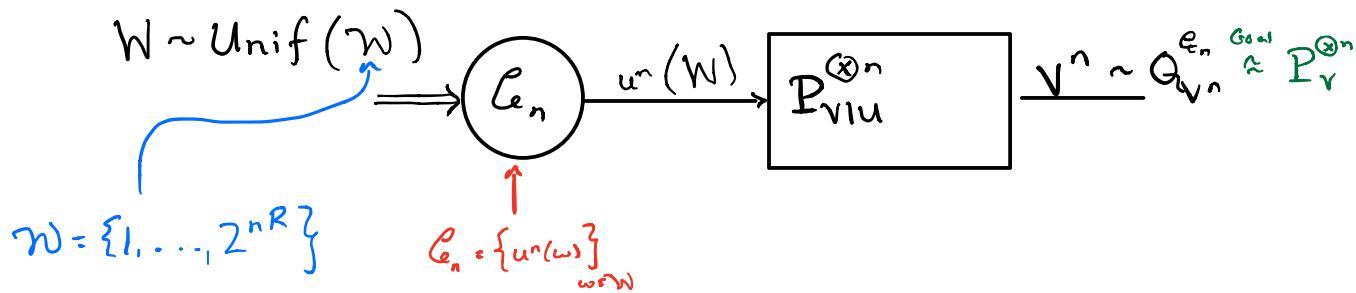


Recap:

We are considering approximate distribution simulation



Theorem (Approximate Distribution Simulation):

Fix $P_u \in \mathcal{P}(U)$, $P_v \in \mathcal{P}(V)$ and let $P_{V|U}$ be a transition kernel from $(U, 2^U)$ to $(V, 2^V)$ such that

$$\sum_{u \in U} P_u(u) P_{V|U}(v|u) = P_v(v) \quad \forall v \in V.$$

Let $C_n = \{U^n(w)\}_{w \in \mathcal{W}}$, $\mathcal{W} = \{1, \dots, 2^{nR}\}$, be an iid codebook distributed according to P_u , i.e. $U^n(w) \stackrel{iid}{\sim} P_u^{\otimes n}$ $\forall w \in \mathcal{W}$. If $R > I_P(U; V)$ ($= I(P_u; P_{V|U})$), then

$$\mathbb{E}_{C_n} \left[\delta_{TV} \left(Q_{V^n}^{(C_n)}, P_v^{\otimes n} \right) \right] \xrightarrow{n \rightarrow \infty} 0$$

where convergence happens exponentially fast and:

$$Q_{V^n}^{(C_n)}(v^n) = \frac{1}{|\mathcal{W}|} \sum_{w \in \mathcal{W}} P_{V|U}^{\otimes n}(v^n | u^n(w))$$

for any fixed codebook $C_n = \{u^n(w)\}_{w \in \mathcal{W}}$.

Proof

We want to separate typical U^n 's from atypical ones.

Define:

$$Q_1^{(C_n)}(v^n) := \frac{1}{|N|} \sum_{\omega \in N} P_{v|u}^{\otimes n}(v^n | U^n(\omega)) \mathbb{I}_{\{(v^n, U^n(\omega)) \in \mathcal{I}_\epsilon^{(C_n)}(P_{uv})\}}$$

$$Q_2^{(C_n)}(v^n) := \frac{1}{|N|} \sum_{\omega \in N} P_{v|u}^{\otimes n}(v^n | U^n(\omega)) \mathbb{I}_{\{(v^n, U^n(\omega)) \notin \mathcal{I}_\epsilon^{(C_n)}(P_{uv})\}}$$

where $\epsilon > 0$ is a fixed small number and $P_{uv} = P_u P_{v|u}$.

A few observations:

① $Q_1^{(C_n)}$, $Q_2^{(C_n)}$ are NOT probability measures on V^n but their sum $Q_1^{(C_n)} + Q_2^{(C_n)} = Q_{vn}^{(C_n)}$ is a probability measure on V^n .

② The expected value of $Q_{vn}^{(C_n)}$ is in fact the target distribution $P_v^{\otimes n}$:

$$\begin{aligned} \mathbb{E}_{C_n} \left[Q_{vn}^{(C_n)}(v^n) \right] &= \mathbb{E}_{C_n} \left[\frac{1}{|N|} \sum_{\omega \in N} P_{v|u}^{\otimes n}(v^n | U^n(\omega)) \right] \\ &= \frac{1}{|N|} \sum_{\omega \in N} \underbrace{\mathbb{E}_{C_n} \left[P_{v|u}^{\otimes n}(v^n | U^n(\omega)) \right]}_{\mathbb{E}_{U^n(1)}} \\ &= \mathbb{E}_{U^n(1)} \left[P_{v|u}^{\otimes n}(v^n | U^n(1)) \right] \\ &= \sum_{u \in U^n} P_u^{\otimes n}(u) P_{v|u}^{\otimes n}(v^n | u) \\ &= P_v^{\otimes n}(v^n) \end{aligned}$$

\implies We thus need to show that the TV distance b/t $Q_{v^n}^{(C_n)}$ and its expected value is small.

We split TV into 3 parts:

$$2 \cdot \mathbb{E}_{C_n} [\delta_{TV}(Q_{v^n}^{(C_n)}, P_v^{\otimes n})] = \mathbb{E}_{C_n} \left[\sum_{v^n \in V^n} \left| Q_{v^n}^{(C_n)}(v^n) - P_v^{\otimes n}(v^n) \right| \right]$$

Decompose

$$= \sum_{v^n \in \mathcal{I}_\epsilon^{(C_n)}(P_v)} \mathbb{E}_{C_n} \left| Q_{v^n}^{(C_n)}(v^n) - P_v^{\otimes n}(v^n) \right| + \sum_{v^n \notin \mathcal{I}_\epsilon^{(C_n)}(P_v)} \mathbb{E}_{C_n} \left| Q_{v^n}^{(C_n)}(v^n) - P_v^{\otimes n}(v^n) \right|$$

Triangle Inequality

$$\leq \sum_{v^n \in \mathcal{I}_\epsilon^{(C_n)}(P_v)} \mathbb{E}_{C_n} \left| Q_1^{(C_n)}(v^n) - \mathbb{E} Q_1^{(C_n)}(v^n) \right|$$

Can sum over all $v^n \in V^n$
which adds more nonnegative terms

$$+ \sum_{v^n \notin \mathcal{I}_\epsilon^{(C_n)}(P_v)} \mathbb{E}_{C_n} \left| Q_2^{(C_n)}(v^n) - \mathbb{E} Q_2^{(C_n)}(v^n) \right|$$

$$+ \sum_{v^n \notin \mathcal{I}_\epsilon^{(C_n)}(P_v)} \mathbb{E}_{C_n} \left| Q_{v^n}^{(C_n)}(v^n) - \mathbb{E}_{C_n} Q_{v^n}^{(C_n)}(v^n) \right|$$

$$\leq \sum_{v^n \in \mathcal{I}_\epsilon^{(C_n)}(P_v)} \mathbb{E}_{C_n} \left| Q_1^{(C_n)}(v^n) - \mathbb{E} Q_1^{(C_n)}(v^n) \right| \textcircled{1}$$

$$+ \sum_{v^n \in V^n} \mathbb{E}_{C_n} \left| Q_2^{(C_n)}(v^n) - \mathbb{E} Q_2^{(C_n)}(v^n) \right| + \sum_{v^n \notin \mathcal{I}_\epsilon^{(C_n)}(P_v)} \mathbb{E}_{C_n} \left| Q_{v^n}^{(C_n)}(v^n) - \mathbb{E}_{C_n} Q_{v^n}^{(C_n)}(v^n) \right|$$

②

③

Item ① is the interesting one and is saved for last.

We start by showing ② and ③ can be made small with n without further conditions.

$$\textcircled{2} = \sum_{v^n \in V^n} \mathbb{E}_{C_n} \left| Q_2^{(C_n)}(v^n) - \mathbb{E} Q_2^{(C_n)}(v^n) \right| \quad (\mathbb{E}|a-b| \leq \mathbb{E}[a] + \mathbb{E}[b])$$

$$\leq 2 \cdot \sum_{v^n \in V^n} \mathbb{E}_{C_n} Q_2^{(C_n)}(v^n)$$

$$= 2 \cdot \sum_{v^n \in V^n} \frac{1}{|w^n|} \sum_{w \in w^n} \mathbb{E}_{C_n} \left[P_{v|w}^{\otimes n}(v^n | w^n(\omega)) \cdot \mathbb{1}_{\{(v^n, w^n(\omega)) \in \mathcal{T}_\epsilon^{C_n}(P_{uv})\}} \right]$$

$$= 2 \cdot \sum_{v^n \in V^n} \frac{1}{|w^n|} \sum_{w \in w^n} \mathbb{E}_{w^n(1)} \left[P_{v|w}^{\otimes n}(v^n | w^n(1)) \cdot \mathbb{1}_{\{(v^n, w^n(1)) \in \mathcal{T}_\epsilon^{C_n}(P_{uv})\}} \right]$$

$$= 2 \cdot \sum_{v^n \in V^n} \sum_{w^n \in w^n} P_w^{\otimes n}(w^n) P_{v|w}^{\otimes n}(v^n | w^n) \mathbb{1}_{\{(v^n, w^n) \in \mathcal{T}_\epsilon^{C_n}(P_{uv})\}}$$

$$= 2 \cdot \sum_{(u^n, v^n) \notin \mathcal{T}_\epsilon^{C_n}(P_{uv})} P_{uv}^{\otimes n}(u^n, v^n)$$

$$= 2 \cdot \mathbb{P}_{P_{uv}^{\otimes n}}((u^n, v^n) \notin \mathcal{T}_\epsilon^{C_n}(P_{uv}))$$

$$\leq 2 \cdot 2 \cdot |u||v| e^{-2n\epsilon^2 \mu_{uv}^2} \xrightarrow{n \rightarrow \infty} 0$$

$\left\{ \begin{array}{l} \text{where} \\ \mu_{uv}^2 = \min_{(u,v) \in \text{supp } P_{uv}} P_{uv}(u,v) \end{array} \right.$

$P_{uv}(u,v) > 0$

$$\textcircled{3} = \sum_{v^n \notin \mathcal{T}_\epsilon^{C_n}(P_v)} \mathbb{E}_{C_n} \left| Q_{v^n}^{(C_n)} - \mathbb{E}_{C_n} Q_{v^n}^{(C_n)} \right|$$

$$\leq 2 \cdot \sum_{v^n \notin \mathcal{T}_\epsilon^{C_n}(P_v)} \mathbb{E} Q_{v^n}^{(C_n)} (v^n)$$

$$= 2 \cdot \sum_{v^n \notin \mathcal{T}_\epsilon^{C_n}(P_v)} P_v^{\otimes n} (v^n)$$

$$= 2 \cdot P_{P_v^{\otimes n}} (v^n \in \mathcal{T}_\epsilon^{C_n}(P_v))$$

$$\leq 2 \cdot 2 |\mathcal{V}| e^{-2n\epsilon^2 \mu_v^2} \xrightarrow{n \rightarrow \infty} 0$$

We are thus left with bounding the first item, $\textcircled{1}$. Since now we are dealing with jointly typical U^n and V^n sequences we need a different analysis.

Let us derive an upper bound on each summand that is independent of $v^n \in \mathcal{T}_\epsilon^{C_n}(P_v)$

$$\mathbb{E}_{C_n} \left| Q_1^{(C_n)} (v^n) - \mathbb{E} Q_1^{(C_n)} (v^n) \right| \leq \sqrt{\mathbb{E}_{C_n} \left[(Q_1^{(C_n)} (v^n) - \mathbb{E} Q_1^{(C_n)} (v^n))^2 \right]}$$

Jensen's Inequality

$$\mathbb{E}[1 \cdot 1] = \mathbb{E}[\sqrt{(-)^2}] \leq \sqrt{\mathbb{E}[(1)^2]} = \sqrt{\text{Var}_{C_n} (Q_1^{(C_n)} (v^n))}$$

Since $C_n = \{U^n(\omega)\}_{\omega \in \mathcal{W}}$ are iid the variance tensorizes:

$$\text{Var}_{C_n} (G_2^{(n)}(v^n)) = \text{Var}_{C_n} \left(\frac{1}{|\mathcal{W}|} \sum_{\omega \in \mathcal{W}} P_{v|u}^{\otimes n}(v^n | U^n(\omega)) \mathbb{1}_{\{(v^n, U^n(\omega)) \in \mathcal{I}_\epsilon^{(n)}(P_{uv})\}} \right)$$

variance of
sum of iid
random variables

$$= \frac{1}{|\mathcal{W}|^2} \sum_{\omega \in \mathcal{W}} \text{Var}_{C_n} \left(P_{v|u}^{\otimes n}(v^n | U^n(\omega)) \mathbb{1}_{\{(v^n, U^n(\omega)) \in \mathcal{I}_\epsilon^{(n)}(P_{uv})\}} \right)$$

Due to symmetry this is independent of $\omega \in \mathcal{W}$ and $U^n(\omega)$ can be replaced with $U^n(1)$ throughout

$$\begin{aligned} &= 2^{-nR} \text{Var}_{U^n(1)} \left(P_{v|u}^{\otimes n}(v^n | U^n(1)) \mathbb{1}_{\{(v^n, U^n(1)) \in \mathcal{I}_\epsilon^{(n)}(P_{uv})\}} \right) \\ &\leq 2^{-nR} \mathbb{E}_{U^n(1)} \left[\left(P_{v|u}^{\otimes n}(v^n | U^n(1)) \mathbb{1}_{\{(v^n, U^n(1)) \in \mathcal{I}_\epsilon^{(n)}(P_{uv})\}} \right)^2 \right] \\ &= 2^{-nR} \sum_{u^n \in U^n} P_u^{\otimes n}(u^n) \left(P_{v|u}^{\otimes n}(v^n | u^n) \right)^2 \mathbb{1}_{\{(v^n, U^n(1)) \in \mathcal{I}_\epsilon^{(n)}(P_{uv})\}} \end{aligned}$$

$$\begin{aligned} &= 2^{-nR} \sum_{u^n \in \mathcal{I}_\epsilon^{(n)}(P_{uv} | v^n)} P_u^{\otimes n}(u^n) \underbrace{P_{v|u}^{\otimes n}(v^n | u^n)}_{\leq 2^{-n(H(v|u) - \delta_1(\epsilon))}} \underbrace{P_{v|u}^{\otimes n}(v^n | u^n)}_{\leq 2^{-n(H(v|u) - \delta_1(\epsilon))}} \\ &\quad \text{def: } \mathcal{I}_\epsilon^{(n)}(P_{uv}) := \{u^n | (u^n, v^n) \in \mathcal{I}_\epsilon^{(n)}(P_{uv})\} \end{aligned}$$

$$\begin{aligned} P_v^{\otimes n}(v^n) &\leq 2^{-nR} 2^{-n(H(V|U) - \delta_1(\epsilon))} \sum_{u^n \in U^n} P_u^{\otimes n} P_{v|u}^{\otimes n}(v^n | u^n) \end{aligned}$$

$$\begin{aligned} &\leq 2^{-n(H(V) - \delta_2(\epsilon))} \\ &\quad \text{def: } \mathcal{I}_\epsilon^{(n)}(P_{uv}) \xrightarrow{\epsilon \rightarrow 0} \emptyset \end{aligned}$$

$$\begin{aligned} &= 2^{-nR} 2^{-n(H(V|U) - \delta_1(\epsilon))} \underbrace{P_v^{\otimes n}(v^n)}_{2^{-n(H(V) - \delta_2(\epsilon))}} \\ &\leq 2^{-nR} 2^{-n(H(V|U) - \delta_1(\epsilon))} 2^{-n(H(V) - \delta_2(\epsilon))} \end{aligned}$$

$$= 2^{-n(R + H(V|U) + H(V) - \delta_1(\epsilon) - \delta_2(\epsilon))}$$

Recap

$$\begin{aligned} \Rightarrow \textcircled{1} &= \sum_{v^n \in \mathcal{I}_\epsilon^{(C_n)}(\mathbb{P}_r)} \mathbb{E}_{c_n} \left| Q_1^{(C_n)}(v^n) - \mathbb{E} Q_1^{(C_n)}(v^n) \right| \\ &\leq \sum_{v^n \in \mathcal{I}_\epsilon^{(C_n)}(\mathbb{P}_r)} 2^{-\frac{n}{2}(R + H(v|u) + H(v) - \delta_1(\epsilon) - \delta_2(\epsilon))} \\ &\leq 2^{n(H(v) + \delta_2(\epsilon))} 2^{-\frac{n}{2}(R + H(v|u) + H(v) - \delta_1(\epsilon) - \delta_2(\epsilon))} \\ &= 2^{-\frac{n}{2}(R + H(v|u) - H(v) - \delta_1(\epsilon) - 3\delta_2(\epsilon))} \end{aligned}$$

where $\delta(\epsilon) = \delta_1(\epsilon) - 3\delta_2(\epsilon) \xrightarrow{\epsilon \rightarrow 0} 0$

\Rightarrow Thus if $R > I_p(u; v)$ then we can take ϵ small enough such that the RHS above will go to 0 as $n \rightarrow \infty$, and converges exponentially fast

Lastly, we will show how the above expected value result can be strengthened into a concentration inequality.

Definition (Functions w/ Bounded Differences): A function $f: \mathcal{X}^n \rightarrow \mathbb{R}$ is said to have $(C_i)_{i=1}^n$ bounded difference if

$$|f(x_1, \dots, x_n) - f(x_1, \dots, x_{i-1}, \tilde{x}, x_{i+1}, \dots, x_n)| \leq c_i \quad \forall x_1, \dots, x_n, \tilde{x} \in \mathcal{X} \quad i=1, 2, \dots, n$$

Theorem (McDiarmid's Inequality): Let X_1, \dots, X_n be \mathcal{X} -valued independent random variables and $f: \mathcal{X}^n \rightarrow \mathbb{R}$ have $(c_i)_{i=1}^n$ -bounded differences. Then

$$\mathbb{P}\left(\left|f(X_1, \dots, X_n) - \mathbb{E}[f(X_1, \dots, X_n)]\right| \geq t\right) \leq 2^{-\frac{2t^2}{\sum_{i=1}^n c_i^2}}$$

Theorem (Strong Approximate Distribution Simulation):

Under the same setup of the expected value claim for approximate distribution simulation, if $R > I_2(U; V)$ then there exists $\gamma_1, \gamma_2 > 0$ such that

$$\mathbb{P}\left(\delta_{TV}(Q_{v^n}^{(e_n)}, P_v^{\otimes n}) > e^{-\gamma_1 n}\right) \leq e^{-e^{\gamma_2 n}}$$

for n large enough.

Proof Outline: Define $f(U_1, \dots, U_n) = \delta_{TV}(Q_{v^n}^{(e_n)}, P_v^{\otimes n})$

$$\begin{aligned} W := (w) \\ &= \frac{1}{2} \mathbb{E}_{P_v^{\otimes n}} \left[\frac{Q_{v^n}^{(e_n)}(v^n)}{P_v^{\otimes n}(v^n)} - 1 \right] \\ &= \frac{1}{2} \mathbb{E}_{P_v^{\otimes n}} \left[\frac{1}{W} \sum_{w=1}^W \frac{P_{v|U}(v^n | U_w)}{P_v^{\otimes n}(v^n)} - 1 \right] \end{aligned}$$

\Rightarrow We can show that $f(U_1, \dots, U_n)$ has $\left(\frac{1}{W}\right)_{i=1}^W$ bounded difference

Thus we can apply McDiarmid's inequality & use the fact that in expectation

$$\mathbb{E}_{C_n} \delta_{TV}(Q_{V^n}, P_r^{\otimes n}) \leq e^{-\gamma n}$$

for some $\gamma > 0$ and n sufficiently large.

Taking t to be exponentially small with n ($t = e^{-\tilde{\gamma}n}$, $\tilde{\gamma} > 0$) and plugging into the above will establish the claim.

Discussion Section

Proof of Strong Approximate Distribution Simulation

$$W = 2^{nR}, U_w = U^n(w)$$

Fix $\ell_n := \{U_w\}_{w \in \mathcal{W}}$ and define

$$f: \ell_n \mapsto \delta_{TV}(Q_{V^n}^{(\ell_n)}, P_r^{\otimes n})$$

i.e.

$$f(\ell_n) = f(U_1, \dots, U_W) := \delta_{TV}(Q_{V^n}^{(\ell_n)}, P_r^{\otimes n})$$

$$\delta_{TV} = D_{\frac{1}{2}}|_{x=1}$$

$$D_f(P||Q) = \mathbb{E}_Q [f(\frac{dP}{dQ})]$$

ratio of pmf in discrete case

$$= \frac{1}{2} \mathbb{E}_{P_r^{\otimes n}} \left| \frac{Q_{V^n}^{(\ell_n)}}{P_r^{\otimes n}} - 1 \right|$$

$$= \frac{1}{2} \mathbb{E}_{P_r^{\otimes n}} \left| \frac{1}{W} \sum_{w=1}^W \frac{P_r^{\otimes n}(V^n | u_w)}{P_r^{\otimes n}(V^n)} - 1 \right|$$

Bounded Differences Property:

Let $u_1, \dots, u_w, \tilde{u} \in \mathcal{U}^n$ and consider:

$$(*) f(u_1, \dots, u_{l-1}, u_l, u_{l+1}, \dots, u_w) - f(u_1, \dots, u_{l-1}, \tilde{u}, u_{l+1}, \dots, u_w) \quad \left. \begin{array}{l} \text{absolute value} \\ \text{omitted for} \\ \text{proof simplicity.} \\ \text{Other direction similar} \end{array} \right\}$$

$$= \frac{1}{2} \mathbb{E}_{P_v^{(n)}} \left| \frac{1}{W} \sum_{w=1}^W \frac{P_{v|u}^{(n)}(V^n | u_w)}{P_v^{(n)}(V^n)} - 1 \right| - \frac{1}{2} \mathbb{E}_{P_v^{(n)}} \left| \frac{1}{W} \left(\sum_{w \neq l} \frac{P_{v|u}^{(n)}(V^n | u_w)}{P_v^{(n)}(V^n)} + \frac{P_{v|\tilde{u}}^{(n)}(V^n | \tilde{u})}{P_v^{(n)}(V^n)} \right) - 1 \right|$$

①

$$\textcircled{1} = \frac{1}{2} \mathbb{E}_{P_v^{(n)}} \left| \frac{1}{W} \sum_{w \neq l} \frac{P_{v|u}^{(n)}(V^n | u_w)}{P_v^{(n)}(V^n)} + \frac{1}{W} \frac{P_{v|u}^{(n)}(V^n | u_l)}{P_v^{(n)}(V^n)} + \frac{1}{W} \frac{P_{v|u}^{(n)}(V^n | \tilde{u})}{P_v^{(n)}(V^n)} - \frac{1}{W} \frac{P_{v|u}^{(n)}(V^n | \tilde{u})}{P_v^{(n)}(V^n)} - 1 \right|$$

(Triangle Ineq.)

$$\leq \frac{1}{2} \mathbb{E}_{P_v^{(n)}} \left| \frac{1}{W} \sum_{w \neq l} \frac{P_{v|u}^{(n)}(V^n | \tilde{u}_w)}{P_v^{(n)}(V^n)} + \frac{1}{W} \frac{P_{v|u}^{(n)}(V^n | \tilde{u})}{P_v^{(n)}(V^n)} - 1 \right|$$

$$+ \frac{1}{2} \mathbb{E}_{P_v^{(n)}} \left| \frac{1}{W} \frac{P_{v|u}^{(n)}(V^n | u_l)}{P_v^{(n)}(V^n)} - \frac{1}{W} \frac{P_{v|u}^{(n)}(V^n | \tilde{u})}{P_v^{(n)}(V^n)} \right|$$

$$\Rightarrow (*) \leq \frac{1}{2} \cdot \frac{1}{W} \mathbb{E} \left| \frac{P_{v|u}^{(n)}(V^n | u_l)}{P_v^{(n)}(V^n)} - \frac{P_{v|u}^{(n)}(V^n | \tilde{u})}{P_v^{(n)}(V^n)} \right|$$

$$\leq \frac{1}{2W} \left[\mathbb{E}_{P_v^{(n)}} \frac{P_{v|u}^{(n)}(V^n | u_l)}{P_v^{(n)}(V^n)} + \mathbb{E}_{P_v^{(n)}} \frac{P_{v|u}^{(n)}(V^n | \tilde{u})}{P_v^{(n)}(V^n)} \right]$$

$$= \frac{1}{W}$$

$\Rightarrow f(u_1, \dots, u_w) = \delta_{rr}(Q_{vn}, P_v^{(n)})$ is a $\left(\frac{1}{W}\right)_{w=1}^W$ -bounded difference function.

Apply McDiarmids inequality

$$\begin{aligned} \Pr\left(f(U_1, \dots, U_W) > \mathbb{E}[f(U_1, \dots, U_W)] + t\right) &\leq e^{-\frac{2t^2}{W \cdot \frac{1}{W}}} \\ &= e^{-2t^2 W} \\ W = 2^{nR} = e^{nR \ln(2)} &= e^{-2t^2 \cdot e^{nR \ln(2)}} \end{aligned}$$

Recall that if $R > I_p(U; V)$, then $\mathbb{E}[f(U_1, \dots, U_W)] \leq e^{\gamma_n}$, for some $\gamma > 0$, and large enough n .
 Let $t = e^{-\tilde{\gamma}n}$; $\tilde{\gamma}$ to be chosen later.

First observe

$$\begin{aligned} \Pr\left(f(U_1, \dots, U_W) > e^{-\gamma n} + e^{-\tilde{\gamma}n}\right) &\leq e^{-2e^{-2\tilde{\gamma}n} \cdot e^{n \cdot R \cdot \ln(2)}} \\ &= e^{-2e^{n(R \ln(2) - 2\tilde{\gamma})}} \end{aligned}$$

Take $0 < \tilde{\gamma} < \frac{R \ln(2)}{2}$,

$$\gamma_1 = \min\{\gamma, \tilde{\gamma}\} > 0 \quad \text{and} \quad \gamma_2 = R \ln(2) - 2\tilde{\gamma} > 0$$

■