

00_X- _Simulation

September 15, 2025

1 X- Framework

- -AB fringe shifts vs. effective flux
- (null spatial EM),
- Cross-Hall drift using a split-step propagation (2D grid + internal rotor treated via),
- Rotor sidebands and level shifts vs. I and

,

- Classical bounce scale min a min

and WDW barrier coefficient,

Shared-range

predictions for gravity/QED (tables + CSV).

Each experiment saves CSVs and figures to /paper for downstream analysis.

1.1 Cell 2: Imports & global constants

```
[52]: # Core imports
from pathlib import Path
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from numpy.fft import fft2, ifft2, fftshift, ifftshift, fftfreq

# Reproducibility
np.random.seed(42)

# Physical constants (SI)
hbar = 1.054_571_817e-34 # J·s
h     = 6.626_070_15e-34 # J·s
G     = 6.674_30e-11      # m^3 kg^-1 s^-2
c     = 299_792_458.0     # m/s
pi    = np.pi
```

```

# repo dirs → paper/*
PROJECT_ROOT_NAME = "X-theta-framework"
repo_root = Path.cwd().resolve()
for p in [repo_root, *repo_root.parents]:
    if p.name == PROJECT_ROOT_NAME:
        repo_root = p; break
PAPER_DIR = repo_root / "paper"
FIG_DIR    = PAPER_DIR / "figs"
TAB_DIR    = PAPER_DIR / "tables"
DATA_DIR   = PAPER_DIR / "data"
TEXT_DIR   = PAPER_DIR / "analysis"
for d in [FIG_DIR, TAB_DIR, DATA_DIR, TEXT_DIR]:
    d.mkdir(parents=True, exist_ok=True)

print("DATA_DIR:", DATA_DIR.resolve())
print("TAB_DIR :", TAB_DIR.resolve())
print("TEXT_DIR:", TEXT_DIR.resolve())

# Helper: save CSV with a small banner
def save_csv(df: pd.DataFrame, path: str):
    df.to_csv(path, index=False)
    print(f"[saved CSV] {path} rows={len(df)}")

# Helper: simple image saver with tight layout
def save_figure(path: str):
    plt.tight_layout()
    plt.savefig(path, dpi=150, bbox_inches='tight')
    print(f"[saved FIG] {path}")

```

```

DATA_DIR: C:\workspace\Physics\X-theta-framework\paper\data
TAB_DIR  : C:\workspace\Physics\X-theta-framework\paper\tables
TEXT_DIR: C:\workspace\Physics\X-theta-framework\paper\analysis

```

1.2 Cell 3:

1.2.1 Experiment 1: $-AB$ fringes vs. effective flux ϕ_θ (null EM)

Goal Demonstrate the 2π -periodic interferometric phase at $\mathbf{E} = \mathbf{B} = 0$:

$$\Delta\varphi_\theta \equiv \phi_\theta \bmod 2\pi$$

Outputs

- **Plot:** Intensity vs. ϕ_θ over several periods.
- **CSV:** `exp1_fringe_vs_phi.csv` with columns (`phi_theta`, `intensity`).

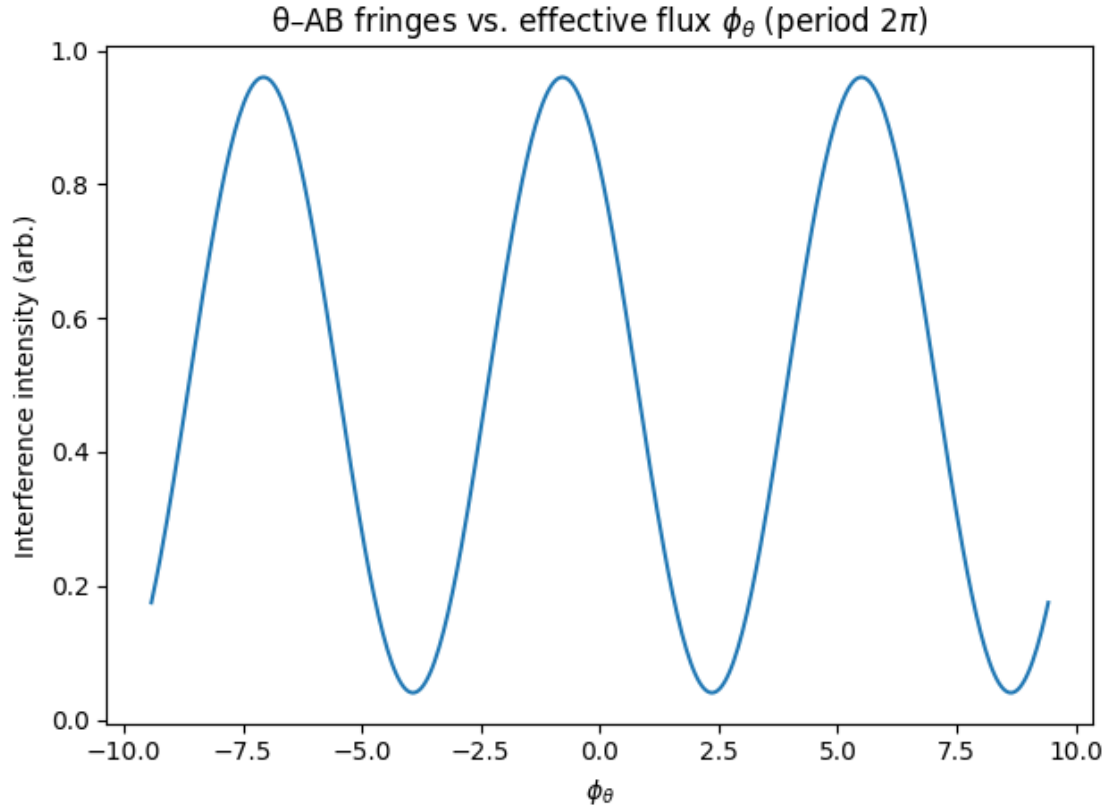
```
[53]: # --- knobs ---
phi_min, phi_max = -3*np.pi, 3*np.pi # sweep range
Nphi = 400 # resolution
visibility = 0.92 # fringe contrast (0..1)
phase_bias = 0.25*np.pi # optional static bias phase

# --- compute ---
phi = np.linspace(phi_min, phi_max, Nphi)
intensity = 0.5*(1 + visibility*np.cos(phi + phase_bias))

# --- figure ---
plt.figure()
plt.plot(phi, intensity)
plt.xlabel(r"$\phi_\theta$")
plt.ylabel("Interference intensity (arb.)")
plt.title("-AB fringes vs. effective flux $\phi_\theta$ (period $2\pi$)")
fig_path = FIG_DIR / "exp1_fringe_vs_phi.png"
#print("FIG_DIR:", FIG_DIR.resolve())
save_figure(fig_path)
plt.show()

# --- csv ---
df1 = pd.DataFrame({"phi_theta": phi, "intensity": intensity})
csv_path = DATA_DIR / "exp1_fringe_vs_phi.csv"
save_csv(df1, str(csv_path))
df1.head()
```

[saved FIG] C:\workspace\Physics\X-theta-framework\paper\figs\exp1_fringe_vs_phi.png



[saved CSV] C:\workspace\Physics\X-theta-framework\paper\data\exp1_fringe_vs_phi.csv rows=400

```
[53]:  phi_theta  intensity
0   -9.424778  0.174731
1   -9.377536  0.190454
2   -9.330294  0.206869
3   -9.283052  0.223937
4   -9.235810  0.241621
```

2 Cell 5:

2.0.1 Experiment 2: Cross-Hall drift via split-step (2D grid, null spatial EM)

Model Schrödinger on (x, y) with internal rotor momentum label ℓ .

With $A_\theta(y) = A_0 + g_y y$ and fixed ℓ , the potential is

$$V(y) = -\frac{\hbar \ell q_\theta}{I} A_\theta(y) + \frac{q_\theta^2}{2I} A_\theta(y)^2 \quad (+\text{constant } \hbar^2 \ell^2 / 2I)$$

which yields a transverse force $\propto \partial_y A_\theta$.

Procedure

1. Propagate a Gaussian packet with initial momentum along $+x$ using split-step FFT.
 2. Record centroid $\langle y(t) \rangle$ and final deflection Δy .
 3. Verify $\Delta y \propto T^2$ by varying total time.
-

Outputs

- **Plot 1:** $\langle y(t) \rangle$ for one run.
- **Plot 2:** Δy vs. T^2 with linear fit.
- **CSV:** exp2_drift_T2.csv with columns (T, T2, delta_y).

```
[54]: # --- knobs (physical & numerical) ---
m      = 1.44e-25      # kg (Rb atom-ish; any test mass OK)
I      = 1.0e-38       # J·s2 (choose to give resolvable effect)
qtheta = 1.0e-34       # "charge" units so that qtheta*A_theta has momentum
    ↪ units
ell     = 1            # rotor Fourier index
A0      = 0.0          # baseline internal potential
gradA   = 5.0e-6       # gradient [A_theta per meter]
Lx      = 2.0e-3       # m
Ly      = 2.0e-3       # m
Nx      = 128
Ny      = 128
x       = np.linspace(-Lx/2, Lx/2, Nx, endpoint=False)
y       = np.linspace(-Ly/2, Ly/2, Ny, endpoint=False)
dx      = x[1]-x[0]
dy      = y[1]-y[0]
X, Y    = np.meshgrid(x, y, indexing='xy')

# K-space for kinetic propagation
kx = 2*np.pi*fftfreq(Nx, d=dx)
ky = 2*np.pi*fftfreq(Ny, d=dy)
KX, KY = np.meshgrid(kx, ky, indexing='xy')
Kfactor = np.exp(-1j * (hbar/(2*m)) * (KX**2 + KY**2)) # this is the dt=1
    ↪ factor; we'll exponentiate to dt later

# A_theta(y), potential V(y)
A_theta = A0 + gradA*Y
V = -(hbar*ell*qtheta/I) * A_theta + (qtheta**2/(2*I)) * (A_theta**2)
# constant rotor term hbar2 ell2/(2I) is omitted (global phase)

# Initial packet
```

```

x0, y0 = -0.6*Lx, 0.0
sigma = 0.12e-3
p0x = 3.0e-27 # kg·m/s
psi0 = np.exp(-((X-x0)**2 + (Y-y0)**2)/(2*sigma**2)) * np.exp(1j * (p0x/
    hbar) * X)
psi0 = psi0 / np.sqrt((np.abs(psi0)**2).sum())

# Time controls
T_list = np.array([0.0020, 0.0025, 0.0030, 0.0035]) # seconds
Nt = 240
dt_base = T_list.max()/Nt

def propagate(psi, T, Nt):
    dt = T/Nt
    # Adjust kinetic factor for dt
    Kdt = np.exp(-1j * (hbar/(2*m)) * (KX**2 + KY**2) * dt)
    psi_t = psi.copy()
    y_centroids = []
    for _ in range(Nt):
        # half potential
        psi_t *= np.exp(-1j * V * (dt/2) / hbar)
        # kinetic (FFT)
        psi_k = fft2(psi_t)
        psi_k *= Kdt
        psi_t = ifft2(psi_k)
        # half potential
        psi_t *= np.exp(-1j * V * (dt/2) / hbar)
        # record centroid
        prob = np.abs(psi_t)**2
        prob /= prob.sum()
        y_centroids.append((prob*Y).sum())
    return np.array(y_centroids), psi_t

# Single-run trace for plotting <y(t)>
T_demo = T_list[1]
y_traj, psi_f = propagate(psi0, T_demo, Nt)

plt.figure()
plt.plot(np.linspace(0, T_demo, len(y_traj)), y_traj)
plt.xlabel("t [s]")
plt.ylabel(r"$\langle y(t) \rangle$ [m]")
plt.title("Transverse centroid vs. time (single run)")
fig_path1 = FIG_DIR / "exp2_y_traj.png"
save_figure(fig_path1)
plt.show()

# Sweep T and verify Delta y ~ T^2

```

```

rows = []
for T in T_list:
    y_traj, _ = propagate(psi0, T, Nt)
    dy = y_traj[-1] - y_traj[0]
    rows.append((T, T**2, dy))

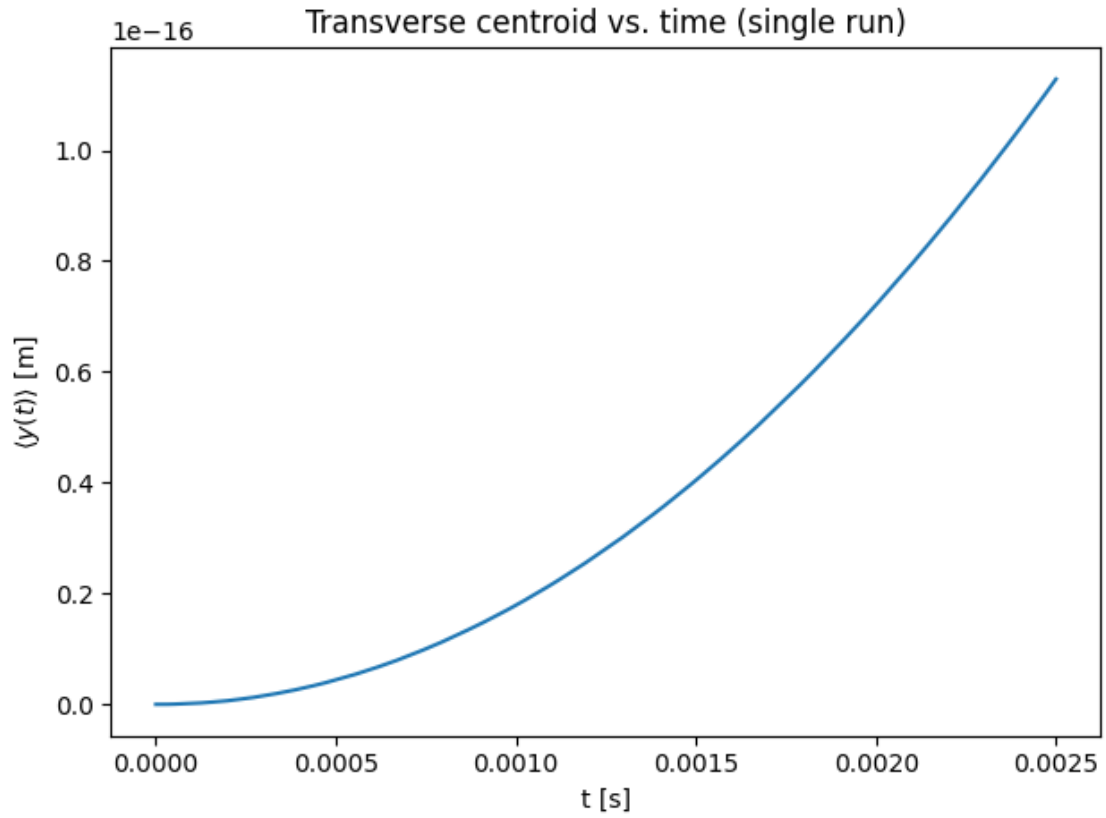
df2 = pd.DataFrame(rows, columns=["T", "T2", "delta_y"])
print(df2)
save_csv(df2, str(DATA_DIR / "exp2_drift_T2.csv"))

# Plot Delta y vs T^2
plt.figure()
plt.scatter(df2["T2"], df2["delta_y"])
# Linear fit
coef = np.polyfit(df2["T2"].values, df2["delta_y"].values, 1)
fit_y = np.polyval(coef, df2["T2"].values)
plt.plot(df2["T2"], fit_y)
plt.xlabel(r"$T^2$ [s$^2$]")
plt.ylabel(r"$\Delta y$ [m]")
plt.title(r"Cross-Hall drift: $\Delta y \propto T^2$ (fit slope = %.3e m/
↪s$^2$)" % coef[0])
fig_path2 = FIG_DIR / "exp2_dy_vs_T2.png"
save_figure(fig_path2)
plt.show()

df2

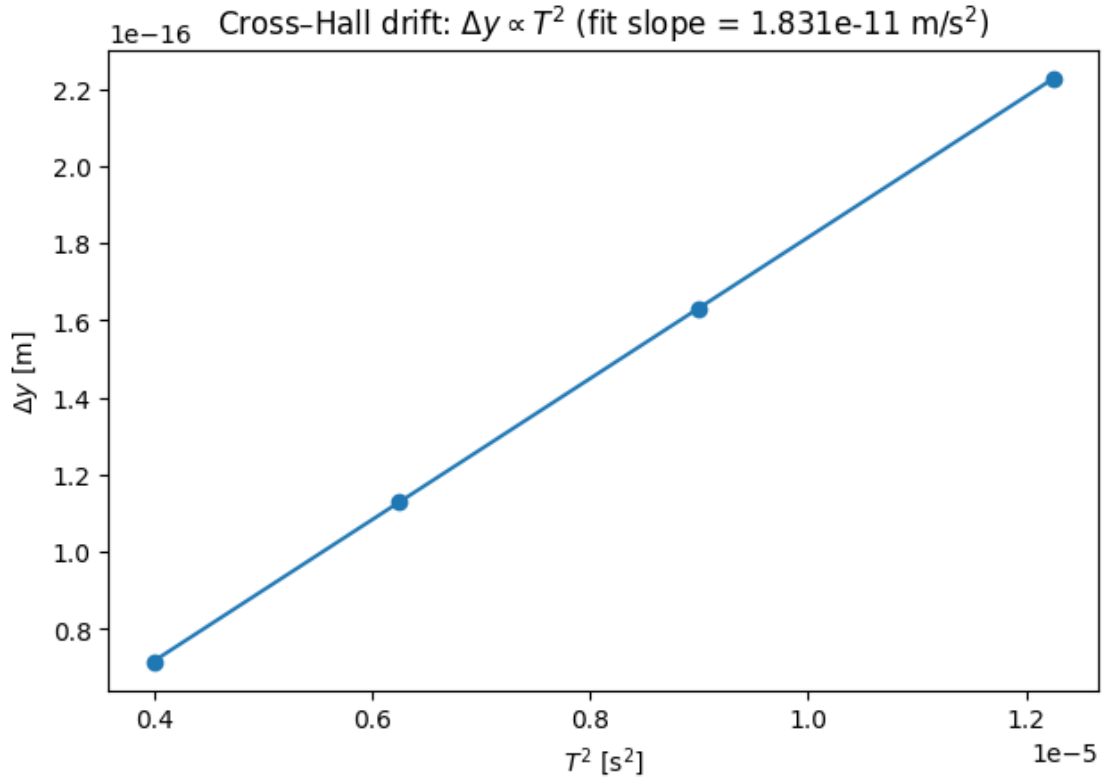
```

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	T	T2	delta_y
0	0.0020	0.000004	7.153024e-17
1	0.0025	0.000006	1.128112e-16
2	0.0030	0.000009	1.631182e-16
3	0.0035	0.000012	2.226138e-16

[saved CSV] C:\workspace\Physics\X-theta-framework\paper\data\exp2_drift_T2.csv
rows=4
[saved FIG] C:\workspace\Physics\X-theta-framework\paper\figs\exp2_dy_vs_T2.png



[54]:

	T	T2	delta_y
0	0.0020	0.000004	7.153024e-17
1	0.0025	0.000006	1.128112e-16
2	0.0030	0.000009	1.631182e-16
3	0.0035	0.000012	2.226138e-16

2.1 Cell 7

2.1.1 Experiment 3: Rotor sidebands and holonomy shift

Model The energy levels E_ℓ of the rotor are given by:

$$E_\ell = \frac{\hbar^2}{2I} \left(\ell - \frac{\phi_\theta}{2\pi} \right)^2$$

Outputs

- **Plot:** E_ℓ vs. ℓ for chosen values of I and ϕ_θ .
- **CSV:** exp3_rotor_levels.csv

```

[55]: # --- knobs ---
from pathlib import Path

I_rot      = 8.4e-39      # J·s2 (~1 kHz spacing)
phi_theta  = 0.6*np.pi   # effective holonomy
Lmax       = 6           # levels from -Lmax..Lmax

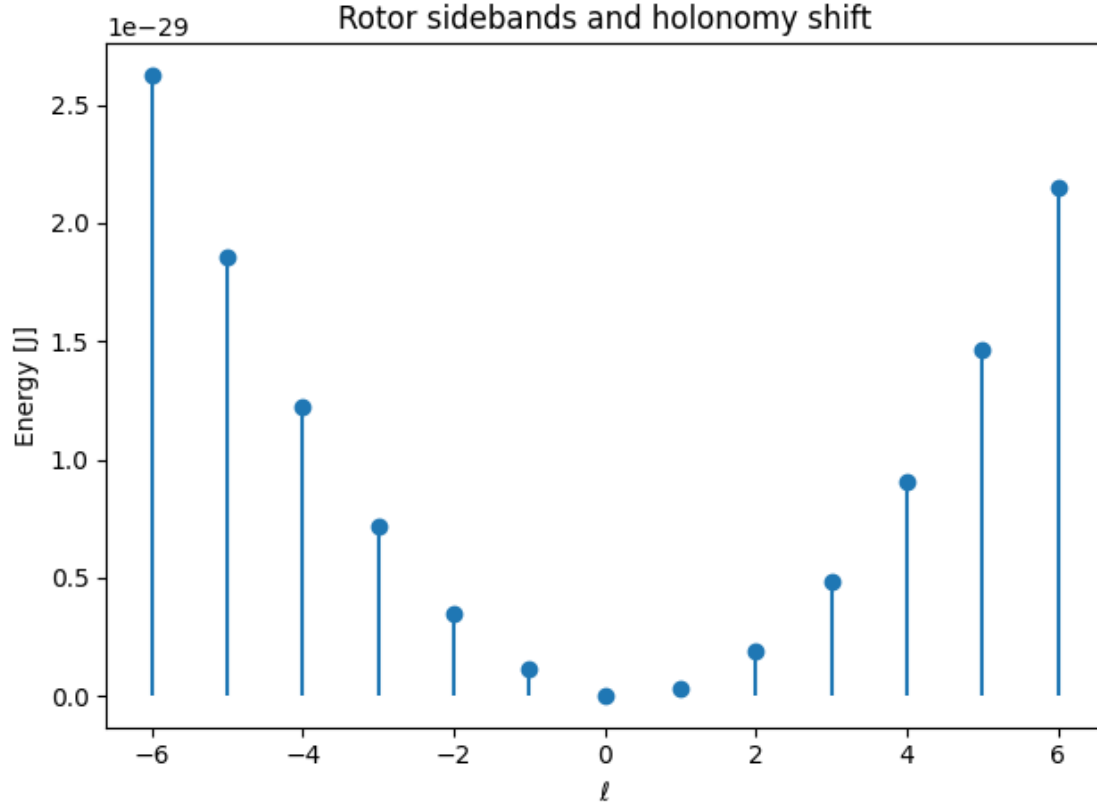
ells = np.arange(-Lmax, Lmax+1)
E = (hbar**2/(2*I_rot)) * (ells - phi_theta/(2*np.pi))**2

# --- plot ---
fig, ax = plt.subplots()
markerline, stemlines, baseline = ax.stem(ells, E) # <-- no use_line_collection
baseline.set_visible(False)                        # optional: hide the
    ↪ horizontal baseline
ax.set_xlabel(r"$\ell$")                          # fix label: single
    ↪ backslashes
ax.set_ylabel("Energy [J]")
ax.set_title("Rotor sidebands and holonomy shift")
fig_path3 = FIG_DIR / "exp3_rotor_levels.png"
save_figure(fig_path3)
plt.show()

# --- csv ---
df3 = pd.DataFrame({"ell": ells, "Energy_J": E})
save_csv(df3, str(DATA_DIR / "exp3_rotor_levels.csv"))
df3.head()

```

[saved FIG] C:\workspace\Physics\X-theta-framework\paper\figs\exp3_rotor_levels.png



[saved CSV] C:\workspace\Physics\X-theta-framework\paper\data\exp3_rotor_levels.csv rows=13

```
[55]:      ell      Energy_J
0    -6  2.627388e-29
1    -5  1.859494e-29
2    -4  1.223996e-29
3    -3  7.208932e-30
4    -2  3.501859e-30
```

2.2 Cell 9

2.2.1 Experiment 4: Classical bounce a_{\min} and WDW barrier coefficient

Formulas Early-time (curvature-assisted) bounce with a positive shear-like term:

$$A = \frac{8\pi G}{3} \frac{\Pi_\theta^2}{2I_0}, \quad a_{\min} = \left(\frac{A + \Sigma^2}{k} \right)^{1/4}, \quad C = \frac{\ell^2 \hbar^2}{2I_0} \quad (\text{WDW barrier coefficient})$$

Outputs

- **Plot:** a_{\min} vs. Π_{θ} (log-log optional).
- **CSV:** exp4_bounce_scan.csv

```
[60]: # --- knobs ---
I0      = 1.0e-38          # J·s2
Sigma2  = (1.0e-6)**2      # dimensionless (enter appropriate units for your
    ↪ normalization)
k_curv  = 1.0e-2           # curvature parameter > 0 (model units)
ell_wdw = 3                # typical Fourier label in WDW
Pi_grid = np.geomspace(1e-40, 1e-34, 60) # scan

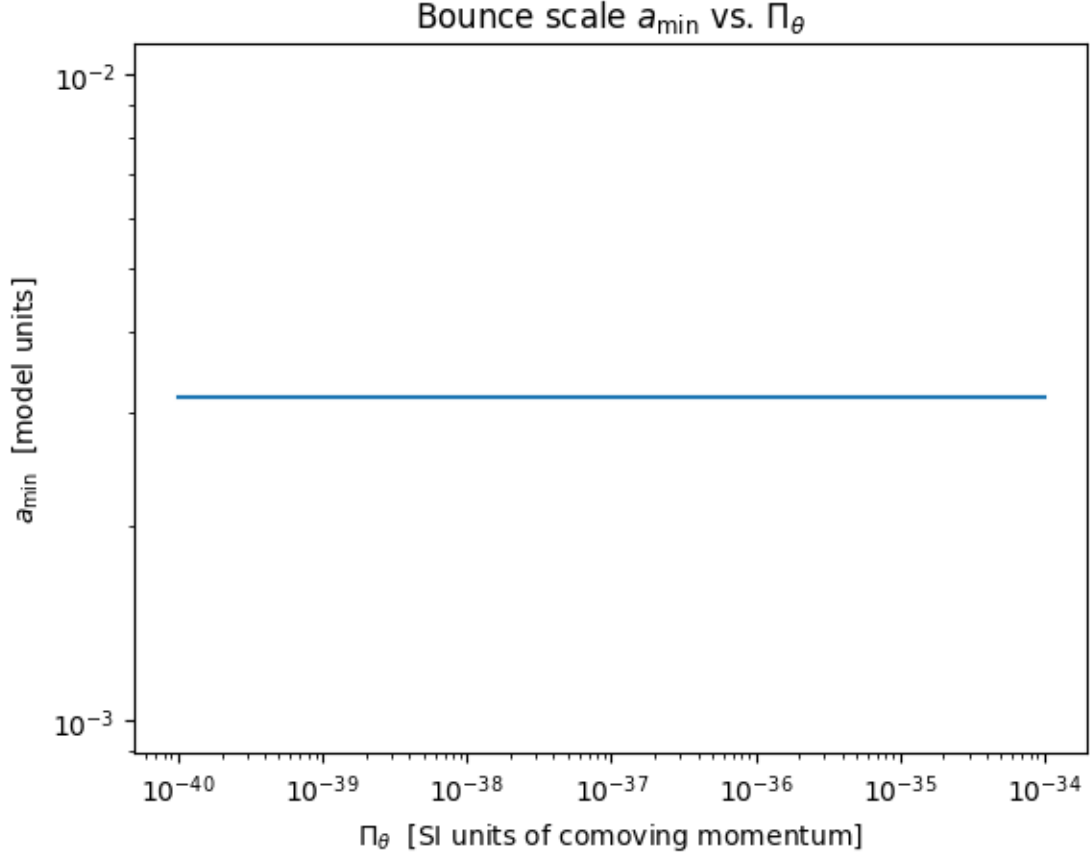
# Compute
A_vals  = (8*np.pi*G/3.0) * (Pi_grid**2)/(2*I0)
a_min   = ((A_vals + Sigma2)/k_curv)**0.25
C_wdw   = (ell_wdw**2 * hbar**2)/(2*I0)

# Plots
plt.figure()
plt.plot(Pi_grid, a_min)
plt.xscale("log")
plt.yscale("log")
plt.xlabel(r"$\Pi_{\theta}$ [SI units of comoving momentum]")
plt.ylabel(r"$a_{\min}$ [model units]")
plt.title("Bounce scale $a_{\min}$ vs. $\Pi_{\theta}$")
fig_path3 = FIG_DIR / "exp4_bounce_scan.png"
#save_figure(fig_path3)
plt.show()

# Save CSV
df4 = pd.DataFrame({"Pi_theta": Pi_grid, "A_value": A_vals, "a_min": a_min})
save_csv(df4, str(DATA_DIR / "exp4_bounce_scan.csv"))

# Print WDW barrier coefficient
print(f"WDW inverse-square barrier coefficient C = {C_wdw:.3e} J (in
    ↪ minisuperspace units; compare within your normalization)")

df4.head()
```



[saved CSV] C:\workspace\Physics\X-theta-framework\paper\data\exp4_bounce_scan.csv rows=60
 WDW inverse-square barrier coefficient $C = 5.005\text{e-}30$ J (in minisuperspace units; compare within your normalization)

[60]:

	Pi_theta	A_value	a_min
0	1.000000e-40	2.795724e-52	0.003162
1	1.263848e-40	4.465645e-52	0.003162
2	1.597312e-40	7.133029e-52	0.003162
3	2.018760e-40	1.139367e-51	0.003162
4	2.551407e-40	1.819926e-51	0.003162

2.3 Cell 11

2.3.1 Experiment 5: Shared-range λ_θ potentials (gravity & QED)

Gravity (fifth-force form) The gravitational potential is modified as follows:

$$V_G(r) = -\frac{Gm_1m_2}{r} [1 + \alpha_G e^{-r/\lambda_\theta}]$$

If $Q_\theta = \beta m$, then the coupling constant α_G is given by:

$$\alpha_G = \frac{g_\theta^2 \beta^2}{4\pi G}$$

(This is composition-independent to leading order).

QED with kinetic mixing The electromagnetic potential is modified as:

$$V_{\text{EM}}(r) = \frac{\alpha Q_1 Q_2}{r} + \varepsilon^2 \alpha Q_1 Q_2 \frac{e^{-r/\lambda_\theta}}{r}, \quad \text{where } \alpha = \frac{e^2}{4\pi}$$

Outputs

- **Plot:** Yukawa factor $e^{-r/\lambda}$ and fractional deviations vs. r .
- **CSV:** `exp5_yukawa_profiles.csv`

```
[62]: # --- knobs ---
lam    = 1.0e-3          # meters (1 mm range)
alphaG = 1.0e-2          # dimensionless strength in gravity channel
eps2    = 1.0e-6          # epsilon^2 for QED mixing
Q1Q2    = 1.0            # unit charges for EM demo
r = np.geomspace(1e-5, 1e-1, 400)

# Profiles
yuk = np.exp(-r/lam)
frac_G = alphaG * yuk          # fractional deviation in gravity
frac_EM = eps2 * yuk          # fractional deviation in EM (on top of
    ↪ Coulomb)

# Plots (two separate figures)
plt.figure()
plt.plot(r, yuk)
plt.xscale("log"); plt.yscale("linear")
plt.xlabel("r [m]")
plt.ylabel("$e^{-r/\\lambda_\\theta}$")
plt.title("Yukawa profile vs. distance")
fig_path4 = FIG_DIR / "exp5_yukawa_profile.png"
save_figure(fig_path4)
plt.show()

plt.figure()
plt.plot(r, frac_G, label="gravity fraction")
plt.plot(r, frac_EM, label="EM fraction")
plt.xscale("log")
plt.xlabel("r [m]")
plt.ylabel("fractional deviation")
```

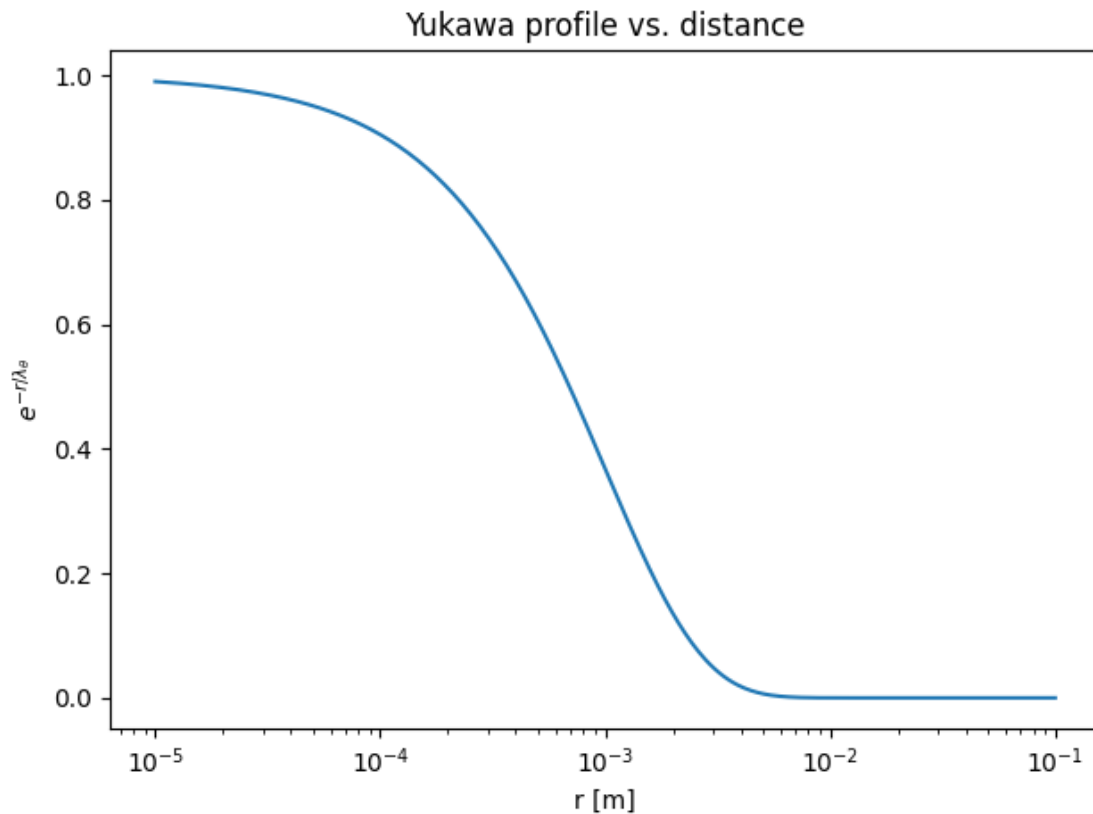
```

plt.title("Fractional deviations vs. distance")
plt.legend()
fig_path5 = FIG_DIR / "exp5_fractional_deviation.png"
save_figure(fig_path5)
plt.show()

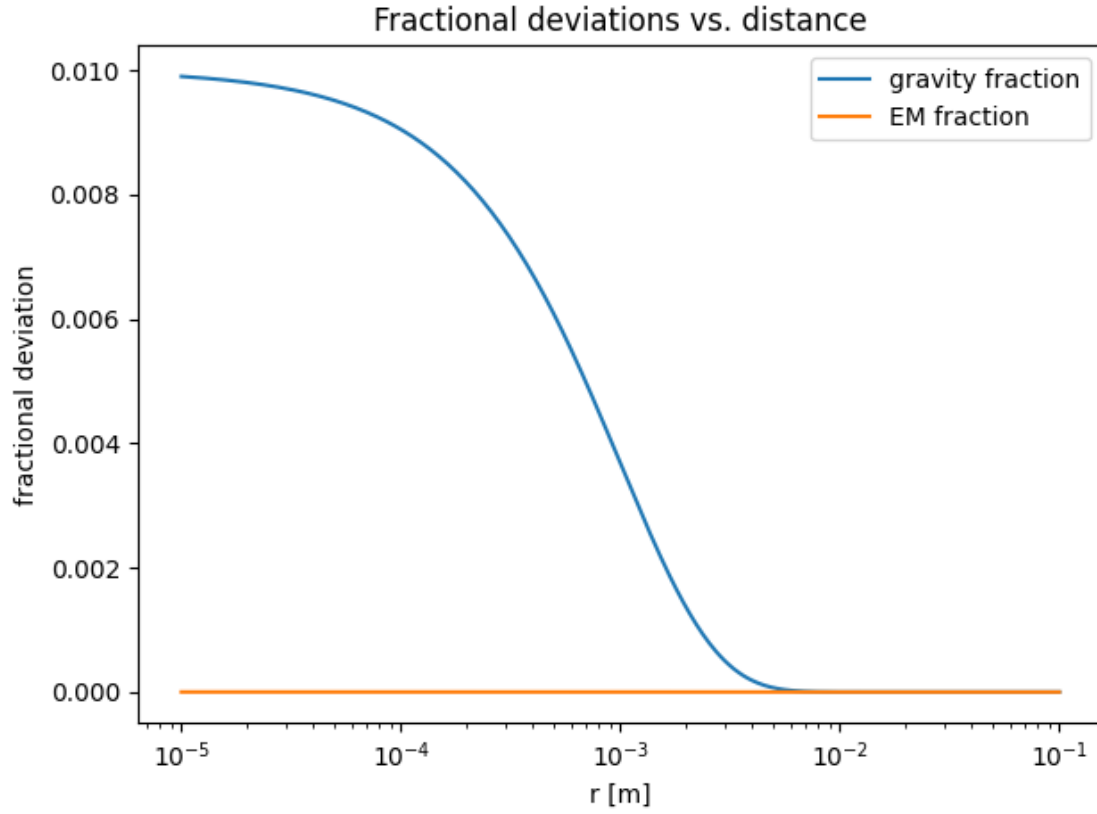
# CSV
df5 = pd.DataFrame({"r_m": r, "yukawa": yuk, "frac_gravity": frac_G, "frac_EM": frac_EM})
save_csv(df5, str(DATA_DIR / "exp5_yukawa_profiles.csv"))
df5.head()

```

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[saved FIG] C:\workspace\Physics\X-theta-framework\paper\figs\exp5_fractional_deviation.png



[saved CSV] C:\workspace\Physics\X-theta-framework\paper\data\exp5_yukawa_profiles.csv rows=400

[62]:

	r_m	yukawa	frac_gravity	frac_EM
0	0.000010	0.990050	0.009900	9.900498e-07
1	0.000010	0.989819	0.009898	9.898187e-07
2	0.000010	0.989582	0.009896	9.895822e-07
3	0.000011	0.989340	0.009893	9.893402e-07
4	0.000011	0.989093	0.009891	9.890926e-07