

```
In [1]: import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt
```

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In [6]: #parameters setting
#lng = 0.02 + 0.02*epsilon + nu
#M = 0.99*g^(-gamma)
#M = 0.99*exp(b_0*(1-g))
epsilon_num = int(1e4)
pre_nu_num = int(1e4)

dis_val = np.log(0.6)
nor_val = 0
dis_probability = 0.02

consumption_growth_intercept = 0.02
consumption_growth_slope = 0.02

#ε (epsilon)
df_consumption_growth = pd.DataFrame(np.random.randn(epsilon_num), columns=

lst_pre_nu = list(np.random.rand(pre_nu_num))

lst_nu = list(map(lambda x: dis_val if x < dis_probability else nor_val, ls

df_consumption_growth1 = df_consumption_growth.copy()
df_consumption_growth1["nu"] = lst_nu
df_consumption_growth1

#lng = 0.02 + 0.02*epsilon + nu
df_consumption_growth2 = df_consumption_growth1.copy()
df_consumption_growth2["lng"] = (consumption_growth_intercept
                                +
                                consumption_growth_slope*df_consumption_gr
                                +
                                df_consumption_growth2["nu"]
                                )

df_consumption_growth3 = df_consumption_growth2.copy()
df_consumption_growth3["g"] = np.exp(df_consumption_growth3["lng"])

#parameters settings
gamma_start = 0
gamma_end = 4.1
gamma_increment = 0.1

M_coefficient = 0.99

#M = 0.99*g^(-gamma)
lst_cols = []

df_consumption_growth4 = df_consumption_growth3.copy()

for gamma in np.arange(gamma_start, gamma_end, gamma_increment):
    col = "M_" + str(round(gamma, 1))
    lst_cols.append(col)
    df_consumption_growth4[col] = M_coefficient*np.exp((1-df_consumption_gr

df_M = df_consumption_growth4[lst_cols].apply(lambda x: [np.mean(x), np.stc
df_M.columns = ["mu_M", "xigma_M", "xigma_M/mu_M"]
df_M

```

Out[6]:

	mu_M	xigma_M	xigma_M/mu_M
M_0.0	0.990000	1.713074e-13	1.730378e-13
M_0.1	0.988852	6.198524e-03	6.268406e-03
M_0.2	0.987743	1.260288e-02	1.275927e-02
M_0.3	0.986677	1.922218e-02	1.948174e-02
M_0.4	0.985652	2.606585e-02	2.644528e-02
M_0.5	0.984672	3.314373e-02	3.365965e-02
M_0.6	0.983738	4.046599e-02	4.113495e-02
M_0.7	0.982850	4.804323e-02	4.888157e-02
M_0.8	0.982010	5.588645e-02	5.691025e-02
M_0.9	0.981221	6.400707e-02	6.523208e-02
M_1.0	0.980483	7.241694e-02	7.385845e-02
M_1.1	0.979798	8.112841e-02	8.280114e-02
M_1.2	0.979169	9.015427e-02	9.207225e-02
M_1.3	0.978596	9.950781e-02	1.016842e-01
M_1.4	0.978082	1.092029e-01	1.116500e-01
M_1.5	0.977630	1.192538e-01	1.219826e-01
M_1.6	0.977240	1.296754e-01	1.326956e-01
M_1.7	0.976915	1.404834e-01	1.438031e-01
M_1.8	0.976658	1.516937e-01	1.553192e-01
M_1.9	0.976470	1.633230e-01	1.672586e-01
M_2.0	0.976354	1.753888e-01	1.796365e-01
M_2.1	0.976313	1.879091e-01	1.924680e-01
M_2.2	0.976350	2.009027e-01	2.057691e-01
M_2.3	0.976466	2.143890e-01	2.195559e-01
M_2.4	0.976666	2.283883e-01	2.338448e-01
M_2.5	0.976951	2.429216e-01	2.486528e-01
M_2.6	0.977326	2.580109e-01	2.639968e-01
M_2.7	0.977792	2.736787e-01	2.798946e-01
M_2.8	0.978354	2.899489e-01	2.963639e-01
M_2.9	0.979015	3.068459e-01	3.134230e-01
M_3.0	0.979779	3.243951e-01	3.310901e-01
M_3.1	0.980649	3.426233e-01	3.493842e-01
M_3.2	0.981630	3.615578e-01	3.683241e-01
M_3.3	0.982724	3.812274e-01	3.879291e-01
M_3.4	0.983938	4.016618e-01	4.082187e-01
M_3.5	0.985274	4.228920e-01	4.292125e-01
M_3.6	0.986738	4.449502e-01	4.509302e-01
M_3.7	0.988335	4.678697e-01	4.733919e-01

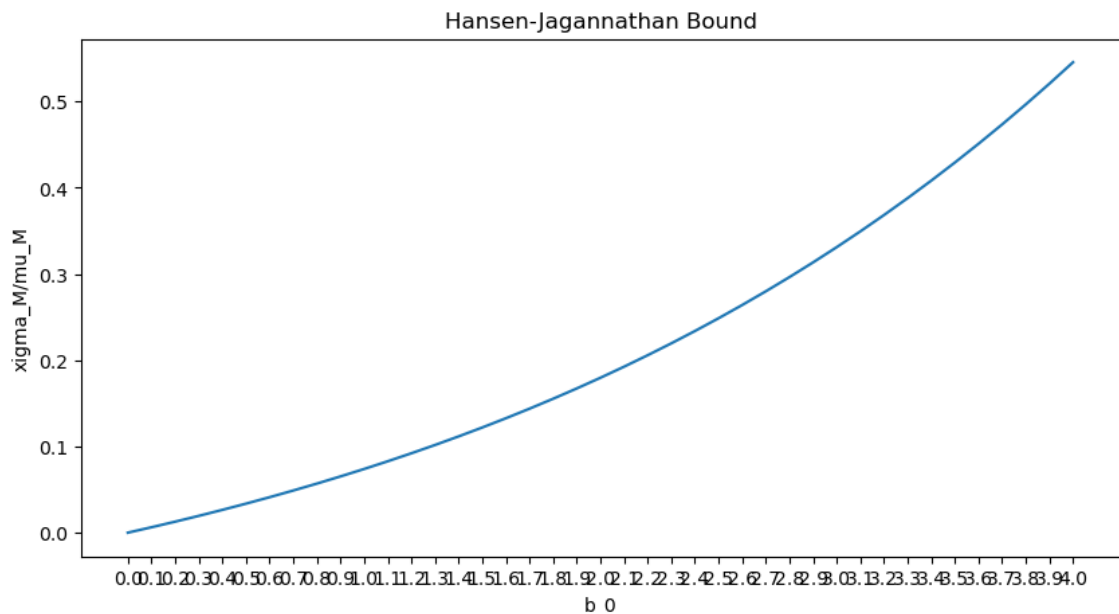
	mu_M	xigma_M	xigma_M/mu_M
M_3.8	0.990068	4.916854e-01	4.966176e-01
M_3.9	0.991945	5.164335e-01	5.206273e-01
M_4.0	0.993969	5.421514e-01	5.454412e-01

```
In [8]: df_M1 = df_M.copy()
df_M1["gamma"] = [round(gamma, 1) for gamma in np.arange(gamma_start, gamma_end, gamma_increment)]

plt.figure(figsize=[10,5])
plt.plot(df_M1["gamma"], df_M1["xigma_M/mu_M"])
plt.xticks(np.arange(gamma_start, gamma_end, gamma_increment))

plt.xlabel("b_0")
plt.ylabel("xigma_M/mu_M")
plt.title("Hansen-Jagannathan Bound")

plt.show()
```



Explain (in words) how to find the smallest value of b_0 for which the Hansen–Jagannathan bound is satisfied, and report the result for your data set.

need to know sharpe ratio first, and let say sharpe ratio = 0.4, then find samllest b_0 that make $xigma_M/mu_M$ larger than sharpe ratio

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In [10]: smallest_gamma = np.min(df_M1[df_M1["xigma_M/mu_M"]>0.4]["gamma"])  
print(smallest_gamma)
```

3.4