

* connection between

$$H_S(\mathcal{X}) \quad \text{and} \quad H(\mathcal{X})$$

- depends on δ
- can be "weak"

when $\theta = 0.25$, $H(\mathcal{X}) = 0.81$

and $\delta = 0.25$, $H_S(\mathcal{X}) = 0$

* Connection between

$$\boxed{H_S(\mathcal{X}^N)} \quad \text{and} \quad H(\mathcal{X})^N = N \cdot H(\mathcal{X})$$

strong(er) for large(r) N

informally: $\frac{1}{N} H_S(\mathcal{X}^N) \approx H(\mathcal{X})$

* intuition behind the connection

$$\begin{array}{llll} \text{many iid} & \text{sample mean} & \approx & \text{expectation} \\ \text{sample} & & & \\ \mathbf{a} = (a_{i1}, a_{i2}, \dots, a_{iN}) & \frac{r_{i1} + r_{i2} + \dots + r_{iN}}{N} & \approx & \sum_i p_i r_i \quad \text{for r.v. } r_i \\ & & & \parallel \\ & \frac{1}{N} \sum_{n=1}^N \log \frac{1}{p_{i_n}} & \approx & H(\mathcal{X}) \quad \log \frac{1}{p_i} \\ & \parallel & & \end{array}$$

$$\therefore \frac{1}{N} \log \frac{1}{P(\underline{a})}$$

$$\text{many } P(\underline{a}) \approx 2^{-NH}$$

$$\underbrace{\left\{ 2^{NH} P(\underline{a}) \right\}}_{\text{typical set}} \approx \text{prob. } 1$$

$$NH \approx H_S(\mathcal{X}^N)$$

*

$$Pr \left(\left(\frac{1}{N} \sum_{n=1}^N u_n - E(u) \right)^2 \geq \alpha \right) \leq \frac{\sigma^2}{\alpha N}$$

weak law of
large numbers

a random variable w/ mean $E(u)$

2 var. $\frac{\sigma^2}{N}$ (for per-un variance σ^2)



$$\Pr\left(\left(r - E(r)\right)^2 \geq \alpha\right) \leq \frac{\sigma_r^2}{\alpha} \quad (\text{Chebyshev's 2nd Ineq.})$$

Ineg.)

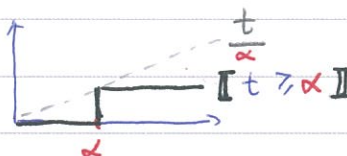


$$\Pr(t \geq \lambda) \leq \frac{E(t)}{\lambda} \quad (\text{Chebyshev's 1st Ineq.})$$

(Chebyshev's 1st Ineq.)

for $t \geq 0$

$$\text{pf. } \sum p(t) \mathbb{I} t \geq \alpha \mathbb{I} \leq \sum p(t) \frac{t}{\alpha}$$



* Shannon's 1st Theorem

Given ϵ , $0 < \epsilon < 1$, there exists N_0 s.t.

$$\forall N \geq N_0, \quad \left| \frac{1}{N} H(\bar{x}^N) - H(\bar{x}) \right| < \epsilon$$

$$(a) \quad \frac{1}{N} H_S(\Sigma^N) \leq H + \epsilon$$

find a "small" subset $U \subseteq \mathbb{X}^N$

$$U: \{ \underline{z}: p(\underline{z}) > 2^{-(H+\epsilon) \cdot N} \}$$



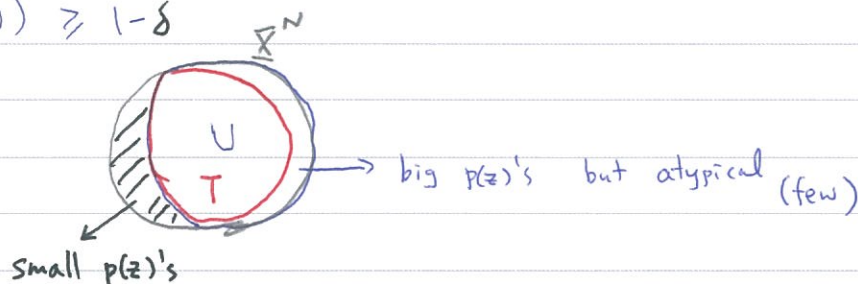
$$|U| < 2^{(H+E) \cdot N}$$

show $H_\delta(\mathbb{R}^N) \leq \log U < (H + \epsilon)$

by showing $\Pr(U) \geq 1 - \delta$

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* $\Pr(U) \geq 1 - \delta$



T_ϵ : typical set at width ϵ

$$2^{-(H+\epsilon)N} < p(z) < 2^{-(H-\epsilon)N}$$

$\Updownarrow$

$$-(H+\epsilon)N < \sum \log p_{i_n} < -(H-\epsilon)N$$

$\Updownarrow$

$$\epsilon > \left(\frac{1}{N} \sum \log \frac{1}{p_{i_n}} - H \right) > -\epsilon$$

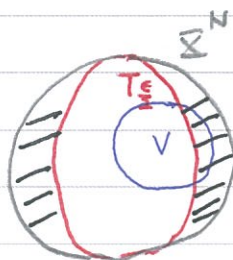
$$\Pr(U) \geq \Pr\left(\underbrace{\left(\frac{1}{N} \sum \log \frac{1}{p_{i_n}} - H\right)^2}_{T} < \epsilon^2\right) \geq 1 - \frac{\sigma^2}{\epsilon^2 N}$$

make N big so $\frac{\sigma^2}{\epsilon^2 N} \leq \delta$

* (b) $\frac{1}{N} H_\delta(\Sigma^N) \geq H - \epsilon$

$\Updownarrow$

every $V \subseteq \Sigma^N$ w/ $|V| < 2^{(H-\epsilon)N}$
must have $\Pr(V) < 1 - \delta$



$$\Pr(V) \leq \Pr(\text{atypical}) + |V| \cdot 2^{-(H-\frac{\epsilon}{2})N}$$

$$< \frac{\sigma^2}{(\frac{\epsilon}{2})^2 N} + 2^{-\frac{\epsilon}{2} \cdot N} \quad \text{max. prob in } T_{\frac{\epsilon}{2}}$$

$< 1 - \delta$ when N big enough