude: 35,786 km
- Speed: 3.07 km/s
- Period: 24 hours (stays over same spot)
- Communication satellites

The Only Equation You Really Need

Orbital Period: $T = 2\pi \sqrt{a^3/\mu}$

Where:

- T = time for one complete orbit (seconds)
- a = semi-major axis (orbit radius in km)
- $\mu = 398,600 \text{ km}^3/\text{s}^2$ (Earth's gravitational parameter)
- $\pi = 3.14159\dots$

Example: ISS at 400km altitude

$a = \text{Earth radius (6,371)} + \text{altitude (400)} = 6,771 \text{ km}$

$$T = 2\pi \sqrt{(6771^3/398600)} = 5,561 \text{ seconds} = 92.7 \text{ minutes}$$

Practical Python Implementation

python

```

import numpy as np

def orbital_period_minutes(altitude_km):
    """Calculate orbital period given altitude above Earth"""
    earth_radius = 6371 # km
    mu = 398600 # km^3/s^2

    a = earth_radius + altitude_km
    period_seconds = 2 * np.pi * np.sqrt(a**3 / mu)
    period_minutes = period_seconds / 60

    return period_minutes

# Test it
print(f"ISS (400km): {orbital_period_minutes(400):.1f} minutes")
print(f"Starlink (550km): {orbital_period_minutes(550):.1f} minutes")
print(f"GPS (20,200km): {orbital_period_minutes(20200):.1f} minutes")

```

Output:

```

ISS (400km): 92.7 minutes
Starlink (550km): 95.8 minutes
GPS (20,200km): 718.3 minutes (11.97 hours)

```

Key Insights for Your Project

1. Similar altitudes = collision risk zone

- Starlink at 550km passes near other 550km satellites constantly
- Natural separation between LEO and GEO

2. Orbital decay matters

- Atmospheric drag at <600km
- Satellites slowly lose altitude
- Lower altitude = faster speed = phasing changes

3. Don't need perfect physics

- SGP4 library handles complicated orbital mechanics
- You focus on geometry and risk scoring

What to Skip

- ✗ Deriving Kepler's laws

- ~~X~~ Hohmann transfer orbits
 - ~~X~~ Lagrange points
 - ~~X~~ Three-body problem
 - ~~X~~ Orbital perturbation math
-

Chapter 2: Two-Line Elements (TLE)

What is a TLE?

A standardized text format for describing satellite orbits. Every satellite tracked by NORAD has one.

Real Example (ISS):

```
ISS (ZARYA)
1 25544U 98067A 23365.5000000 .00016717 00000-0 10270-3 0 9005
2 25544 51.6400 247.4627 0006703 130.5360 325.0288 15.72125391414289
```

The Fields That Matter

Line 0: Name

```
ISS (ZARYA)
```

Human-readable name (not used by computers)

Line 1: Identification & Epoch

```
1 25544U 98067A 23365.5000000 .00016717 00000-0 10270-3 0 9005
```

 ^ ^
 NORAD ID Epoch (when measured)

- **25544:** Unique satellite ID (NORAD catalog number)
- **23365.5000000:** Year 2023, day 365.5 (Dec 31, noon UTC)

Line 2: Orbital Elements

```
2 25544 51.6400 247.4627 0006703 130.5360 325.0288 15.72125391414289
```

 ^ ^ ^ ^ ^ ^
 Incl RAAN Ecc ArgPer MeanAnom MeanMotion

- **51.6400°:** Inclination (orbit tilt)

- **0006703**: Eccentricity $\times 10^7$ (0.0006703 - nearly circular)
- **15.72125391**: Mean motion (orbits per day)

Python: Parse TLE

Never parse manually. Use libraries:

```
python

from skyfield.api import EarthSatellite, load

# TLE text (get from Space-Track.org or Celestrak)
name = "ISS (ZARYA)"
line1 = "1 25544U 98067A 23365.5000000 .00016717 00000-0 10270-3 0 9005"
line2 = "2 25544 51.6400 247.4627 0006703 130.5360 325.0288 15.72125391414289"

# Create satellite object
satellite = EarthSatellite(line1, line2, name)

# Extract useful info
print(f"Name: {satellite.name}")
print(f"Norad ID: {satellite.model.satnum}")
print(f"Epoch: {satellite.epoch.utc_iso()}")
print(f"Inclination: {satellite.model.inclo * 57.2958:.2f}°")
print(f"Mean Motion: {satellite.model.no_kozai / (2*3.14159) * 1440:.2f} orbits/day")
```

TLE Freshness - CRITICAL

TLEs expire because:

- Atmospheric drag changes orbit
- Satellites maneuver
- Measurement errors accumulate

Check TLE age:

```
python
```

```

from datetime import datetime, timezone

def tle_age_hours(satellite):
    """How old is this TLE?"""
    epoch = satellite.epoch.utc_datetime()
    now = datetime.now(timezone.utc)
    age_hours = (now - epoch).total_seconds() / 3600
    return age_hours

def tle_confidence(age_hours):
    """Confidence in TLE accuracy"""
    if age_hours < 24:
        return "HIGH (< 1 day old)"
    elif age_hours < 72:
        return "MEDIUM (1-3 days old)"
    elif age_hours < 168:
        return "LOW (3-7 days old)"
    else:
        return "VERY LOW (> 1 week old)"

age = tle_age_hours(satellite)
print(f"TLE age: {age:.1f} hours")
print(f"Confidence: {tle_confidence(age)}")

```

Rule of thumb:

- < 24 hours: Good for collision prediction
- 24-72 hours: Acceptable with increased uncertainty
- 72 hours: High uncertainty, refresh TLE
- 1 week: Don't use for collision analysis

Getting TLEs

Free sources:

1. **Space-Track.org** (Best, requires free account)

python

```

import requests

# Login and get session
login_url = "https://www.space-track.org/ajaxauth/login"
query_url = "https://www.space-track.org/basicspacedata/query/class/gp/NORAD_CAT_ID/25544/orderby/EPOCH%20des"

session = requests.Session()
session.post(login_url, data={'identity': 'YOUR_EMAIL', 'password': 'YOUR_PASSWORD'})

response = session.get(query_url)
tle_text = response.text

```

2. CelesTrak (Public, no login, but less frequent updates)

python

```

import requests

url = "https://celestrak.org/NORAD/elements/gp.php?GROUP=stations&FORMAT=tle"
response = requests.get(url)
tles = response.text.split('\n')

```

What to Skip

- ✗ TLE checksum validation (library does it)
- ✗ Manual coordinate conversions
- ✗ Understanding every field in detail
- ✗ Historical TLE formats (pre-2000)

Chapter 3: Orbital Elements Deep Dive

The 6 Classical Orbital Elements

Every orbit is defined by exactly 6 numbers. Think of them as GPS coordinates for space.

1. Semi-Major Axis (a) - Size

What it is: Average distance from Earth's center

Small a = low orbit = fast satellite

Large a = high orbit = slow satellite

Examples:

ISS: $a = 6,771$ km (400 km altitude)

Starlink: $a = 6,921$ km (550 km altitude)

GPS: $a = 26,560$ km (20,189 km altitude)

GEO: $a = 42,164$ km (35,786 km altitude)

Why it matters:

- Groups satellites by altitude
- Similar \boxed{a} = potential collision zone
- Different \boxed{a} = natural separation

Python:

```
python

def altitude_from_period(period_minutes):
    """Calculate altitude given orbital period"""
    mu = 398600 # km^3/s^2
    period_sec = period_minutes * 60
    a = (mu * (period_sec / (2 * np.pi))**2)**(1/3)
    altitude = a - 6371 # subtract Earth radius
    return altitude

# ISS orbits every 92 minutes
print(f"Altitude: {altitude_from_period(92):.0f} km")
```

2. Eccentricity (e) - Shape

What it is: How "stretched" the orbit is

$e = 0$: Perfect circle

$e = 0.01$: Slightly elliptical (most satellites)

$e = 0.5$: Very elliptical (Molniya orbit)

$e = 1.0$: Parabola (escape trajectory)

Most LEO satellites: $e < 0.01$ (treat as circular)

Why it matters:

- Low e (circular) = predictable paths

- High e (elliptical) = varying speed and altitude
- For collision risk: usually ignore unless $e > 0.1$

Practical implication:

```
python

def is_circular(eccentricity, threshold=0.01):
    """Check if orbit is approximately circular"""
    return eccentricity < threshold

# ISS eccentricity = 0.0006703
if is_circular(0.0006703):
    print("Treat as circular orbit for collision analysis")
```

3. Inclination (i) - Tilt

What it is: Angle between orbit plane and equator

$i = 0^\circ$: Equatorial (along equator)
 $i = 28.5^\circ$: Cape Canaveral launches (Florida latitude)
 $i = 51.6^\circ$: ISS (Baikonur, Kazakhstan)
 $i = 90^\circ$: Polar (passes over both poles)
 $i = 98^\circ$: Sun-synchronous (always same local time)
 $i > 90^\circ$: Retrograde (orbits "backwards")

Why it REALLY matters:

Collision risk is highest when inclinations match:

Satellite A: $i = 51.6^\circ$
 Satellite B: $i = 51.5^\circ$
 → They cross paths twice per orbit!

Satellite A: $i = 51.6^\circ$
 Satellite C: $i = 98.0^\circ$
 → Paths cross at steep angles, brief encounters

Python: Inclination similarity check

```
python
```

```

def inclination_risk_group(inclination):
    """Group satellites by inclination for risk analysis"""
    if inclination < 30:
        return "EQUATORIAL"
    elif 45 <= inclination <= 55:
        return "ISS_ZONE" # Very crowded!
    elif inclination > 85:
        return "POLAR"
    else:
        return "MID_INCLINATION"

def collision_risk_from_inclination(i1, i2):
    """Higher risk if inclinations are similar"""
    diff = abs(i1 - i2)

    if diff < 5:
        return "HIGH (similar orbital planes)"
    elif diff < 30:
        return "MEDIUM (crossing orbits)"
    else:
        return "LOW (different planes)"

# Example
print(collision_risk_from_inclination(51.6, 52.1)) # HIGH
print(collision_risk_from_inclination(51.6, 98.0)) # MEDIUM

```

4. RAAN (Ω) - Right Ascension of Ascending Node

What it is: Longitude where orbit crosses equator going northward

Think: "Where does the orbit cross the equator?"

RAAN = 0° : Crosses at prime meridian

RAAN = 90° : Crosses at 90°E longitude

RAAN = 180° : Crosses at 180°E

RAAN = 270° : Crosses at 90°W

Why it matters:

- Changes slowly over time (precession from Earth's bulge)
- Satellites with similar RAAN are in same "longitude slice"
- Sun-synchronous orbits: RAAN precesses $1^\circ/\text{day}$ to track sun

Key insight for collision risk:

Similar inclination + similar RAAN = HIGH RISK

(orbits nearly overlap)

Similar inclination + different RAAN = MEDIUM RISK

(orbits in same plane but shifted)

Python:

```
python
```

```
def raan_difference(raan1, raan2):
    """Calculate angular difference between RAANs"""
    diff = abs(raan1 - raan2)
    # Handle wrap-around (350° and 10° are 20° apart)
    if diff > 180:
        diff = 360 - diff
    return diff

def combined_orbital_similarity(i1, i2, raan1, raan2):
    """Combined risk score from inclination and RAAN"""
    inc_diff = abs(i1 - i2)
    raan_diff = raan_difference(raan1, raan2)

    # Both must be similar for high risk
    if inc_diff < 5 and raan_diff < 30:
        return "VERY HIGH RISK (nearly identical orbits)"
    elif inc_diff < 10 and raan_diff < 60:
        return "HIGH RISK (similar orbital plane)"
    else:
        return "MODERATE/LOW RISK"

# Starlink constellation example
print(combined_orbital_similarity(53.0, 53.2, 100, 110))
```

5. Argument of Perigee (ω) - Orientation of Ellipse

What it is: Where in the orbit the satellite is closest to Earth

For circular orbits ($e \approx 0$): IGNORE THIS

- Most collision analysis ignores ω because orbits are nearly circular

For elliptical orbits:

When it matters:

- Highly elliptical orbits ($e > 0.1$)
- Molniya orbits (Russian communication satellites)
- Transfer orbits

Skip for now unless working with HEO (Highly Elliptical Orbits)

6. Mean Anomaly (M) - Position in Orbit

What it is: Where the satellite is right now in its orbit

$M = 0^\circ$: At perigee (closest to Earth)

$M = 90^\circ$: 1/4 through orbit

$M = 180^\circ$: At apogee (farthest from Earth)

$M = 270^\circ$: 3/4 through orbit

$M = 360^\circ/0^\circ$: Back at perigee

Why it matters:

- Combined with epoch, tells you exact satellite position
- Changes quickly (completes 0-360° each orbit)
- For collision analysis: SGP4 handles this automatically

Practical use:

python

```

# You never manually use mean anomaly
# SGP4 library converts M → true position

from skyfield.api import EarthSatellite, load

ts = load.timescale()
t = ts.now()

# This internally uses mean anomaly + orbital elements
position = satellite.at(t)
x, y, z = position.position.km

print(f"Position: ({x:.0f}, {y:.0f}, {z:.0f}) km")

```

Summary Table

Element	Symbol	What It Controls	Collision Relevance
Semi-Major Axis	a	Orbit size & speed	CRITICAL - groups satellites
Eccentricity	e	Orbit shape	Low (most circular)
Inclination	i	Orbit tilt	CRITICAL - crossing geometry
RAAN	Ω	Equator crossing longitude	HIGH - orbital plane location
Argument of Perigee	ω	Ellipse orientation	Low (for circular orbits)
Mean Anomaly	M	Current position	Handled by SGP4

Practical Risk Grouping

python

```

def orbital_risk_group(a, e, i, raan):
    """Create risk group ID for similar orbits"""
    # Round to bins
    altitude_bin = round(a - 6371, -1) # nearest 10km
    inclination_bin = round(i, 0)      # nearest degree
    raan_bin = round(raan / 30) * 30  # 30° bins

    group_id = f"ALT{altitude_bin}_INC{inclination_bin}_RAAN{raan_bin}"
    return group_id

# Examples
print(orbital_risk_group(6771, 0.001, 51.6, 247))
# Output: ALT400_INC52_RAAN240

print(orbital_risk_group(6921, 0.001, 53.0, 105))
# Output: ALT550_INC53_RAAN120

# Satellites in same group have higher collision risk

```

What to Skip

- ✗ Converting between different anomaly types (mean, true, eccentric)
 - ✗ Orbital element perturbation equations
 - ✗ Coordinate system transformations (TEME, J2000, etc.)
 - ✗ Historical epoch systems
-

Chapter 4: Coordinate Frames (ECI vs ECEF)

Why Multiple Coordinate Systems?

Problem: Earth rotates, but satellites orbit in (mostly) fixed planes.

Two perspectives:

1. **Inertial (space-fixed):** Ignores Earth's rotation - good for orbital mechanics
2. **Earth-fixed:** Rotates with Earth - good for ground locations

ECI - Earth-Centered Inertial

What it is: Origin at Earth's center, axes fixed relative to stars

Imagine freezing Earth's rotation:

- X-axis points toward vernal equinox (fixed star direction)
- Z-axis points to North Pole
- Y-axis completes right-hand system

Satellites orbit in (nearly) fixed ECI planes

Earth rotates underneath them

When to use:

- Calculating satellite positions
- Computing relative distances between satellites
- Orbital mechanics calculations

Python example:

```
python

from skyfield.api import EarthSatellite, load

ts = load.timescale()
t = ts.now()

# Get position in ECI (TEME frame, similar to ECI)
position = satellite.at(t)
x, y, z = position.position.km

print(f"ECI Position: ({x:.0f}, {y:.0f}, {z:.0f}) km")
# Example: (3421, -5102, 2814) km
```

ECEF - Earth-Centered Earth-Fixed

What it is: Origin at Earth's center, axes rotate with Earth

Axes fixed to Earth's surface:

- X-axis points to 0° longitude (prime meridian) on equator
- Z-axis points to North Pole
- Y-axis points to 90°E on equator

Ground locations have fixed ECEF coordinates

Satellites move in ECEF as Earth rotates

When to use:

- Ground station locations

- Sub-satellite points (lat/lon below satellite)
- Visualizing on map/globe

Python example:

```
python

# Get geographic location (converts ECI → ECEF → Lat/Lon)
geographic = position.subpoint()

print(f"Latitude: {geographic.latitude.degrees:.2f}°")
print(f"Longitude: {geographic.longitude.degrees:.2f}°")
print(f"Altitude: {geographic.elevation.km:.0f} km")
# Example: Lat: 28.5°, Lon: -80.1°, Alt: 420 km
```

For Collision Analysis: Use ECI

Critical rule: Always compute distances in ECI, never mix frames

```
python

# CORRECT: Both positions in same frame (ECI)
pos1_eci = sat1.at(t).position.km
pos2_eci = sat2.at(t).position.km
distance = np.linalg.norm(pos1_eci - pos2_eci)

# WRONG: Mixing frames
pos1_eci = sat1.at(t).position.km
pos2_latlon = sat2.at(t).subpoint() # Different frame!
# distance = ??? # This makes no sense!
```

Practical Code for Your Project

```
python
```

```

import numpy as np
from skyfield.api import EarthSatellite, load

def compute_distance_km(sat1, sat2, time):
    """Compute 3D distance between two satellites"""
    # Both in ECI frame automatically
    pos1 = sat1.at(time).position.km
    pos2 = sat2.at(time).position.km

    # Euclidean distance in 3D
    distance = np.linalg.norm(pos1 - pos2)
    return distance

def get_ground_track(satellite, time):
    """Get lat/lon for visualization"""
    pos = satellite.at(time)
    geo = pos.subpoint()

    return {
        'lat': geo.latitude.degrees,
        'lon': geo.longitude.degrees,
        'alt': geo.elevation.km
    }

# Usage
ts = load.timescale()
t = ts.now()

# Collision analysis: ECI distances
dist = compute_distance_km(sat1, sat2, t)
print(f"Separation: {dist:.2f} km")

# Visualization: ground tracks
track1 = get_ground_track(sat1, t)
track2 = get_ground_track(sat2, t)
print(f"Sat1 over: {track1['lat']:.1f}°, {track1['lon']:.1f}°")

```

Common Mistake to Avoid

python

```

# BAD: Computing distance from lat/lon
lat1, lon1 = 40.7, -74.0 # New York
lat2, lon2 = 34.0, -118.2 # Los Angeles

# This is ground distance, NOT satellite distance!
# Satellites at same lat/lon can be 1000 km apart vertically!

# GOOD: Always use 3D ECI positions
pos1_eci = [x1, y1, z1]
pos2_eci = [x2, y2, z2]
distance_3d = sqrt((x2-x1)**2 + (y2-y1)**2 + (z2-z1)**2)

```

What to Skip

- ✗ TEME vs J2000 vs GCRF frame differences
 - ✗ Precession and nutation corrections
 - ✗ Manual frame transformation matrices
 - ✗ Geodetic vs geocentric coordinates
-

Chapter 5: SGP4 Propagation Model

What is SGP4?

Simplified General Perturbations 4 - The standard algorithm for predicting satellite positions from TLEs.

Think of it as: GPS navigation for satellites

- Input: TLE + time
- Output: Position & velocity

Why "Simplified"?

- Doesn't model every tiny force
- Good enough for most collision analysis
- Fast to compute (thousands of satellites/second)

What SGP4 Accounts For

- ✓ Earth's gravity (main force)
- ✓ Earth's oblateness (J2 perturbation - equatorial bulge)
- ✓ Atmospheric drag (approximate model)
- ✓ Solar radiation pressure (approximate)

Third-body effects (sun/moon gravity) (approximate)

NOT modeled in detail:

- Solar weather variations
- Satellite maneuvers
- High-precision drag
- Relativistic effects

When SGP4 Breaks Down

Good for:

- LEO satellites (200-2000 km)
- Time spans < 7 days from TLE epoch
- Circular/near-circular orbits

Degraded accuracy:

- Very low altitude (< 200 km) - drag dominates
- Highly elliptical orbits
- > 7 days from TLE epoch
- After satellite maneuvers

Python: Using SGP4

Method 1: Direct sgp4 library

```
python
```

```
from sgp4.api import Satellite, jday
from datetime import datetime

# Parse TLE
line1 = "1 25544U 98067A 23365.50000000 ..."
line2 = "2 25544 51.6400 247.4627 ..."

satellite = Satellite(line1, line2)

# Propagate to specific time
year = 2023
month = 12
day = 31
hour = 12
minute = 0
second = 0

jd, fr = jday(year, month, day, hour, minute, second)
error_code, position, velocity = satellite.sgp4(jd, fr)

if error_code == 0:
    print(f"Position: {position} km")
    print(f"Velocity: {velocity} km/s")
else:
    print(f"Error: {error_code}")
```

Method 2: Skyfield (easier, recommended)

```
python
```

```
from skyfield.api import EarthSatellite, load

ts = load.timescale()

# Create satellite
satellite = EarthSatellite(line1, line2, "ISS")

# Propagate to now
t = ts.now()
position = satellite.at(t)

# Get position (ECI frame)
x, y, z = position.position.km
vx, vy, vz = position.velocity.km_per_s

print(f"Position: ({x:.0f}, {y:.0f}, {z:.0f}) km")
print(f"Velocity: ({vx:.2f}, {vy:.2f}, {vz:.2f}) km/s")
```

Propagation Over Time Window

For collision analysis, check multiple time steps:

python

```

import numpy as np

def propagate_window(satellite, start_time, duration_hours, step_minutes=10):
    """
    Propagate satellite over time window
    """

    Returns:
        times: array of time objects
        positions: array of (x,y,z) positions
        velocities: array of (vx,vy,vz) velocities
    """

    ts = load.timescale()

    # Generate time array
    num_steps = int(duration_hours * 60 / step_minutes)
    times = [start_time + i * step_minutes / (24 * 60) for i in range(num_steps)]

    positions = []
    velocities = []

    for t in times:
        pos = satellite.at(ts.from_datetime(t))
        positions.append(pos.position.km)
        velocities.append(pos.velocity.km_per_s)

    return times, np.array(positions), np.array(velocities)

# Example: Propagate ISS for 24 hours in 10-minute steps
from datetime import datetime, timezone

start = datetime.now(timezone.utc)
times, positions, velocities = propagate_window(satellite, start, 24, 10)

print(f"Propagated {len(times)} time steps")
print(f"First position: {positions[0]}")
print(f"Last position: {positions[-1]}")

```

Checking Propagation Errors

python

```
def check_sgp4_health(satellite, time):
    """Check if SGP4 propagation succeeded"""

    try:
        pos = satellite.at(time)
        position = pos.position.km

        # Check for NaN or unrealistic values
        if np.any(np.isnan(position)):
            return False, "NaN in position"

    distance_from_
```