

# Math 254A Lecture 13 Notes

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## 1 Observing Macroscopic Quantities From Microscopic States

### 1.1 Recap

We have a phase space  $(M, \lambda)$  which is a  $\sigma$  finite but not finite measure space. The energy of one particle is  $\varphi : M \rightarrow [0, \infty)$ , where  $\min \varphi = \text{ess min } \varphi = 0$ . Then we know that

$$\begin{aligned} \lambda^{\times n} \left( \left\{ (p_1, \dots, p_n) \in M^n : \frac{1}{n} \Phi_n(p_1, \dots, p_n) := \frac{1}{n} \sum_{i=1}^n \varphi(p_i) \in I \right\} \right) \\ = \exp \left( n \cdot \sup_{x \in I} s(x) + o(n) \right), \end{aligned}$$

where

$$s(x) = \inf_{\beta > 0} \{s^*(\beta) + \beta x\}.$$

We also have the Fenchel-Legendre transform

$$s^*(\beta) = \log \int e^{-\beta \varphi}.$$

$\beta$  achieves equality in the definition of  $s$

$$\begin{aligned} &\iff s \text{ has a tangent of slope } \beta \text{ at } x \\ &\iff D_+ s(x) \leq \beta \leq D_- s(x) \\ &\iff s^*(\beta + (-s(x))) = -\beta x \\ &\iff D_- s^*(\beta) \leq -x \leq D_+ s^*(\beta) \\ &\iff s^* \text{ has a tangent of slope } -x \text{ at } \beta. \end{aligned}$$

Using  $s^*$ , we can prove:

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$$s^*(\beta) \rightarrow \begin{cases} \log \lambda(\{\varphi = 0\}) & \beta \rightarrow \infty \\ \infty & \beta \downarrow 0. \end{cases}$$

- $s^*$  is strictly decreasing and strictly convex.

- $s^*$  is differentiable on  $(0, \infty)$ .

•

$$s(x) \rightarrow \begin{cases} \log \lambda(\{\varphi = 0\}) & x \downarrow 0 \\ \infty & x \rightarrow \infty. \end{cases}$$

- $s$  is strictly increasing and strictly concave.

- $s$  is differentiable on  $(0, \infty)$ .

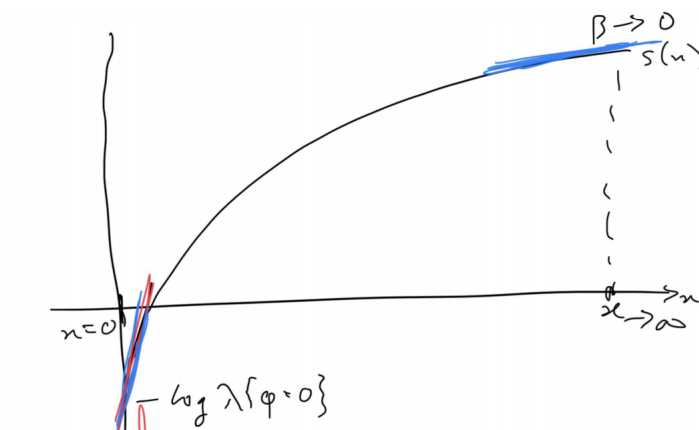
## 1.2 Behavior of $s'$

Let's analyze the behavior of  $s'$ :

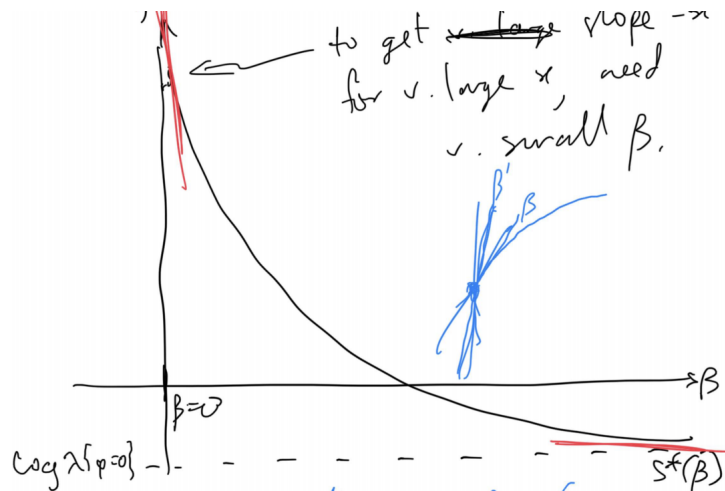
**Proposition 1.1.**

$$s'(x) \rightarrow \begin{cases} 0 & x \rightarrow \infty \\ \infty & x \rightarrow 0. \end{cases}$$

Instead of a formal proof, here are some pictures. Look at the possible slopes we can get for points on the graph of  $s$  and how they correspond to slopes for points on the graph for  $s^*$ .



To get slope  $-x$  for very large  $x$  in the graph of  $s^*$ , we need very small  $\beta$ .



### 1.3 Observing macroscopic quantities from microscopic states

Now imagine we are looking at some other macroscopic observable quantity of the microscopic state  $(p_1, \dots, p_n) \in M^n$ . We will study functions for the form

$$\Psi_n(p_1, \dots, p_n) = \sum_{i=1}^n \psi(p_i).$$

If  $M = \mathbb{R}^3 \times \mathbb{R}^3$ , we could take  $\psi(r, p) = \mathbb{1}_D(r)$ , which indicates whether a particle is in  $D$  or not in  $D$ ; then  $\Psi_n$  would be the total number of particles in  $D$ .

We need some regularity. A simple sufficient condition is that  $\psi$  is bounded. A weaker but still sufficient condition is that for every  $\beta > 0$ , there is an  $\varepsilon > 0$  such that  $\int e^{-\beta\varphi} e^{-\gamma\psi} d\lambda < \infty$  for all  $\gamma \in (-\varepsilon, \varepsilon)$ .

Let's assume  $\psi$  is bounded, and we'll ask about the distribution of  $\Psi_n$  on the approximate level set  $\{\frac{1}{n}\Phi_n \in I\}$ , where  $I$  is a small interval. We need to compare  $\lambda^{\times n}(\{\frac{1}{n}\Phi_n \in I\})$  and  $\lambda^{\times n}(\{\frac{1}{n}\Phi_n \in I, \frac{1}{n}\Psi_n \in J\})$ . We use the generalized type-counting machinery with  $\mathbb{R}^2$  to get an asymptotic for this:

$$\begin{aligned} \lambda^{\times n} \left( \left\{ \frac{1}{n} \Phi_n \in I, \frac{1}{n} \Psi_n \in J \right\} \right) &= \lambda^{\times n} \left( \left\{ (p_1, \dots, p_n) \in M^n : \frac{1}{n} \sum_{i=1}^n (\varphi(p_i), \psi(p_i)) \in I \times J \right\} \right) \\ &= \exp \left( n \cdot \sup_{(x,y) \in I \times J} \widetilde{s}(x, y) + o(n) \right), \end{aligned}$$

where  $\tilde{s}(x, y) : \mathbb{R}^2 \rightarrow [-\infty, \infty)$  is an upper semicontinuous, concave function with

$$\tilde{s}(x, y) = \inf_{\beta, \gamma} \{ \tilde{s}^*(\beta, \gamma) + \beta x + \gamma y \}.$$

and Fenchel-Legendre transform

$$\tilde{s}^*(\beta, \gamma) = \log \int e^{-\beta\varphi} e^{-\gamma\psi} d\lambda.$$

Here, we assume  $\psi$  is bounded,  $|\psi| \leq M$ , so

$$\tilde{s}^*(\beta, \gamma) = \begin{cases} \infty & \beta = 0 \\ < \infty & \beta > 0. \end{cases}$$

Here,  $\tilde{s}(x, y) \leq s(x)$  for all  $y \in \mathbb{R}$ . We want to find a  $y_0$  such that  $\tilde{s}(x, y_0) = s(x)$  and  $\tilde{s}(x, y) < s(x)$  for any other  $y$ . This will tell us that conditioned on  $\Phi$  being  $x$ , we are likely to have  $\Psi$  be  $y_0$  and not likely to have any other  $y$ . We have

$$s(x) = \inf_{\beta > 0} \left\{ \log \int e^{-\beta\varphi} d\lambda + \beta x \right\},$$

which is greater than or equal to

$$\tilde{s}(x, y) = \inf_{\beta > 0, \gamma \in \mathbb{R}} \left\{ \log \int e^{-\beta\varphi} e^{-\gamma\psi} d\lambda + \beta x + \gamma y \right\}.$$

**Lemma 1.1.**  $\tilde{s}(x, y_0) = s(x)$  and  $\tilde{s}(x, y) < s(x)$  for any other  $y$ , where

$$y_0 = \int \psi e^{-\beta\varphi} d\lambda = \langle \psi, \mu_\beta \rangle$$

and

$$d\mu_\beta(p) = \frac{e^{-\beta\varphi(p)} d\lambda(p)}{\int e^{-\beta\varphi} d\lambda}$$

is the **Gibbs measure** obtained from  $\lambda, \varphi, \beta$ .

*Proof.* First,  $s$  is differentiable, so for every  $x > 0$ , there is a unique  $\beta > 0$  such that  $s(x) = \log \int e^{-\beta\varphi} d\lambda + \beta x$ . To achieve  $\tilde{s}(x, y_0) = s(x)$ , we must have that the function  $\gamma \mapsto \log \int e^{-\beta\varphi} e^{-\gamma\psi} d\lambda + \beta x + \gamma y_0$  achieves its minimum uniquely at  $\gamma = 0$ . This function of  $\gamma$  is convex (by Hölder), strictly convex if  $\psi$  is not a.s. constant, and differentiable. Assuming  $\psi$  is not a.s. constant, we need  $y_0$  such that

$$\frac{\partial}{\partial \gamma} \left\{ \log \int e^{-\beta\varphi} e^{-\gamma\psi} d\lambda + \beta x + \gamma y_0 \right\} = 0$$

at  $\gamma = 0$ . This is the derivative of the log of the moment generating function. Differentiate under the integral to get

$$\frac{\partial}{\partial \gamma} \log \int e^{-\beta\varphi} e^{-\gamma\psi} d\lambda = \frac{\int -\psi e^{-\beta\varphi} e^{-\gamma\psi} d\lambda}{\int e^{-\beta\varphi} e^{-\gamma\psi} d\lambda} \Big|_{\gamma=0} = -\langle \psi, \mu_\beta \rangle.$$

So  $\frac{\partial}{\partial \gamma} [\dots] |_{\gamma=0} = -\langle \psi, \mu_\beta \rangle + y_0$ , and this equals 0 iff  $y_0 = \langle \psi, \mu_\beta \rangle$ . □

**Corollary 1.1.**

$$\lambda^{\times n} \left( \left\{ \left| \frac{1}{n} \Psi_n - \langle \psi, \mu_\beta \rangle \right| > \varepsilon \right\} \mid \left\{ \frac{1}{n} \Phi_n \in I \right\} \right) \leq e^{-c \cdot n + o(n)},$$

where  $c$  is a constant,  $I$  is a short enough interval containing  $x$ , and we are using conditional probability notation.

**Remark 1.1.** Given  $\frac{1}{n} \Phi_n \approx x$ , we found that

$$\begin{aligned} \Psi_n &\approx n(\text{its average over } \{\frac{1}{n} \Phi_n \approx n\}^1) \\ &\approx n \langle \psi, \mu_\beta \rangle \\ &= \langle \psi(p_1) + \cdots + \psi(p_n), \mu_\beta^{\times n} \rangle \\ &= \int \Psi_n d\mu_{\beta,n}, \end{aligned}$$

where

$$d\mu_{\beta,n}(p_1, \dots, p_n) = \frac{e^{-\beta \Psi_n(p_1, \dots, p_n)} d\lambda^{\times n}(p)}{\int e^{-\beta \Psi_n} d\lambda^{\times n}} = \mu_\beta \times \cdots \times \mu_\beta$$

is called the **canonical ensemble measure**.

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<sup>1</sup>This is called the **microcanonical ensemble**.