

Caso03

November 10, 2025

1 Evaluación del caso “Kicked Kepler con impulsos periódicos”

Analizamos un sistema Kepleriano perturbado: una estrella central y una partícula ligera que recibe impulsos periódicos en velocidad. El objetivo es encontrar masas dentro de rangos estrechos que mitiguen la respuesta caótica y favorezcan órbitas cuasi periódicas pese a las perturbaciones.

Interpretación - La métrica de Lyapunov indica qué tan sensible es la órbita ante el “kick” periódico; valores más bajos implican trayectorias controladas. - La penalización de periodicidad ayuda a que el estado tras cada ciclo completo de impulsos permanezca próximo al inicial.

1.1 Preparación del entorno

Localizamos la carpeta raíz `two_body`, la añadimos a `sys.path` y definimos `PARENT` para acceder a utilidades compartidas. Así evitamos problemas de importación sin importar dónde se ejecute el notebook.

```
[1]: import sys
from pathlib import Path

PROJECT_ROOT = Path.cwd().resolve()
while PROJECT_ROOT.name != "two_body" and PROJECT_ROOT.parent != PROJECT_ROOT:
    PROJECT_ROOT = PROJECT_ROOT.parent

if PROJECT_ROOT.name != "two_body":
    raise RuntimeError("No se encontró la carpeta two_body")

PARENT = PROJECT_ROOT.parent # directorio que contiene a two_body
if str(PARENT) not in sys.path:
    sys.path.insert(0, str(PARENT))

print("PYTHONPATH += ", PARENT)
```

PYTHONPATH +=
C:\Users\emicr\Documents\CODIGOS_FUENTES\TrabajoTerminal\collision_of_two_bodies

1.2 Dependencias clave del pipeline

Importamos la configuración (`Config`), el controlador híbrido, los visualizadores 2D/3D, el adaptador `REBOUND` base y utilidades como `numpy` y `Path` para manejar resultados y artefactos.

```
[2]: from two_body import Config, set_global_seeds
from two_body.core.telemetry import setup_logger
from two_body.logic.controller import ContinuousOptimizationController
from two_body.presentation.visualization import Visualizer as PlanarVisualizer
from two_body.presentation.triDTry import Visualizer as Visualizer3D
from two_body.simulation.rebound_adapter import ReboundSim
import numpy as np
from pathlib import Path # si quieres guardar animaciones/figuras
```

1.3 Instrumentación de rendimiento

Activamos `PERF_TIMINGS_ENABLED` y cargamos las herramientas de la suite de tiempos (`time_block`, `latest_timing_csv`, etc.). Esto permitirá inspeccionar los costos de cada fase del experimento Kicked Kepler.

```
[3]: import os
os.environ["PERF_TIMINGS_ENABLED"] = "1"
os.environ.setdefault("PERF_TIMINGS_JSONL", "0")

from two_body.perf_timings.timers import time_block
from two_body.perf_timings import latest_timing_csv, read_timings_csv, □
    ↪parse_sections_arg, filter_rows
```

1.4 Formato de logging amigable en notebook

Creamos un `NotebookHandler` que acumula los mensajes del optimizador y los imprime progresivamente. Mantiene el contexto de ejecución visible sin saturar la salida con trazas crudas.

```
[ ]: import logging
from IPython.display import display, Markdown

class NotebookHandler(logging.Handler):
    def __init__(self):
        super().__init__()
        self.lines = []

    def emit(self, record):
        msg = self.format(record)
        self.lines.append(msg)
        print(msg)

handler = NotebookHandler()
handler.setFormatter(logging.Formatter("[%(asctime)s] %(levelname)s -□
    ↪%(message)s"))

logger = setup_logger(level="INFO")
logger.handlers.clear()
logger.addHandler(handler)
```

```
logger.setLevel(logging.INFO)
```

1.5 Configuración del escenario “Kicked Kepler” físico

Trabajamos en unidades astronómicas reales (UA, años, masas solares): - Órbita casi circular para un planeta tipo Tierra alrededor de una estrella ~ 1 Msun. - Impulsos periódicos pequeños (`kick_period`, `kick_delta_v`) que simulan maniobras tangenciales. - Presupuestos moderados del GA para iteraciones rápidas.

```
[ ]: # Kicked Kepler (unidades físicas)
case = {
    # Integración (UA, años, masas solares)
    "t_end_short": 0.5,
    "t_end_long": 4.0,
    "dt": 2.0e-4,
    "integrator": "ias15",

    # Estado inicial: estrella  $\sim 1$  Msun en reposo y planeta tipo Tierra a 1 UA
    "r0": (
        (0.0, 0.0, 0.0),           # estrella central
        (1.0, 0.0, 0.0),           # planeta en el eje X
    ),
    "v0": (
        (0.0, 0.0, 0.0),           # estrella fija (masa dominante)
        (0.0, 6.2831853072, 0.0),  # velocidad circular  $2\pi$  UA/año
    ),

    # Parámetros físicos
    "mass_bounds": (
        (0.8, 1.2),                # estrella cerca de 1 Msun
        (2.8e-6, 3.4e-2),           # planeta tipo Tierra (3e-6 Msun)
    ),
    "G": 39.47841760435743,
    "periodicity_weight": 0.05,

    # Impulso periódico ("kick")
    "kick_period": 0.25,          # cada 0.25 años
    "kick_delta_v": (0.0, 0.05, 0.0), # refuerzo tangencial  $\sim 0.24$  km/s
    "kick_target": 1,              # aplica el impulso al planeta

    # GA / búsqueda continua
    "pop_size": 180,
    "n_gen_step": 5,
    "mutation": 0.2,
    "crossover": 0.85,
    "elitism": 2,
    "seed": 314159,
```

```

# Control de ejecución
"max_epochs": 50,
"top_k_long": 12,
"stagnation_window": 5,
"stagnation_tol": 1.25e-4,
"local_radius": 0.04,
"radius_decay": 0.85,
"time_budget_s": 1800.0,
"eval_budget": 16000,

# Artefactos / salida
"artifacts_dir": "artifacts/kicked_kepler_real",
"save_plots": True,
"headless": False,
}

```

```

[6]: from two_body.logic.controller import ContinuousOptimizationController
from two_body.core.config import Config
from two_body.core.telemetry import setup_logger
from two_body.core.cache import HierarchicalCache

_kick_keys = (
    "kick_period",
    "kick_delta_v",
    "kick_target",
)
config_kwargs = {k: v for k, v in case.items() if k not in _kick_keys}

cfg = Config(**config_kwargs)
set_global_seeds(cfg.seed)
logger = setup_logger()

```

1.6 Adaptador REBOUND con impulsos periódicos

Subclasamos `ReboundSim` para construir `KickedKeplerSim`, que aplica un incremento de velocidad Δv al cuerpo objetivo cada `kick_period`. Registraremos la función en `post_timestep_modifications` y monkeypatcheamos el módulo `rebound_adapter` para que el resto del pipeline use este integrador “pateado”.

```

[ ]: from two_body.simulation.rebound_adapter import ReboundSim as _BaseReboundSim
kick_period = case["kick_period"]
kick_delta_v = case["kick_delta_v"]
kick_target = case["kick_target"]

class KickedKeplerSim(_BaseReboundSim):
    def __init__(self, **kwargs):

```

```

super().__init__(**kwargs)
self._kick_period = kick_period
self._kick_delta_v = kick_delta_v
self._kick_target = kick_target

def setup_simulation(self, *args, **kwargs):
    sim = super().setup_simulation(*args, **kwargs)
    next_kick = {"t": sim.t + self._kick_period}
    dv = self._kick_delta_v

    def apply_kick(_ptr=None):
        while sim.t >= next_kick["t"] - 1e-12:
            p = sim.particles[self._kick_target]
            p.vx += dv[0]
            p.vy += dv[1]
            p.vz += dv[2]
            next_kick["t"] += self._kick_period

    sim.post_timestep_modifications = apply_kick
    return sim

# Patch para que todo el pipeline use el adaptador pateado
from two_body.simulation import rebound_adapter
rebound_adapter.ReboundSim = KickedKeplerSim

```

```
[8]: from two_body.simulation.lyapunov import LyapunovEstimator

masses = tuple(np.mean(bounds) for bounds in cfg.mass_bounds)
sim = ReboundSim(G=cfg.G, integrator=cfg.integrator).setup_simulation(
    masses,
    cfg.r0[: len(masses)],
    cfg.v0[: len(masses)],
)
estimator = LyapunovEstimator()
ret = estimator.mLCE({"sim": sim, "dt": cfg.dt, "t_end": cfg.t_end_short, "masses": masses})
print(ret)

{'lambda': 5.617102963610244, 'series': None, 'meta': {'steps': 2500, 'dt': 0.0002, 'n_bodies': 2, 'masses': (1.0, 0.0170014), 'impl': 'rebound_megno'}}

c:\Users\emicr\anaconda3\envs\grav2body\Lib\site-
packages\rebound\__init__.py:58: UserWarning: pkg_resources is deprecated as an
API. See https://setuptools.pypa.io/en/latest/pkg_resources.html. The
pkg_resources package is slated for removal as early as 2025-11-30. Refrain from
using this package or pin to Setuptools<81.
    import pkg_resources
```

```
[9]: print(cfg.mass_bounds, cfg.max_epochs, cfg.eval_budget)
```

```
((0.8, 1.2), (2.8e-06, 0.034)) 50 16000
```

1.7 Ejecución del controlador híbrido

Instanciamos Config, el logger y el ContinuousOptimizationController. Envolvemos controller.run() dentro de time_block("notebook_run") para medir la ejecución completa (GA + refinamiento continuo) bajo el régimen de impulsos periódicos.

```
[10]: with time_block("notebook_run", extra={"source": "Caso03.ipynb"}):
    controller = ContinuousOptimizationController(cfg, logger=logger)
    results = controller.run()
```

```
[2025-11-02 18:03:24,171] INFO - Starting optimization | pop=180 | dims=2 |
time_budget=1800.0s | eval_budget=16000
[2025-11-02 18:03:34,260] INFO - Epoch 0 | new global best (short)
lambda=-0.608653 | fitness=-0.130490 | penalty=14.782862 | masses=(1.076408,
0.009538)
[2025-11-02 18:03:39,255] INFO - Epoch 0 | new global best (long)
lambda=0.078349 | fitness=-0.112894 | penalty=0.690898 | masses=(1.112924,
0.018475)
[2025-11-02 18:03:39,255] INFO - Epoch 0 complete | lambda_short=-0.608653 |
fitness_short=-0.130490 | lambda_best=0.078349 | fitness_best=-0.112894 | evals
short/long=180/12 | total evals=192 | radius=0.0400
[2025-11-02 18:03:49,009] INFO - Epoch 1 | new global best (short)
lambda=-1.175435 | fitness=0.522169 | penalty=13.065319 | masses=(0.893851,
0.005434)
[2025-11-02 18:03:54,206] INFO - Epoch 1 complete | lambda_short=-1.175435 |
fitness_short=0.522169 | lambda_best=-1.175435 | fitness_best=0.522169 | evals
short/long=180/12 | total evals=384 | radius=0.0400
[2025-11-02 18:04:09,359] INFO - Epoch 2 complete | lambda_short=-0.328350 |
fitness_short=-0.411257 | lambda_best=-1.175435 | fitness_best=0.522169 | evals
short/long=180/12 | total evals=576 | radius=0.0400
[2025-11-02 18:04:24,495] INFO - Epoch 3 complete | lambda_short=0.077930 |
fitness_short=-0.814718 | lambda_best=-1.175435 | fitness_best=0.522169 | evals
short/long=180/12 | total evals=768 | radius=0.0400
[2025-11-02 18:04:39,871] INFO - Epoch 4 complete | lambda_short=-0.284319 |
fitness_short=-0.459426 | lambda_best=-1.175435 | fitness_best=0.522169 | evals
short/long=180/12 | total evals=960 | radius=0.0400
[2025-11-02 18:04:55,263] INFO - Epoch 5 complete | lambda_short=-1.054964 |
fitness_short=0.327293 | lambda_best=-1.175435 | fitness_best=0.522169 | evals
short/long=180/12 | total evals=1152 | radius=0.0400
[2025-11-02 18:05:05,541] INFO - Epoch 6 | new global best (short)
lambda=-2.384170 | fitness=1.644374 | penalty=14.795926 | masses=(1.079796,
0.00437)
[2025-11-02 18:05:10,891] INFO - Epoch 6 complete | lambda_short=-2.384170 |
fitness_short=1.644374 | lambda_best=-2.384170 | fitness_best=1.644374 | evals
short/long=180/12 | total evals=1344 | radius=0.0400
```

```
[2025-11-02 18:05:26,349] INFO - Epoch 7 complete | lambda_short=0.506178 |
fitness_short=-1.068172 | lambda_best=-2.384170 | fitness_best=1.644374 | evals
short/long=180/12 | total evals=1536 | radius=0.0400
[2025-11-02 18:05:41,846] INFO - Epoch 8 complete | lambda_short=-1.061196 |
fitness_short=0.388638 | lambda_best=-2.384170 | fitness_best=1.644374 | evals
short/long=180/12 | total evals=1728 | radius=0.0400
[2025-11-02 18:05:57,269] INFO - Epoch 9 complete | lambda_short=1.128368 |
fitness_short=-1.739622 | lambda_best=-2.384170 | fitness_best=1.644374 | evals
short/long=180/12 | total evals=1920 | radius=0.0400
[2025-11-02 18:06:12,802] INFO - Epoch 10 complete | lambda_short=-0.296447 |
fitness_short=-0.440785 | lambda_best=-2.384170 | fitness_best=1.644374 | evals
short/long=180/12 | total evals=2112 | radius=0.0400
[2025-11-02 18:06:28,600] INFO - Stagnation detected; reseeding around best
candidate.
[2025-11-02 18:06:28,600] INFO - Epoch 11 complete | lambda_short=-0.540672 |
fitness_short=-0.122644 | lambda_best=-2.384170 | fitness_best=1.644374 | evals
short/long=180/12 | total evals=2304 | radius=0.0340
[2025-11-02 18:06:44,314] INFO - Epoch 12 complete | lambda_short=-0.634294 |
fitness_short=-0.097665 | lambda_best=-2.384170 | fitness_best=1.644374 | evals
short/long=180/12 | total evals=2496 | radius=0.0340
[2025-11-02 18:06:54,590] INFO - Epoch 13 | new global best (short)
lambda=-2.569516 | fitness=1.825080 | penalty=14.888726 | masses=(1.051172,
3e-06)
[2025-11-02 18:06:59,971] INFO - Epoch 13 complete | lambda_short=-2.569516 |
fitness_short=1.825080 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=2688 | radius=0.0340
[2025-11-02 18:07:15,685] INFO - Epoch 14 complete | lambda_short=0.877516 |
fitness_short=-1.602276 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=2880 | radius=0.0340
[2025-11-02 18:07:31,304] INFO - Epoch 15 complete | lambda_short=-0.109073 |
fitness_short=-0.633062 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=3072 | radius=0.0340
[2025-11-02 18:07:47,503] INFO - Epoch 16 complete | lambda_short=-1.035795 |
fitness_short=0.305423 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=3264 | radius=0.0340
[2025-11-02 18:08:04,529] INFO - Epoch 17 complete | lambda_short=-2.562232 |
fitness_short=1.818351 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=3456 | radius=0.0340
[2025-11-02 18:08:21,860] INFO - Stagnation detected; reseeding around best
candidate.
[2025-11-02 18:08:21,860] INFO - Epoch 18 complete | lambda_short=0.000140 |
fitness_short=-0.743577 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=3648 | radius=0.0289
[2025-11-02 18:08:38,244] INFO - Epoch 19 complete | lambda_short=-0.250048 |
fitness_short=-0.492338 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=3840 | radius=0.0289
[2025-11-02 18:08:54,744] INFO - Epoch 20 complete | lambda_short=-0.328173 |
fitness_short=-0.416361 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
```

```
short/long=180/12 | total evals=4032 | radius=0.0289
[2025-11-02 18:09:11,746] INFO - Epoch 21 complete | lambda_short=-0.907719 |
fitness_short=0.163288 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=4224 | radius=0.0289
[2025-11-02 18:09:28,692] INFO - Epoch 22 complete | lambda_short=0.253645 |
fitness_short=-0.993983 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=4416 | radius=0.0289
[2025-11-02 18:09:45,890] INFO - Stagnation detected; reseeding around best
candidate.
[2025-11-02 18:09:45,890] INFO - Epoch 23 complete | lambda_short=-0.776251 |
fitness_short=0.035909 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=4608 | radius=0.0246
[2025-11-02 18:10:02,312] INFO - Epoch 24 complete | lambda_short=-2.542646 |
fitness_short=1.800844 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=4800 | radius=0.0246
[2025-11-02 18:10:19,404] INFO - Epoch 25 complete | lambda_short=-0.350518 |
fitness_short=-0.393468 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=4992 | radius=0.0246
[2025-11-02 18:10:36,470] INFO - Epoch 26 complete | lambda_short=-0.495415 |
fitness_short=-0.248981 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=5184 | radius=0.0246
[2025-11-02 18:10:53,478] INFO - Epoch 27 complete | lambda_short=-2.479341 |
fitness_short=1.740821 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=5376 | radius=0.0246
[2025-11-02 18:11:10,386] INFO - Stagnation detected; reseeding around best
candidate.
[2025-11-02 18:11:10,386] INFO - Epoch 28 complete | lambda_short=-0.799035 |
fitness_short=0.055854 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=5568 | radius=0.0209
[2025-11-02 18:11:27,295] INFO - Epoch 29 complete | lambda_short=-0.836794 |
fitness_short=0.092289 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=5760 | radius=0.0209
[2025-11-02 18:11:44,300] INFO - Epoch 30 complete | lambda_short=-0.708816 |
fitness_short=-0.031383 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=5952 | radius=0.0209
[2025-11-02 18:12:01,063] INFO - Epoch 31 complete | lambda_short=-2.549279 |
fitness_short=1.807070 | lambda_best=-2.569516 | fitness_best=1.825080 | evals
short/long=180/12 | total evals=6144 | radius=0.0209
[2025-11-02 18:12:12,443] INFO - Epoch 32 | new global best (short)
lambda=-2.569736 | fitness=1.825861 | penalty=14.877505 | masses=(1.045259,
3e-06)
[2025-11-02 18:12:18,320] INFO - Epoch 32 complete | lambda_short=-2.569736 |
fitness_short=1.825861 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=6336 | radius=0.0209
[2025-11-02 18:12:35,088] INFO - Epoch 33 complete | lambda_short=-0.873832 |
fitness_short=0.132522 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=6528 | radius=0.0209
[2025-11-02 18:12:52,174] INFO - Epoch 34 complete | lambda_short=-2.511178 |
```

```
fitness_short=1.766649 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=6720 | radius=0.0209
[2025-11-02 18:13:08,867] INFO - Epoch 35 complete | lambda_short=0.231212 |
fitness_short=-0.972695 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=6912 | radius=0.0209
[2025-11-02 18:13:25,660] INFO - Epoch 36 complete | lambda_short=-0.467087 |
fitness_short=-0.277482 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=7104 | radius=0.0209
[2025-11-02 18:13:42,800] INFO - Stagnation detected; reseeding around best
candidate.
[2025-11-02 18:13:42,800] INFO - Epoch 37 complete | lambda_short=-0.588201 |
fitness_short=-0.156385 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=7296 | radius=0.0177
[2025-11-02 18:13:59,351] INFO - Epoch 38 complete | lambda_short=-0.401247 |
fitness_short=-0.343302 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=7488 | radius=0.0177
[2025-11-02 18:14:16,849] INFO - Epoch 39 complete | lambda_short=-0.536986 |
fitness_short=-0.206327 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=7680 | radius=0.0177
[2025-11-02 18:14:33,732] INFO - Epoch 40 complete | lambda_short=0.586378 |
fitness_short=-1.330922 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=7872 | radius=0.0177
[2025-11-02 18:14:50,043] INFO - Epoch 41 complete | lambda_short=-0.458505 |
fitness_short=-0.286085 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=8064 | radius=0.0177
[2025-11-02 18:15:06,875] INFO - Stagnation detected; reseeding around best
candidate.
[2025-11-02 18:15:06,876] INFO - Epoch 42 complete | lambda_short=-0.522419 |
fitness_short=-0.221399 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=8256 | radius=0.0151
[2025-11-02 18:15:23,814] INFO - Epoch 43 complete | lambda_short=-0.348077 |
fitness_short=-0.392797 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=8448 | radius=0.0151
[2025-11-02 18:15:40,489] INFO - Epoch 44 complete | lambda_short=-0.919933 |
fitness_short=0.175671 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=8640 | radius=0.0151
[2025-11-02 18:15:57,639] INFO - Epoch 45 complete | lambda_short=0.591647 |
fitness_short=-1.335774 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=8832 | radius=0.0151
[2025-11-02 18:16:15,157] INFO - Epoch 46 complete | lambda_short=-0.802527 |
fitness_short=0.058030 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=9024 | radius=0.0151
[2025-11-02 18:16:32,481] INFO - Stagnation detected; reseeding around best
candidate.
[2025-11-02 18:16:32,481] INFO - Epoch 47 complete | lambda_short=-0.551856 |
fitness_short=-0.191244 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=9216 | radius=0.0128
[2025-11-02 18:16:49,298] INFO - Epoch 48 complete | lambda_short=-2.546347 |
```

```

fitness_short=1.804238 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=9408 | radius=0.0128
[2025-11-02 18:17:05,979] INFO - Epoch 49 complete | lambda_short=-0.266154 |
fitness_short=-0.477131 | lambda_best=-2.569736 | fitness_best=1.825861 | evals
short/long=180/12 | total evals=9600 | radius=0.0128
[2025-11-02 18:17:05,980] INFO - Optimization completed | epochs=50 | evals=9600
| best lambda=-2.569736 | wall=821.8s

```

1.8 Métricas de referencia y solución óptima

Calculamos el fitness del punto central de los rangos de masa (`center`) y lo comparamos con `results["best"]`. Obtenemos `baseline_details` para revisar cómo afectan los kicks al exponente de Lyapunov y conservamos `metrics` para graficar la evolución de γ y del fitness.

```

[11]: metrics = controller.metrics
results

[11]: {'status': 'completed',
       'best': {'masses': [1.045258635678602, 2.8e-06],
                 'lambda': -2.569736031111161,
                 'fitness': 1.825860801467893,
                 'm1': 1.045258635678602,
                 'm2': 2.8e-06},
       'evals': 9600,
       'epochs': 50}

[12]: from two_body.core.cache import HierarchicalCache
from two_body.logic.fitness import FitnessEvaluator

#original_masses = tuple((lo + hi) / 2.0 for lo, hi in cfg.mass_bounds)
original_masses = (1.2, 3.3e-4)
center = original_masses

baseline_entry = ret
baseline_lambda_short = baseline_entry.get("lambda")
if baseline_lambda_short is None or not np.isfinite(baseline_lambda_short):
    baseline_lambda_short = -float(baseline_entry.get("fitness", 0.0))

best_payload = results.get("best") or {}
best_lambda_short = best_payload.get("lambda")
if best_lambda_short is None and best_payload.get("fitness") is not None:
    best_lambda_short = -float(best_payload["fitness"])

print(
    f" original (short) = {baseline_lambda_short:.6f}, "
    f" optimizado (short) = {best_lambda_short:.6f}"
    if best_lambda_short is not None
    else f" original (short) = {baseline_lambda_short:.6f}, optimizado = N/A"
)

```

```

)
RUN_LONG_CHECK = False
if RUN_LONG_CHECK and best_payload.get("masses"):
    evaluator = FitnessEvaluator(HierarchicalCache(), cfg)
    orig_fits, orig_details = evaluator.evaluate_batch(
        [original_masses],
        horizon="long",
        return_details=True,
    )
    opt_fits, opt_details = evaluator.evaluate_batch(
        [tuple(best_payload["masses"])],
        horizon="long",
        return_details=True,
    )
    lambda_orig_long = orig_details[0].get("lambda")
    if lambda_orig_long is None or not np.isfinite(lambda_orig_long):
        lambda_orig_long = -orig_fits[0]
    lambda_opt_long = opt_details[0].get("lambda")
    if lambda_opt_long is None or not np.isfinite(lambda_opt_long):
        lambda_opt_long = -opt_fits[0]
    print(
        f" original (long) = {lambda_orig_long:.6f}, "
        f" optimizado (long) = {lambda_opt_long:.6f}"
    )

```

original (short) = 5.617103, optimizado (short) = -2.569736

```

[13]: sim_builder = ReboundSim(G=cfg.G, integrator=cfg.integrator)
best_masses = tuple(results["best"]["masses"])

def _slice_vectors(vectors, count):
    if len(vectors) < count:
        raise ValueError("Config no tiene suficientes vectores iniciales")
    return tuple(tuple(float(coord) for coord in vectors[i]) for i in range(count))

r0 = _slice_vectors(cfg.r0, len(best_masses))
v0 = _slice_vectors(cfg.v0, len(best_masses))

sim = sim_builder.setup_simulation(best_masses, r0, v0)
traj = sim_builder.integrate(sim, t_end=cfg.t_end_long, dt=cfg.dt)
print("Trayectoria calculada con masas óptimas.")
print(traj.shape)
print(traj[-1])
xyz_tracks = [traj[:, i, :3] for i in range(traj.shape[1])]

```

Trayectoria calculada con masas óptimas.

```
(20000, 2, 6)
[[ 1.43029738e-06 -2.05143528e-06  0.00000000e+00  1.44315179e-05
  1.08237131e-05  0.00000000e+00]
 [-5.33939533e-01  7.65814443e-01  0.00000000e+00 -5.38738169e+00
 -4.04056415e+00  0.00000000e+00]]
```

```
[14]: from two_body.scripts.demo_tierra import (
    summarize_trajectory,
    compute_total_energy,
    estimate_orbital_period,
)

summarize_trajectory(
    logger=logger,
    traj=traj,
    masses=best_masses,
    cfg=cfg,
)
```

```
[2025-11-02 18:17:06,527] INFO - Resumen de simulacion
[2025-11-02 18:17:06,527] INFO -      pasos=20000 | dt=0.000200 anos | duracion
total=4.000 anos
[2025-11-02 18:17:06,527] INFO -      masas=(1.045258635678602, 2.8e-06) (M_sun) |
G=39.478418
[2025-11-02 18:17:06,528] INFO -      centro de masa: desplazamiento maximo =
2.066e-21 UA
[2025-11-02 18:17:06,530] INFO -      cuerpo 0 -> radio[min, max]=[0.0000, 0.0000]
UA | radio sigma=7.7750e-08 | velocidad media=0.0000 UA/ano
[2025-11-02 18:17:06,531] INFO -      cuerpo 1 -> radio[min, max]=[0.9170, 1.0000]
UA | radio sigma=2.9024e-02 | velocidad media=6.5501 UA/ano
[2025-11-02 18:17:06,534] INFO -      energia total (media)=-6.027280e-05 |
variacion relativa=1.574e-15
[2025-11-02 18:17:06,536] INFO -      periodo orbital estimado para la Tierra ==
0.917852 anos
[2025-11-02 18:17:06,536] INFO -      error relativo vs 1 ano =~ 8.215e-02
```

1.9 Visualización y seguimiento de convergencia

Mostramos: - `plot_lambda_evolution` y `plot_fitness_evolution` para estudiar cómo responde el GA ante los impulsos repetidos. - Las trayectorias integradas con las masas óptimas (`xyz_tracks`) tanto en vista rápida 2D (`quick_view`) como en la animación 3D final, verificando si la órbita se mantiene acotada tras sucesivos kicks.

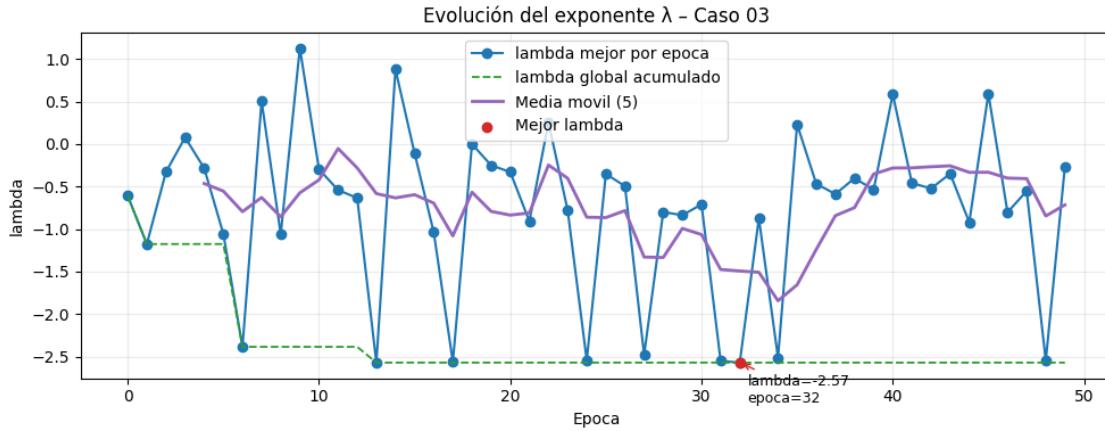
```
[15]: viz_3d = Visualizer3D(headless=cfg.headless)

_ = viz_3d.plot_lambda_evolution(
    lambda_history=metrics.best_lambda_per_epoch,
    epoch_history=metrics.epoch_history,
```

```

        title="Evolución del exponente - Caso 03",
        moving_average_window=5,    # opcional
)

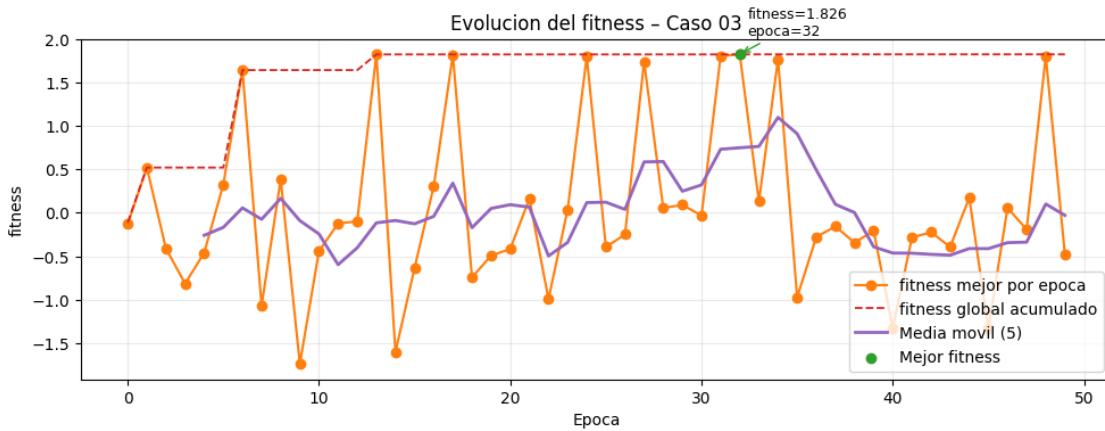
```



```

[16]: _ = viz_3d.plot_fitness_evolution(
    fitness_history=metrics.best_fitness_per_epoch,
    epoch_history=metrics.epoch_history,
    title="Evolucion del fitness - Caso 03",
    moving_average_window=5,
)

```



```

[17]: sim_builder = ReboundSim(G=cfg.G, integrator=cfg.integrator)
best_masses = tuple(results["best"]["masses"])

def _slice_vectors(vectors, count):
    if len(vectors) < count:

```

```

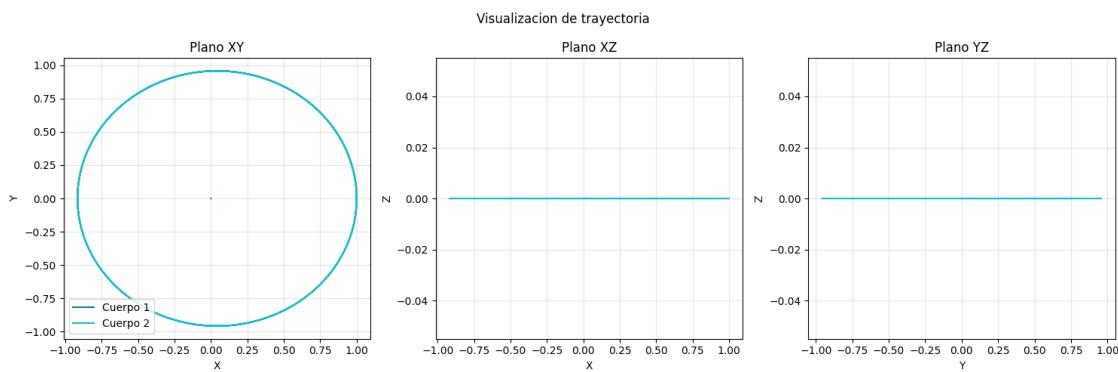
        raise ValueError("Config no tiene suficientes vectores iniciales")
    return tuple(tuple(float(coord) for coord in vectors[i]) for i in
    range(count))

r0 = _slice_vectors(cfg.r0, len(best_masses))
v0 = _slice_vectors(cfg.v0, len(best_masses))

sim = sim_builder.setup_simulation(best_masses, r0, v0)
traj = sim_builder.integrate(sim, t_end=cfg.t_end_long, dt=cfg.dt)
xyz_tracks = [traj[:, i, :3] for i in range(traj.shape[1])]

```

[18]: viz_planar = PlanarVisualizer(headless=cfg.headless)
 $_$ = viz_planar.quick_view(xyz_tracks)

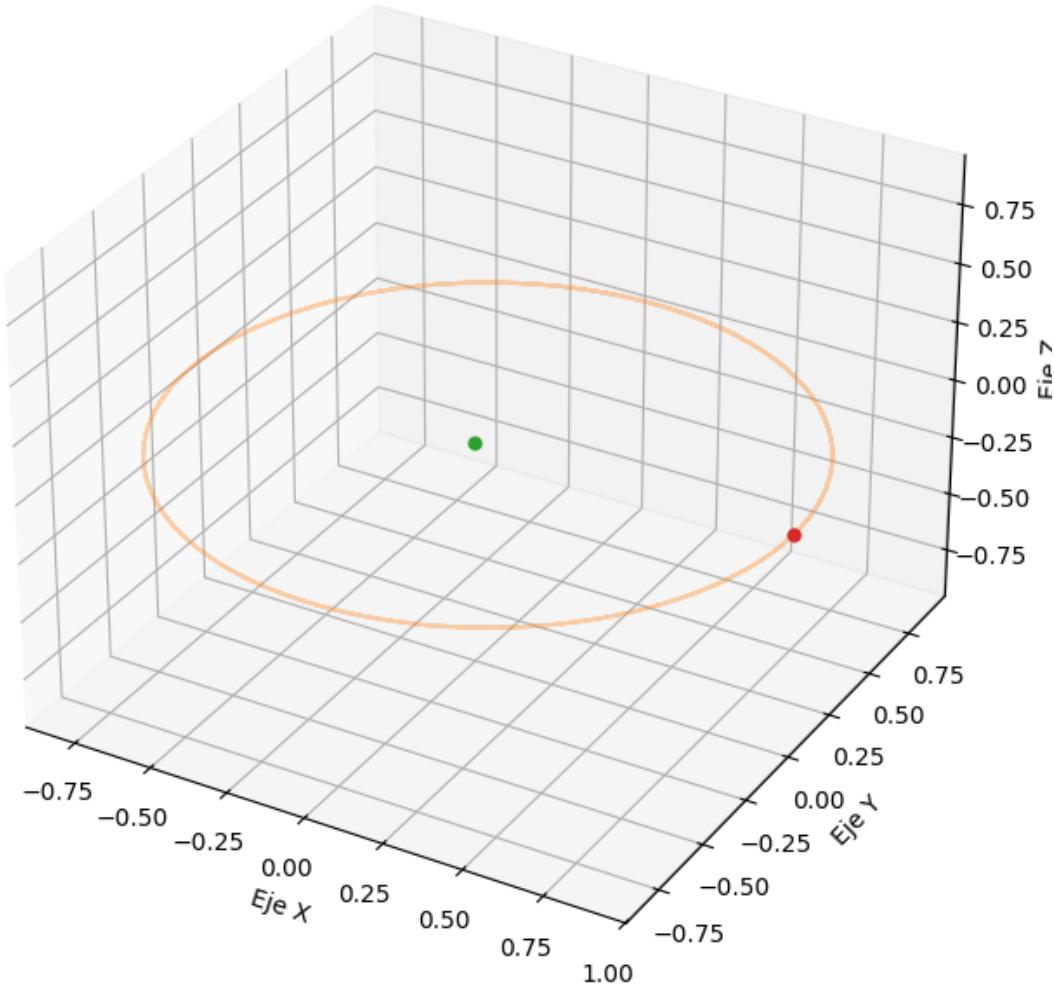


[19]: from IPython.display import HTML
import matplotlib as mpl
mpl.rcParams['animation.embed_limit'] = 1024 # MB

[20]: viz_3d = Visualizer3D(headless=False)

anim = viz_3d.animate_3d(
 trajectories=xyz_tracks,
 interval_ms=50,
 title=f"Trayectorias 3D m1={best_masses[0]:.3f}, m2={best_masses[1]:.3f}",
 total_frames=len(xyz_tracks[0]),
)
#HTML(anim.to_jshtml())

Trayectorias 3D $m_1=1.045$, $m_2=0.000$



1.10 Exportación de animaciones

Configuramos un `FFMpegWriter`, creamos `artifacts/caso03` (ajusta la ruta si lo prefieres) y guardamos los MP4 de la trayectoria y de la comparación de masas. Puedes modificar `fps`, `bitrate` o el preset de `ffmpeg` si necesitas acelerar el render.

```
[21]: from matplotlib.animation import FFMpegWriter # o PillowWriter para GIF

writer = FFMpegWriter(
    fps=20,
    bitrate=1800,
    extra_args=["-vcodec", "libx264", "-preset", "ultrafast", "-crf", "28"]
```

```

)
output_path = Path("artifacts/caso03")                                # ajusta a tu gusto
output_path.mkdir(parents=True, exist_ok=True)

anim.save(output_path / "trayectoria_optima.mp4", writer=writer)

[ ]: # Trayectoria con las masas originales (center)
sim_orig = sim_builder.setup_simulation(center, r0, v0)
traj_original = sim_builder.integrate(sim_orig, t_end=cfg.t_end_long, dt=cfg.dt)

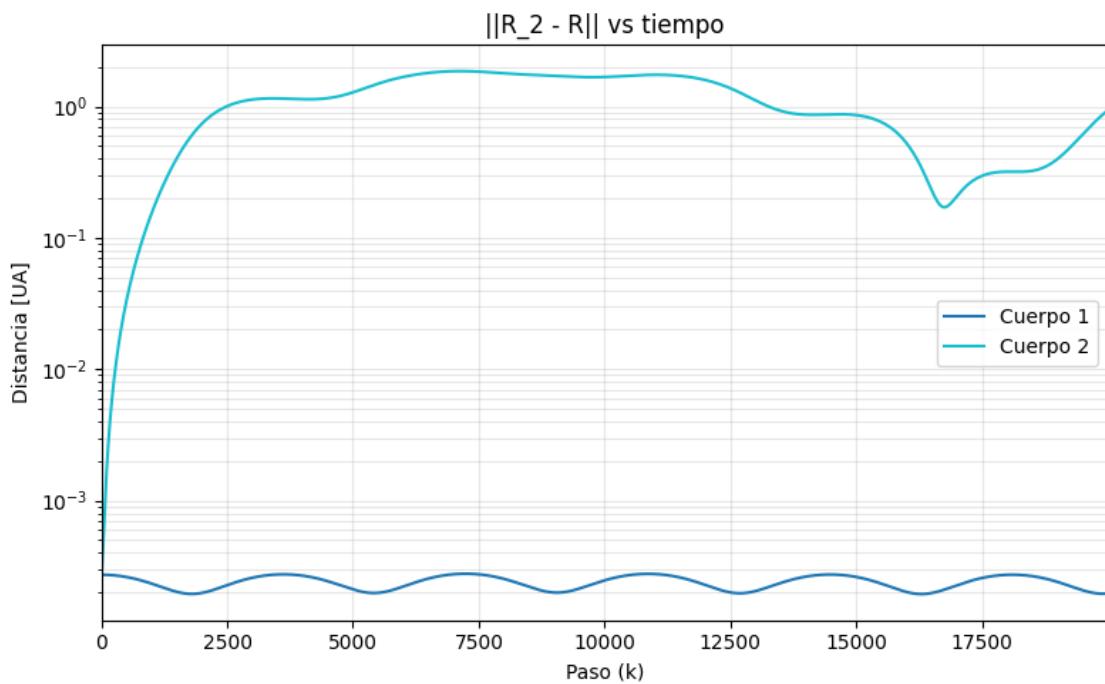
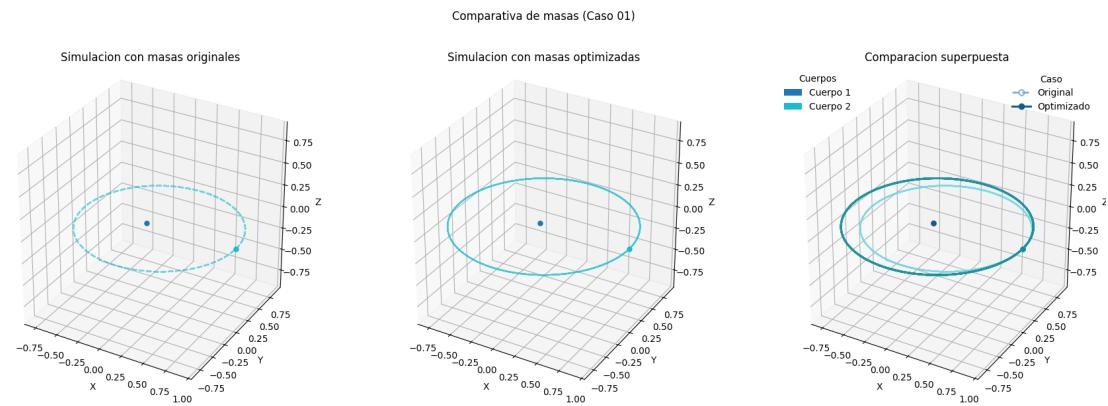
traj_opt = traj # 'traj' es la integración que acabas de calcular con
               ↵best_masses

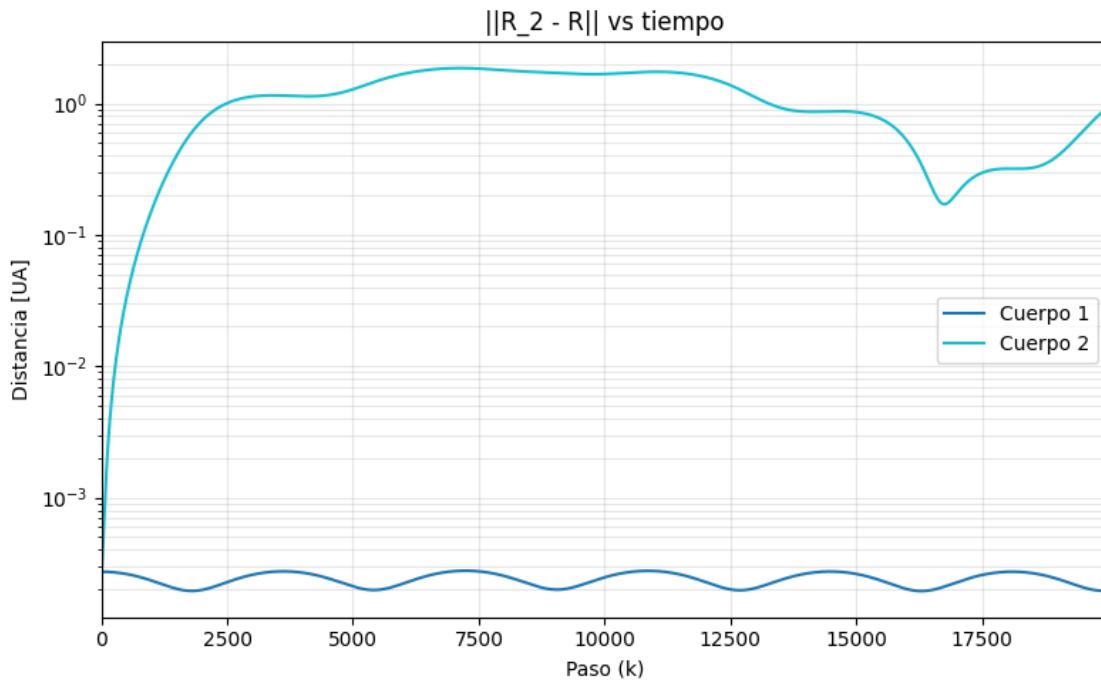
[ ]: orig_tracks = [traj_original[:, i, :3] for i in range(traj_original.shape[1])]
opt_tracks = [traj_opt[:, i, :3] for i in range(traj_opt.shape[1])]

anim_mass = viz_3d.plot_mass_comparison(
    original_tracks=orig_tracks,
    optimized_tracks=opt_tracks,
    original_masses=center,           # opcional, útil para etiquetar/distancias
    optimized_masses=best_masses,     # idem
    body_labels=[f"Cuerpo {i+1}" for i in range(len(opt_tracks))],
    dt=cfg.dt,
    title="Comparativa de masas (Caso 01)",
)

if anim_mass is not None:
    dist_fig = viz_3d.plot_mass_distance_evolution(
        comparison_data=anim_mass.mass_comparison_data,
        title="||R_2 - R|| vs tiempo",
    )
    if dist_fig is not None:
        display(dist_fig)
# display(HTML(anim_mass.to_jshtml())) # descomenta para ver la animación
               ↵embebida

```





```
[24]: anim_mass.save(output_path / "comparativa_masas.mp4", writer=writer)
```

1.11 Reporte de tiempos de ejecución

Leemos el CSV generado por la instrumentación, mostramos una vista preliminar y agregamos estadísticas por sección para detectar qué partes del flujo (simulación con kicks, evaluación larga, etc.) dominan el costo computacional en este caso.

```
[ ]: import pandas as pd

csv_path = latest_timing_csv()
display(f"Usando CSV: {csv_path}")

rows = read_timings_csv(csv_path)
df = pd.DataFrame(rows)
display(df.head(10))

# Estadísticas rápidas por sección
section_stats = (
    df.groupby("section")["duration_us"]
    .agg(["count", "mean", "sum"])
    .sort_values("sum", ascending=False)
)
section_stats
```

```

NameError Traceback (most recent call last)
Cell In[2], line 3
      1 import pandas as pd
----> 3 csv_path = latest_timing_csv()
      4 display(f"Usando CSV: {csv_path}")
      5 rows = read_timings_csv(csv_path)

NameError: name 'latest_timing_csv' is not defined

```

```

[1]: import os
import subprocess
from pathlib import Path
from IPython.display import Image, display

PROJECT_ROOT = Path.cwd()
while PROJECT_ROOT.name != "two_body" and PROJECT_ROOT.parent != PROJECT_ROOT:
    PROJECT_ROOT = PROJECT_ROOT.parent

env = os.environ.copy()
env["PYTHONPATH"] = str(PROJECT_ROOT)

run_id = df["run_id"].iloc[0]
cmd = [
    sys.executable,
    "scripts/plot_timings.py",
    "--run-id", str(run_id),
    "--top-n", "5",
]

print("Ejecutando:", " ".join(cmd))
result = subprocess.run(cmd, cwd=PROJECT_ROOT, env=env, text=True, capture_output=True)
print(result.stdout)
print(result.stderr)
result.check_returncode()

reports_dir = PROJECT_ROOT / "reports"

display(
    Image(filename=str(reports_dir / f"timeline_by_individual_{run_id}.png")),
    Image(filename=str(reports_dir / f"timeline_by_batch_{run_id}.png")),
    Image(filename=str(reports_dir / f"timeline_simulation_{run_id}.png")),
    Image(filename=str(reports_dir / f"pie_sections_{run_id}.png")),
)

```

```
)
```

```
NameError Traceback (most recent call last)
Cell In[1], line 13
      10 env = os.environ.copy()
      11 env["PYTHONPATH"] = str(PROJECT_ROOT)
---> 13 run_id = df["run_id"].iloc[0]
      14 cmd = [
      15     sys.executable,
      16     "scripts/plot_timings.py",
      17     "--run-id", str(run_id),
      18     "--top-n", "5",
      19 ]
      22 print("Ejecutando:", " ".join(cmd))

NameError: name 'df' is not defined
```