

Since we changed our variables from n_H to s , I thought it'd be better to replot all the older graphs first.

So, first I plotted

- ① s vs pdf
- ② s vs $\log(\text{pdf})$
- ③ $\log(\lambda_f)$ vs s
- ④ s vs X_{H_2}

I have detailed the algorithm & the verification also.
So far everything seems as expected.

$$(I) M = 5$$

$$\sigma_s = \sqrt{\log(1 + 0.3^2 M^2)}$$

$$\bar{s} = -\frac{1}{2} \sigma_s^2$$

$$s_{\min} = -3\sigma_s + \bar{s}$$

$$s_{\max} = +3\sigma_s + \bar{s}$$

$$ds = (s_{\max} - s_{\min})/1000$$

$$\bar{n}_H = 10,000$$

~~for i = 0 to 1000~~

for ($i \rightarrow 0$ to 1000)

$$s[i] = s_{\min} + i \cdot ds$$

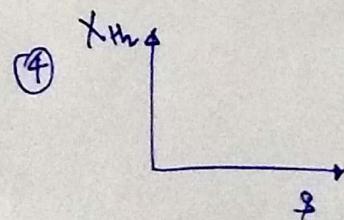
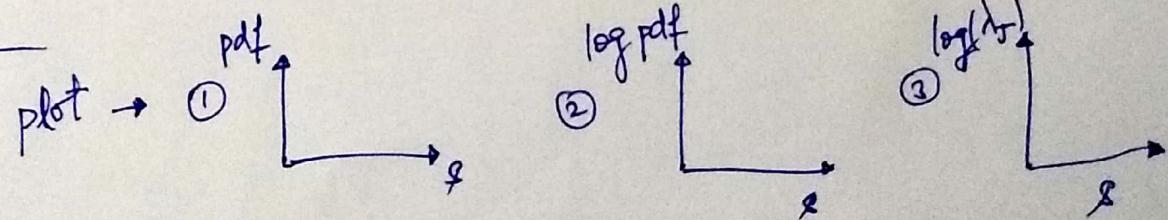
$$\text{pdf} = \frac{1}{\sqrt{2\pi\sigma_s^2}} \cdot \exp\left[-\frac{1}{2}\left(\frac{s - \bar{s}}{\sigma_s}\right)^2\right]$$

$$n_H = \bar{n}_H \cdot \exp(-s)$$

$$\lambda_J = \frac{\sqrt{\frac{K_b \cdot T_{\text{mean}}}{m_p}}}{\sqrt{4\pi G n_H \cdot m_p}}$$

$$\exp(-\tau) = \exp(-k^* n_H * \lambda_J)$$

$$X_{H_2} = \left[\frac{1}{2 + \left(\frac{DC \cdot \exp(-\tau) \cdot \text{rad-out}}{CC \cdot Z \cdot n_H} \right)} \right]$$



$$T = 10K, K_b = 1.3806 \times 10^{-16} \text{ erg/K}$$

$$m_p = 1.6726 \times 10^{-24} \text{ g}, G = 6.674 \times 10^{-8} \text{ dyne cm}^2/\text{g}^2$$

$$c_s = \sqrt{\frac{k_b T}{m_p}} = 28730.5 \text{ cm/s} \quad | \quad n_H = \bar{n}_H \cdot \exp(z)$$

$$\bar{n}_H = 10,000$$

$$(S)_{\min} = -3.8463038$$

$$(S)_{\max} = 2.661134902$$

$$(n_H)_{\min} = 213.585350$$

$$(n_H)_{\max} = 143125 \cdot 232046$$

~~From Analytical:~~

$$(\lambda_J)_{\min} = \frac{c_s}{\sqrt{4\pi G m_p (n_H)_{\max}}} = 6.41 \times 10^{16}$$

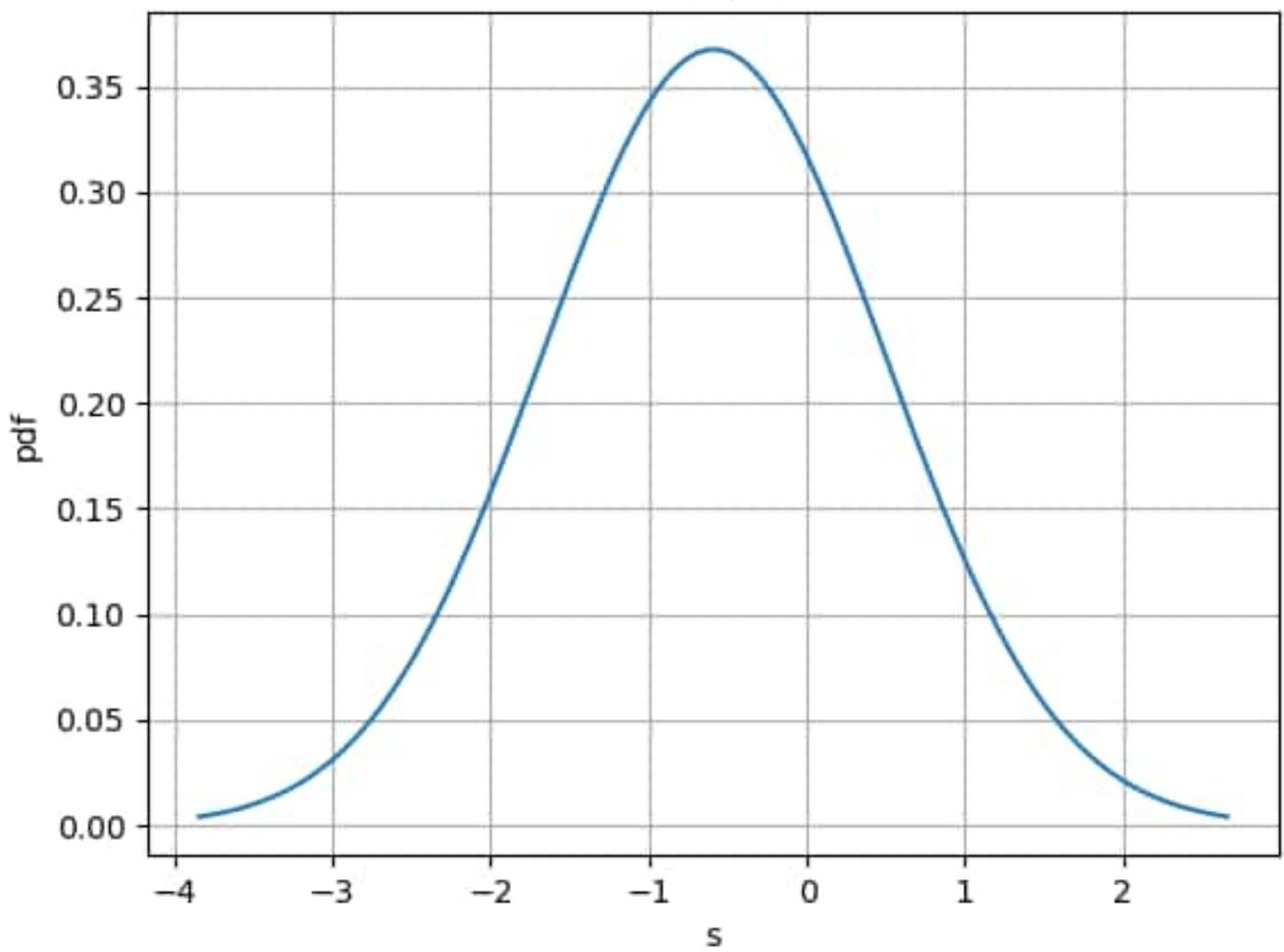
$$(\lambda_J)_{\max} = \frac{c_s}{\sqrt{4\pi G m_p (n_H)_{\min}}} = 1.659 \times 10^{18}$$

From Code:

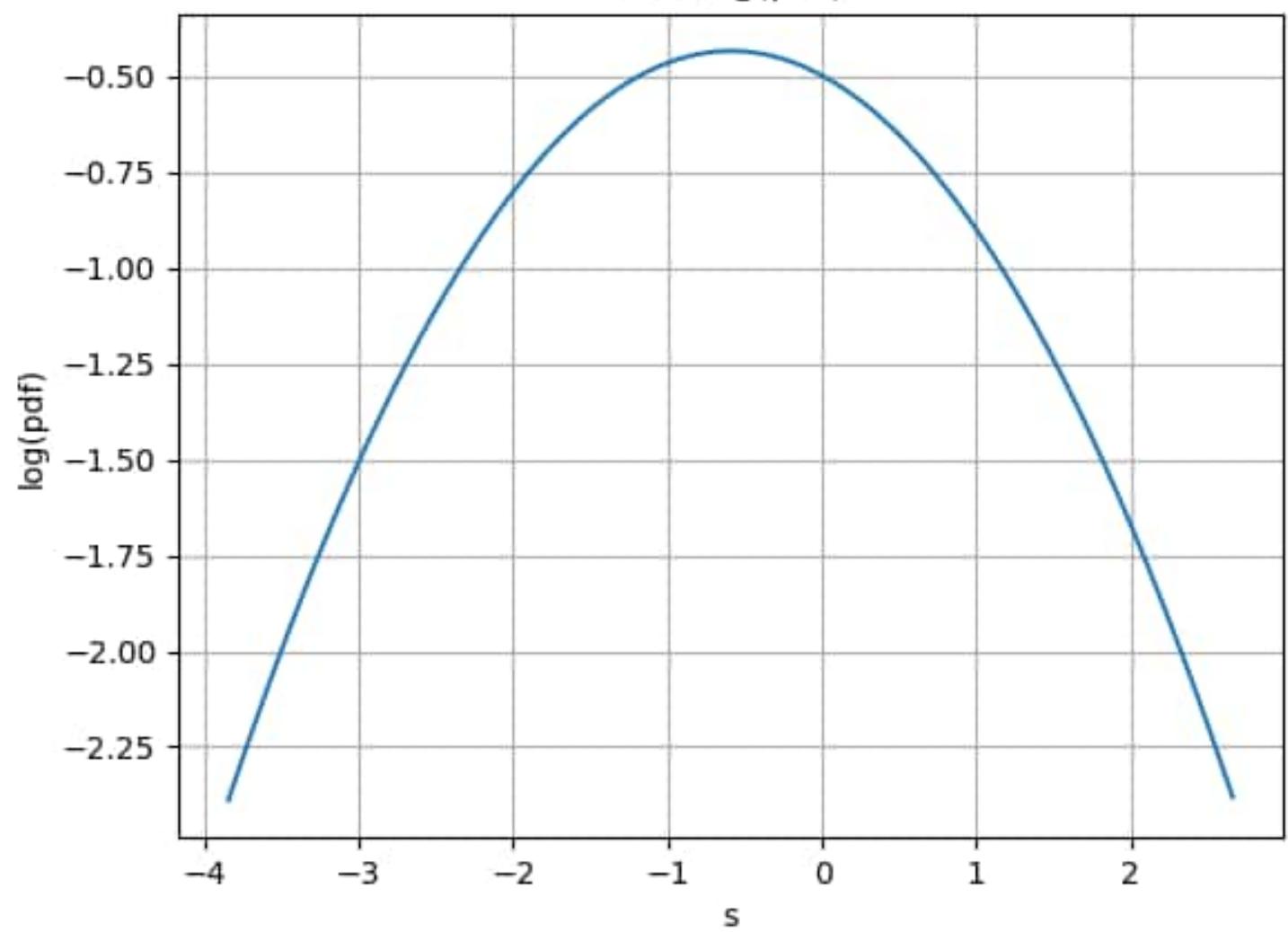
$$(\lambda_J)_{\min} = 6.4118 \times 10^{15} \quad | \quad \log(\lambda_{J,\min}) = 16.81$$

$$(\lambda_J)_{\max} = 1.6598 \times 10^{18} \quad | \quad \log(\lambda_{J,\max}) = \cancel{18.22} 18.22$$

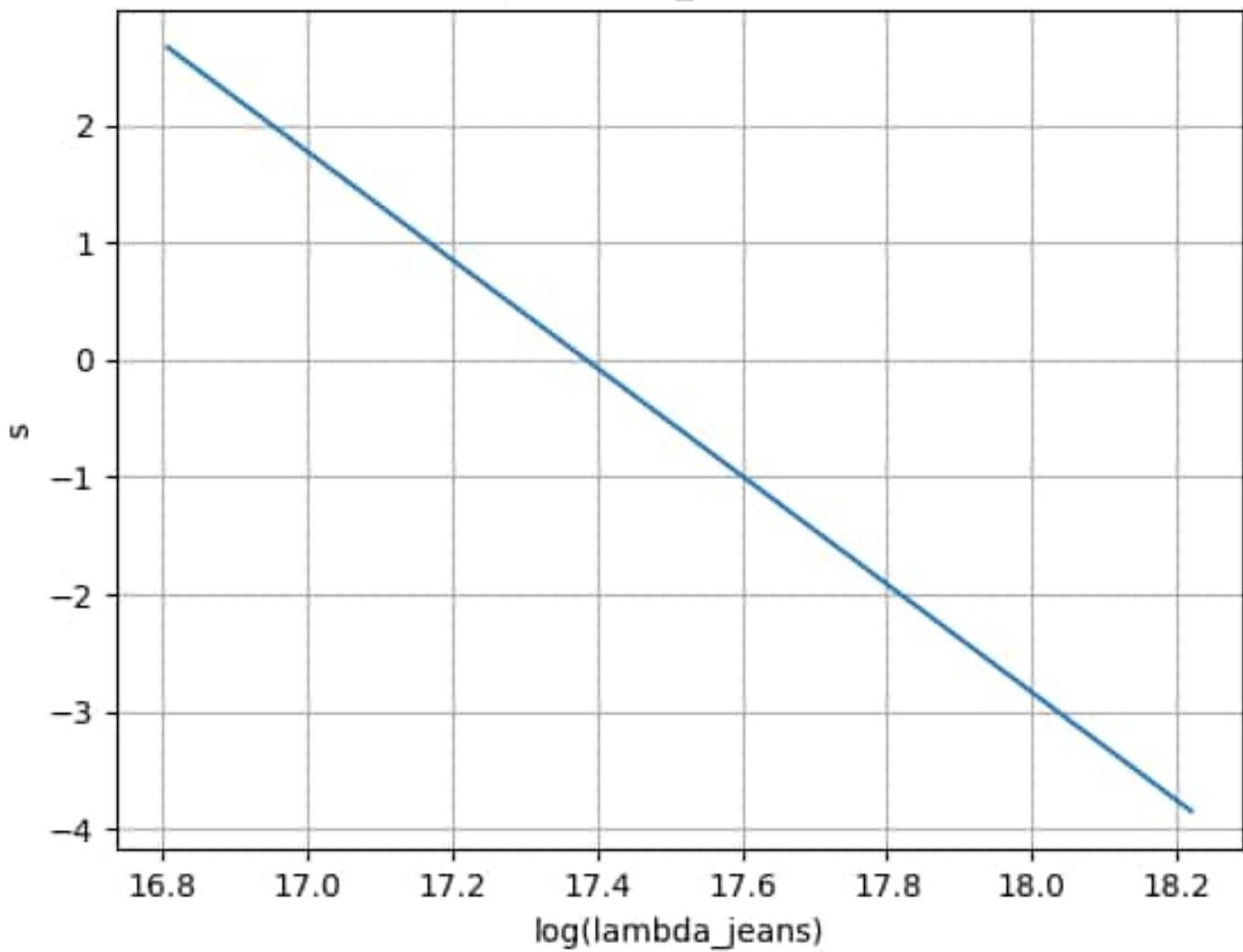
s vs pdf



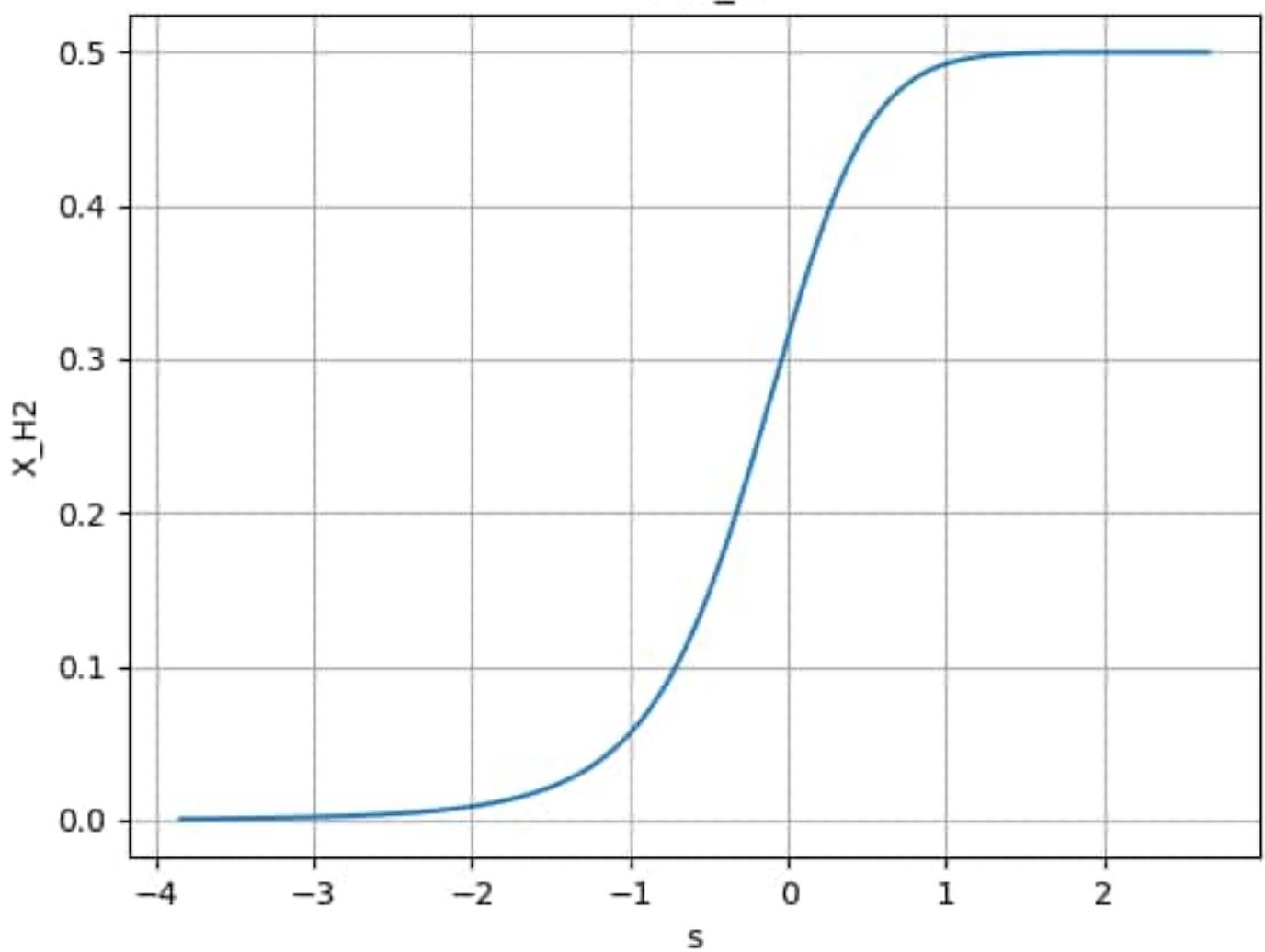
s vs log(pdf)



$\log(\lambda_{\text{jeans}})$ vs s



s vs X_H2



Then, I tried to integrate to find $\int e^s P(s) ds$ &
 $\int e^s P(s) ds \cdot \bar{X}_{H_2}$.

$$\text{But, } \text{tot-} \bar{n}_H = \int e^s P(s) ds \neq 1$$

so, obviously $\bar{X}_{H_2} = \int e^s P(s) ds \cdot \bar{X}_{H_2}$ wasn't as expected.

Just to see $\bar{X}_{H_2} \rightarrow$ I plotted s vs \bar{X}_{H_2} .

(II) for ($i \rightarrow 0$ to 1000)

$$\left[\begin{array}{l} \text{tot-}\bar{n}_H = \text{tot-}\bar{n}_H + [\exp(s) \cdot \text{pdf} \cdot ds] \\ \bar{X}_{H_2} = \bar{X}_{H_2} + [\text{tot-}\bar{n}_H \cdot X_{H_2}] = N \end{array} \right]$$

plotting : (just to see the data p.t.s)

$$\textcircled{1} \quad \bar{X}_{H_2} \uparrow = x_{NM2} ; \quad \bar{s} + 2 \Rightarrow \text{tot-}\bar{n}_H \neq 1$$

$$\text{tot-}\bar{n}_H = \frac{(N_{NM2} - x_{NM2})}{20} = 2.17 \times 10^{-4}$$

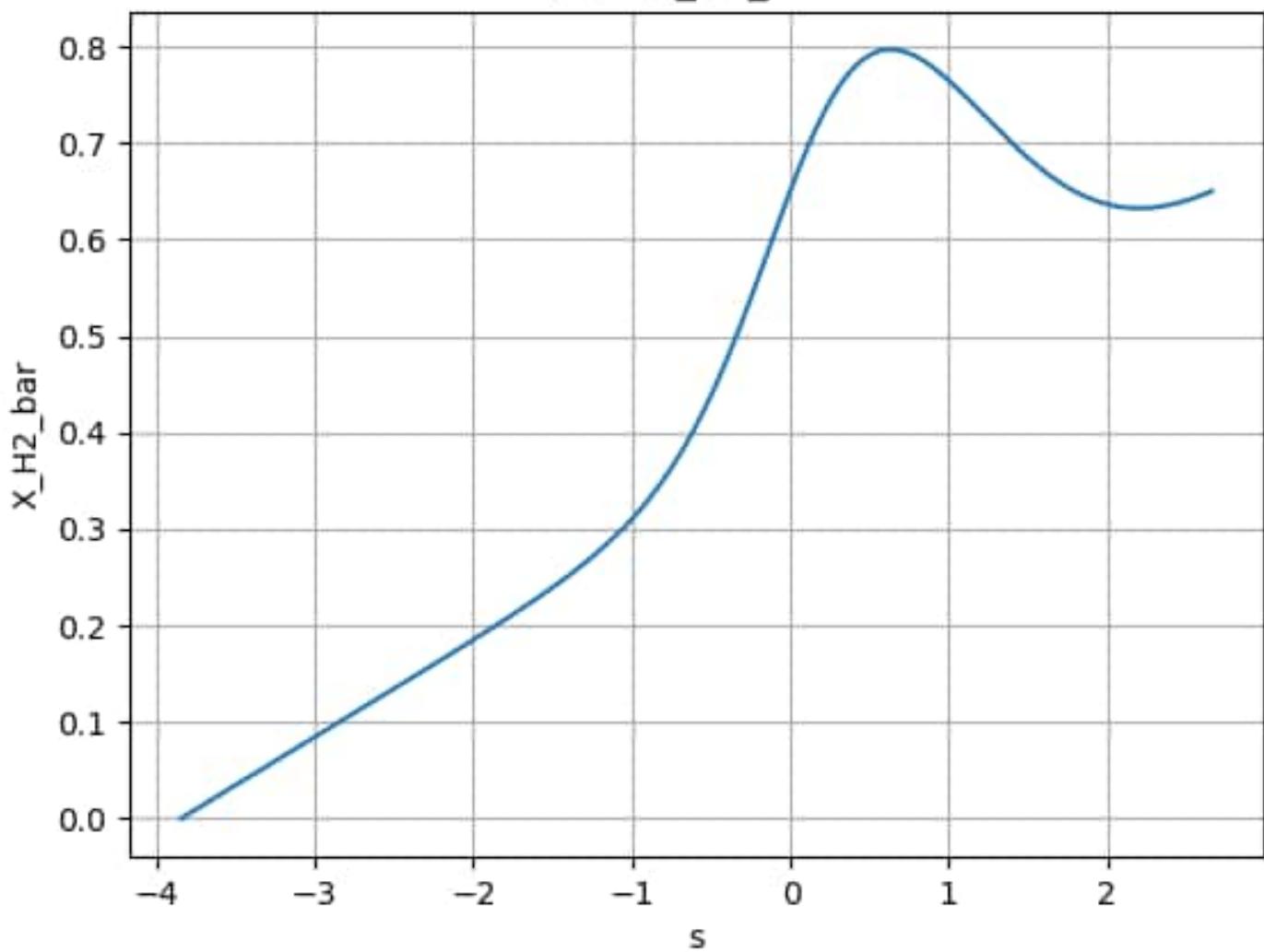
$$(\text{tot-}\bar{n}_H)_{\min} = 5.68 \times 10^{-4}$$

$$(\text{tot-}\bar{n}_H)_{\max} = 2.17$$

$$(\text{total of } 0 \leftarrow i)$$

$$20 \cdot i + N_{NM2} = [i]_2$$

s vs X_H2_bar



I skipped this problem for the time being, because I wanted to plot \bar{X}_{H_2} . So, I assumed that $\text{tot-}\bar{n}_H = 0.999686855$ so that I can see whether the problem was only in $\text{tot-}\bar{n}_H$. And plotted λ vs \bar{X}_{H_2} , and it was close to what was expected. So the problem is that $\text{tot-}\bar{n}_H \neq 1$. $0.999686855 \rightarrow$ value I got when n_H was the variable (last week).

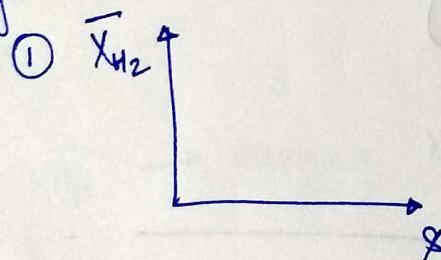
(III) setting $\text{tot-}\bar{n}_H = \cancel{0.999} 0.9996868549$

for ($i \rightarrow 0$ to 1000)

$$\text{tot-}\bar{n}_H = \text{tot-}\bar{n}_H + [\exp(x) \cdot \text{pdf} \cdot dx]$$

$$\bar{X}_{H_2} = \bar{X}_{H_2} + [0.9996868549 \cdot X_{H_2}]$$

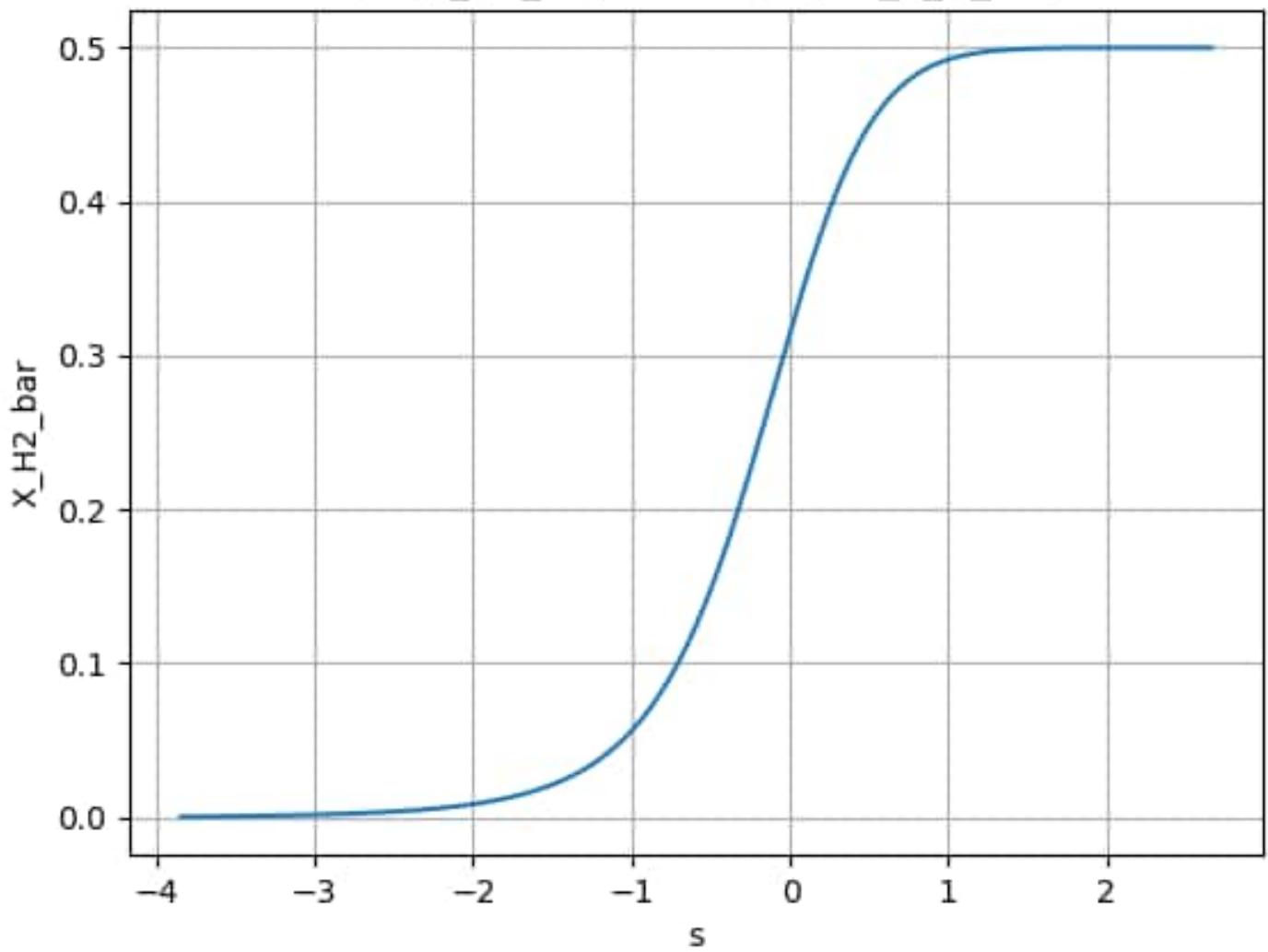
plotting:



→ almost as expected

→ problem in tot-\bar{n}_H

s vs X_H2_bar - assumed tot_n_H_bar



For my next step, I assumed $\text{tot-}\bar{n}_H = 0.999686855$

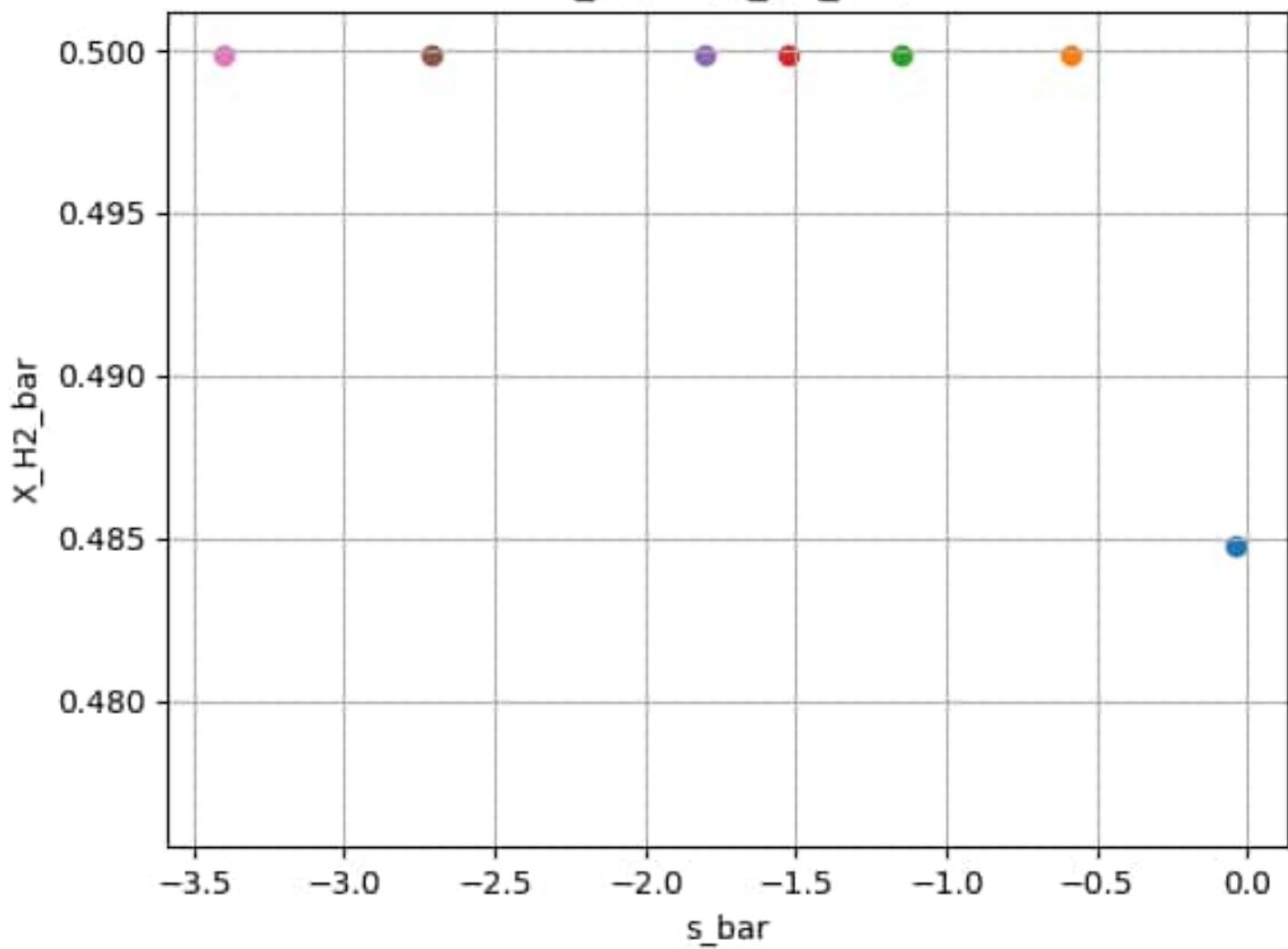
Now I took mach no. M in the form of an array.
I varied ~~this~~ M so that $O_s, \bar{s}, s_{\min}, s_{\max}$ & ds
would vary, so that I could plot \bar{s} vs \bar{X}_{H_2} . Each
point is a different value of \bar{X}_{H_2} for different \bar{s} .

(IV) M-array = [1, 5, 10, 15, 20, 50, 100]
 for (m → ~~Marray~~ → 0 to len(M-array))
 {
 M = M-array[m]
 $\sigma_s = \sqrt{\log(1 + 0.3^2 M^2)}$
 $\bar{s} = -\frac{1}{2} \sigma_s^2$
 smin = $-3\sigma_s + \bar{s}$; smax = $+3\sigma_s + \bar{s}$
 ds = (smax - smin) / 1000
 $\bar{n}_H = 10,000$
 for (i → 0 to 1000)
 {
 s[i] = smin + i * ds
~~n_H[i] = $\bar{n}_H \cdot \exp(s[i])$~~
 $X_{H_2}[i] = \text{_____}$
 pdf[i] = _____
 tot- $\bar{n}_H[i] = \cancel{\exp(s[i])} \cdot \exp(s[i]) \cdot pdf[i] \cdot ds$
 $\bar{X}_{H_2}[i] = 0.9996868549 \cdot X_{H_2}[i]$
 plot (\bar{s} & ~~\bar{X}_{H_2}~~)
 }
 }
 saving the plot

$$\text{tot-}\bar{n}_H \neq 1$$

varying M → varying \bar{s}
 ↳ then plotting \bar{s} vs \bar{X}_{H_2}

$s_{\bar{b}} \text{ vs } X_{H_2\bar{b}}$



Now, again, I assumed tot- $\bar{n}_H = 0.999686855$

This time, I fixed $M=5$, & varied \bar{n}_H .

Varying \bar{n}_H would change n_H (because $n_H = \bar{n}_H \cdot \exp(s)$), which will change X_{H_2} & \bar{X}_{H_2} .

Then I plotted \bar{n}_H vs \bar{X}_{H_2} .

This plot is very similar to the plot of s vs X_{H_2} .

Each pt. is a different value for X_{H_2} as \bar{n}_H is varied.

First I varied \bar{n}_H as : $[10, 10^2, 10^3, 10^4, 10^5, 10^6]$

But then I noticed there were big gaps in the curve, so I added more data-points : $[10, 10^2, 250, 500, 750, 10^3, 2500, 5000, 10^4, 10^5, 10^6]$

(I) The array ~~10, 500, 2000, 1000, 100, 250, 500, 750~~

$$\overline{n}_H\text{-array} = [10, 10^2, 10^3, \sqrt{10^4}, 10^5, 10^6]$$

for ($m \rightarrow 0$ to $\text{len}(\overline{n_H\text{-array}})$)

$$\hat{M} = 5$$

$\sigma_s, \bar{s}, s_{\min}, s_{\max}, ds \rightarrow \text{fixed}$

$$\overline{n}_H = \overline{n}_H\text{-array}[m]$$

for ($i \rightarrow 0$ to 1000)

$s[i] = -$

$$n_h[i] = \bar{n}_h \cdot \exp(g[i])$$

$$X_{H_2}[i] = -$$

pdf[i] = -

$$\text{tot_}\bar{n}_H[i] = -$$

$$\bar{X}_{H_2}[i] = 0.99 - x_{H_2}[i]$$

plot ($\log(n_H)$ & X_{H_2})

sawing the plot

$$tot - \bar{n}_H \neq 1$$

varying $\overline{n_H}$ \rightarrow varying n_H \rightarrow varying X_{H_2}

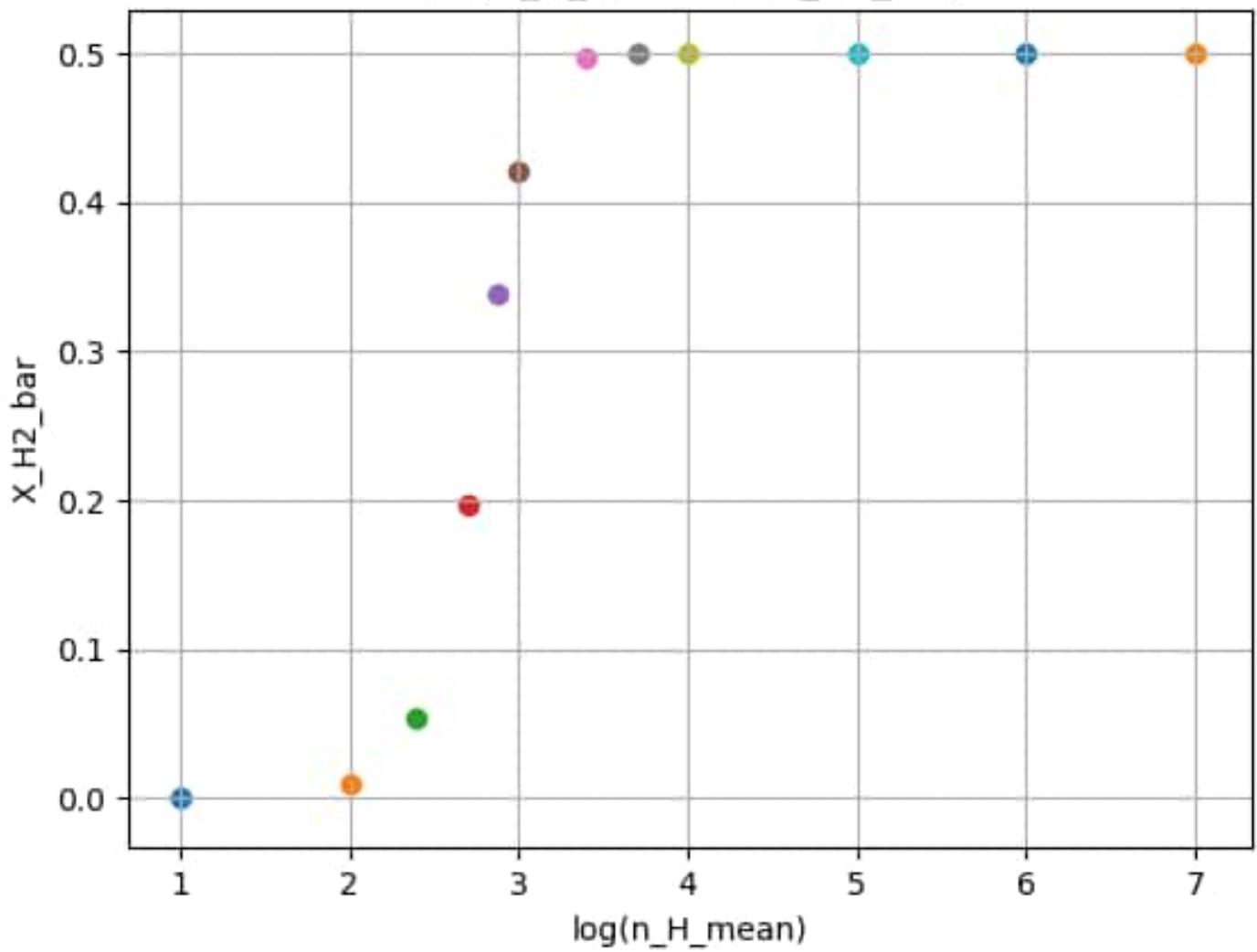
↓
varying $\overline{X_{H_2}}$

1

then plotting

$\log(\bar{n}_H)$ vs \bar{X}_{H_2}

$\log(n_{\text{H}} \text{mean})$ vs $X_{\text{H}_2 \text{bar}}$



If the plot of \overline{n}_H vs \overline{X}_{H_2} is correct, then the only problem is that $\text{tot-}\overline{n}_H \neq 1$.