

APPENDIX A
PROOF OF SECTION IV

A. Auxiliary Lemmas and Propositions

Lemma 3. We recall $\alpha^* = \arg \max_{\alpha \in (0,1)} D_\alpha(\varphi_0 \parallel \varphi_1)$ and Y has density function φ_{α^*} , then $\sum_{j=1}^n \mathbb{E}[Z_j] = 0$.

Proof. By definition of Y , we have

$$\mathbb{E}\left[\log \frac{\varphi_0(Y)}{\varphi_1(Y)}\right] = \int_{\Omega} \varphi_{\alpha^*} \log \frac{\varphi_0(y)}{\varphi_1(y)} d\mu(y) = e^{-D_{\alpha^*}(\varphi_0 \parallel \varphi_1)} \int_{\Omega} \varphi_0^{1-\alpha^*} \varphi_1^{\alpha^*} \log \frac{\varphi_1}{\varphi_0} d\mu.$$

We omit the variable y in the integral in the rest of the proof. Recall that we assume the Kullback–Leibler divergence $D_{\text{KL}}(\varphi_0 \parallel \varphi_1)$ and $D_{\text{KL}}(\varphi_1 \parallel \varphi_0)$ exist, so for $\alpha \in (0, 1)$, $|\varphi_0^{1-\alpha} \varphi_1^\alpha \log \frac{\varphi_0}{\varphi_1}| \leq |(\varphi_0 + \varphi_1) \log \frac{\varphi_0}{\varphi_1}|$ is integrable. By mean value theorem and dominated convergence theorem,

$$\int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^\alpha \log \frac{\varphi_1}{\varphi_0} d\mu = \int_{\Omega} \frac{d}{d\alpha} \varphi_0^{1-\alpha} \varphi_1^\alpha d\mu = \frac{d}{d\alpha} \int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^\alpha d\mu.$$

Since $\alpha \mapsto -\varphi_0(x)^{1-\alpha} \varphi_1(x)^\alpha$ is convex for $x \in \Omega$, $\alpha \mapsto -\int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^\alpha d\mu$ is also convex, and it is indeed strictly convex if $\varphi_0 \neq \varphi_1$ on a set with nonzero measure. Therefore, $D_\alpha(\varphi_0 \parallel \varphi_1)$ achieves maximum if and only if $\frac{d}{d\alpha} \int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^\alpha d\mu = 0$, which is true if we evaluate at $\alpha = \alpha^*$. Hence $\mathbb{E}[\log l(Y)] = 0$. \square

Proposition 1. Let Φ be the cumulative distribution function of standard normal distribution, then for $x > 0$,

$$\frac{1}{x} - \frac{1}{x^3} \leq \sqrt{2\pi} e^{x^2/2} \Phi(-x) \leq \frac{1}{x}. \quad (16)$$

and

$$\sqrt{2\pi} e^{x^2/2} \left(\Phi(x) - \frac{1}{2} \right) \geq x + \frac{x^3}{3}. \quad (17)$$

In particular,

$$\Phi(x) \geq \frac{1}{2} + \frac{1}{\sqrt{2\pi}} \left(x - \frac{2x^3}{3} \right). \quad (18)$$

Proof. For $x > 0$, we have divergent series expanded at ∞ :

$$\sqrt{2\pi} e^{x^2/2} \Phi(-x) = \frac{1}{x} + \sum_{i=1}^{\infty} \frac{(-1)^i (2i-1)!}{2^{i-1} (i-1)!} \frac{1}{x^{2i+1}} = \frac{1}{x} - \frac{1}{x^3} + \frac{3}{x^5} - \frac{15}{x^7} + \dots$$

which implies (16). We also have the power series expanded at 0. Let $n!! = 1 \cdot 3 \cdot \dots \cdot n$ for odd integer n , we have

$$\sqrt{2\pi} e^{x^2/2} \left(\Phi(x) - \frac{1}{2} \right) = \sum_{i=0}^{\infty} \frac{x^{2i+1}}{(2i+1)!!} = x + \frac{x^3}{3} + \frac{x^5}{3 \cdot 5} + \frac{x^7}{3 \cdot 5 \cdot 7} + \dots$$

which implies (17). Then we expand $e^{-x^2/2} \left(x + \frac{x^3}{3} \right)$ and have

$$e^{-x^2/2} \left(x + \frac{x^3}{3} \right) = x - \frac{2x^3}{3} + \frac{7x^5}{30} - \dots$$

Then we obtain (18). \square

Lemma 4. Let $g_\alpha(x) = \exp(\min(\alpha x, (\alpha-1)x))$. Suppose $Z \sim \mathcal{N}(0, \sigma^2)$, then

$$\frac{1}{\sqrt{2\pi}\sigma\alpha(1-\alpha)} - \frac{1}{\sqrt{2\pi}\sigma^3\alpha^3(1-\alpha)^3} \leq \mathbb{E}[g_\alpha(Z)] \leq \frac{1}{\sqrt{2\pi}\sigma\alpha(1-\alpha)}.$$

Proof. We have

$$\begin{aligned} \mathbb{E}[g_\alpha(Z)] &= \mathbb{E}[\exp(\min(\alpha Z, (\alpha-1)Z))] \\ &= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^0 e^{\alpha x} e^{-\frac{x^2}{2\sigma^2}} dx + \frac{1}{\sqrt{2\pi}} \int_0^{\infty} e^{(\alpha-1)x} e^{-\frac{x^2}{2\sigma^2}} dx \\ &= \frac{e^{\sigma^2\alpha^2/2}}{\sqrt{2\pi}\sigma} \int_{-\infty}^0 e^{-\frac{(x-\sigma^2\alpha)^2}{2\sigma^2}} dx + \frac{e^{\sigma^2(1-\alpha)^2/2}}{\sqrt{2\pi}\sigma} \int_0^{\infty} e^{-\frac{(x-\sigma^2(1-\alpha))^2}{2\sigma^2}} dx \end{aligned}$$

$$= \exp\left(\frac{\sigma^2 \alpha^2}{2}\right) \Phi(-\sigma \alpha) + \exp\left(\frac{\sigma^2 (1-\alpha)^2}{2}\right) \Phi(-\sigma(1-\alpha))$$

. By (16), we have

$$\mathbb{E}[g_\alpha(Z)] \leq \frac{1}{\sqrt{2\pi}\sigma\alpha} + \frac{1}{\sqrt{2\pi}\sigma(1-\alpha)} = \frac{1}{\sqrt{2\pi}\sigma\alpha(1-\alpha)},$$

and

$$\begin{aligned} \mathbb{E}[g_\alpha(Z)] &\geq \frac{1}{\sqrt{2\pi}\sigma\alpha} - \frac{1}{\sqrt{2\pi}\sigma^3\alpha^3} + \frac{1}{\sqrt{2\pi}\sigma(1-\alpha)} - \frac{1}{\sqrt{2\pi}\sigma^3(1-\alpha)^3} \\ &\geq \frac{1}{\sqrt{2\pi}\sigma\alpha(1-\alpha)} - \frac{1}{\sqrt{2\pi}\sigma^3\alpha^3(1-\alpha)^3}. \end{aligned}$$

This completes the proof. \square

Theorem 3 (Berry-Esseen theorem). *Let $\{Z_i\}_{i=1}^n$ be independent random variables with zero means and $\mathbb{E}[\sum_{i=1}^n Z_i^2] = \sigma^2$. Let F be the distribution function of $\sum_{i=1}^n Z_i/\sigma$, then there exists an absolute constant $C_0 \leq 0.56$ such that for every $x \in \mathbb{R}$,*

$$|F(x) - \Phi(x)| \leq \frac{C_0 \sum_{i=1}^n \mathbb{E}[|Z_i|^3]}{\sigma^3}. \quad (19)$$

Proof. The proof is given in [23]. \square

B. Proof of Theorem 1

We recall that $\varphi_\alpha = \varphi_0^{1-\alpha} \varphi_1^\alpha e^{D_\alpha(\varphi_0\|\varphi_1)}$. In this proof, we briefly write $\int_\Omega \min(\varphi_0(x), \varphi_1(x)) d\mu(x)$ as $\int_\Omega \min(\varphi_0, \varphi_1) d\mu$. We have

$$\begin{aligned} \int_\Omega \min(\varphi_0, \varphi_1) d\mu &= \int_\Omega \varphi_0^{1-\alpha} \varphi_1^\alpha \min\left(\left(\frac{\varphi_0}{\varphi_1}\right)^\alpha, \left(\frac{\varphi_1}{\varphi_0}\right)^{1-\alpha}\right) d\mu \\ &= e^{-D_\alpha(\varphi_0\|\varphi_1)} \int_\Omega \varphi_\alpha \exp\left(\min\left(\log \frac{\varphi_0}{\varphi_1}, (\alpha-1) \log \frac{\varphi_0}{\varphi_1}\right)\right) d\mu \\ &= e^{-D_\alpha(\varphi_0\|\varphi_1)} \mathbb{E}\left[g_\alpha\left(\log \frac{\varphi_0(Y)}{\varphi_1(Y)}\right)\right] \\ &= e^{-D_\alpha(\varphi_0\|\varphi_1)} \mathbb{E}\left[g_\alpha\left(\sum_{j=1}^n Z_j\right)\right], \end{aligned}$$

where $g_\alpha(x) = \exp(\min(\alpha x, (\alpha-1)x))$ from Lemma 3 and Y and Z_j have been defined in Section IV. Now it suffices to show that

$$\mathbb{E}\left[g_\alpha\left(\log \frac{\varphi_0(Y)}{\varphi_1(Y)}\right)\right] \asymp \frac{1}{\sqrt{n}\bar{\sigma}_n\alpha(1-\alpha)}$$

when $\alpha = \alpha^*$. In this case, we have $\mathbb{E}[\sum_{j=1}^n Z_j] = 0$ by Lemma 3. Let us define $\sigma^2 = \mathbb{E}[\sum_{j=1}^n Z_j^2] = -D''_{\alpha^*}(\varphi_0\|\varphi_1)$. By assumption $\sum_{j=1}^n \mathbb{E}[Z_j]^3 \leq C_1 n \bar{\sigma}_n^2$, the distribution function F of $\sum_{j=1}^n Z_j/\sigma$ satisfies for $x \in \mathbb{R}$, $|F(x) - \Phi(x)| \leq \frac{C}{\sigma}$ where $C := 0.56C_1$. Recall that $\log l(Y) = \sum_{j=1}^n Z_j$, we have

$$\begin{aligned} \mathbb{E}\left[g_\alpha\left(\log \frac{\varphi_0(Y)}{\varphi_1(Y)}\right)\right] &= \int_0^1 \mathbb{P}\left(g_\alpha\left(\log \frac{\varphi_0(Y)}{\varphi_1(Y)}\right) > x\right) dx \\ &= \int_0^1 \mathbb{P}\left(\frac{\log x}{\alpha\sigma} < \frac{1}{\sigma} \log \frac{\varphi_0(Y)}{\varphi_1(Y)} < \frac{\log x}{(\alpha-1)\sigma}\right) dx \\ &= \int_0^1 F\left(\frac{\log x}{(\alpha-1)\sigma}\right) - F\left(\frac{\log x}{\alpha\sigma}\right) dx \\ &\leq \int_0^1 \Phi\left(\frac{\log x}{(\alpha-1)\sigma}\right) - \Phi\left(\frac{\log x}{\alpha\sigma}\right) dx + \frac{2C}{\sigma} \\ &= \frac{1}{\sqrt{2\pi}\sigma\alpha(1-\alpha)} + \frac{2C}{\sigma} \leq \frac{1+C/2}{\sigma\alpha(1-\alpha)}. \end{aligned}$$

This completes the proof of upper bound. On the other hand, for any $t \in [0, 1]$,

$$\begin{aligned}\mathbb{E}\left[g_\alpha\left(\log \frac{\varphi_0(Y)}{\varphi_1(Y)}\right)\right] &= \int_0^1 F\left(\frac{\log x}{(\alpha-1)\sigma}\right) - F\left(\frac{\log x}{\alpha\sigma}\right) dx \\ &\geq \int_0^t \Phi\left(\frac{\log x}{(\alpha-1)\sigma}\right) - \Phi\left(\frac{\log x}{\alpha\sigma}\right) - \frac{2C}{\sigma} dx \\ &= \int_0^t \Phi\left(\frac{\log x}{(\alpha-1)\sigma}\right) - \Phi\left(\frac{\log x}{\alpha\sigma}\right) dx - \frac{2tC}{\sigma}.\end{aligned}$$

By Fubini's theorem,

$$\begin{aligned}\int_0^t \Phi\left(\frac{\log x}{\sigma(\alpha-1)}\right) - \Phi\left(\frac{\log x}{\sigma\alpha}\right) dx &= \int_0^t \frac{1}{\sqrt{2\pi}} \int_{\frac{\log x}{\sigma\alpha}}^{\frac{\log x}{\sigma(\alpha-1)}} e^{-y^2/2} dy dx \\ &= \frac{1}{\sqrt{2\pi}} \left[\int_{-\infty}^{\frac{\log t}{\sigma\alpha}} e^{\sigma\alpha y - \frac{y^2}{2}} dy + \int_{\frac{\log t}{\sigma(\alpha-1)}}^{\infty} e^{\sigma(\alpha-1)x - \frac{y^2}{2}} dy \right] + t \left[\Phi\left(\frac{\log t}{\sigma(\alpha-1)}\right) - \Phi\left(\frac{\log t}{\sigma\alpha}\right) \right].\end{aligned}\tag{20}$$

The first integral in the last line can be evaluated as

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\frac{\log t}{\sigma\alpha}} e^{\sigma\alpha y - \frac{y^2}{2}} dy = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\frac{\log t}{\sigma\alpha}} e^{\frac{\sigma^2\alpha^2}{2} - \frac{(y+\sigma\alpha)^2}{2}} dy = e^{\frac{\sigma^2\alpha^2}{2}} \Phi\left(\frac{\log t}{\sigma\alpha} - \sigma\alpha\right).$$

For the first integral, we similarly have

$$\frac{1}{\sqrt{2\pi}} \int_{\frac{\log t}{\sigma(\alpha-1)}}^{\infty} e^{\sigma(\alpha-1)x - \frac{y^2}{2}} dy = e^{\frac{\sigma^2(1-\alpha)^2}{2}} \Phi\left(\frac{\log t}{\sigma(1-\alpha)} - \sigma(1-\alpha)\right).$$

Assuming $\sigma\alpha(1-\alpha) \geq \sqrt{2\pi}C \vee 2$ and letting $t = \exp[-2\sqrt{2\pi}C(1-\alpha)\alpha]$, using $\alpha(1-\alpha) \leq 1/4$, we have

$$-\frac{\log t}{\sigma\alpha} \leq \frac{2\sqrt{2\pi}C\alpha(1-\alpha)}{\sqrt{2\pi}C} \leq \frac{1}{2} \quad \text{and} \quad \sigma\alpha - \frac{\log t}{\sigma\alpha} \leq \sigma\alpha + \frac{1}{2} \leq \frac{5\sigma\alpha}{4}.$$

By (16) and the fact that the function $\frac{1}{x} - \frac{1}{x^3}$ is decreasing on $[\sqrt{3}, \infty]$, we have

$$\begin{aligned}e^{\frac{\sigma^2\alpha^2}{2}} \Phi\left(\frac{\log t}{\sigma\alpha} - \sigma\alpha\right) &\geq \frac{1}{\sqrt{2\pi}} \exp\left(\frac{\sigma^2\alpha^2 - \left(\frac{\log t}{\sigma\alpha} - \sigma\alpha\right)^2}{2}\right) \left[\frac{1}{\sigma\alpha - \frac{\log t}{\sigma\alpha}} - \frac{1}{\left(\sigma\alpha - \frac{\log t}{\sigma\alpha}\right)^3} \right] \\ &\geq \frac{1}{\sqrt{2\pi}} \exp\left(\log t - \frac{1}{2}\left(\frac{\log t}{\sigma\alpha}\right)^2\right) \left[\frac{4}{5\sigma\alpha} - \frac{64}{125\sigma^3\alpha^3} \right] \\ &\geq \frac{1}{\sqrt{2\pi}} \exp(-2\sqrt{2\pi}C(1-\alpha)\alpha - 1/8) \left[\frac{4}{5\sigma\alpha} - \frac{64}{125(4\sigma\alpha)} \right] \\ &\geq \frac{\exp(-2\sqrt{2\pi}C)}{5\sigma\alpha}.\end{aligned}$$

Similarly, we have

$$e^{\frac{\sigma^2(1-\alpha)^2}{2}} \Phi\left(\frac{\log t}{\sigma(1-\alpha)} - \sigma(1-\alpha)\right) \geq \frac{\exp(-2\sqrt{2\pi}C)}{5\sigma(1-\alpha)}.$$

Hence the integral in (20) has lower bound

$$\frac{1}{\sqrt{2\pi}} \left[\int_{-\infty}^{\frac{\log t}{\sigma\alpha}} e^{\sigma\alpha y - \frac{y^2}{2}} dy + \int_{\frac{\log t}{\sigma(\alpha-1)}}^{\infty} e^{\sigma(\alpha-1)x - \frac{y^2}{2}} dy \right] \geq \frac{\exp(-2\sqrt{2\pi}C)}{5\sigma\alpha} + \frac{\exp(-2\sqrt{2\pi}C)}{5\sigma(1-\alpha)} = \frac{\exp(-2\sqrt{2\pi}C)}{5\sigma\alpha(1-\alpha)}.\tag{21}$$

Now we consider another term $\Phi\left(\frac{\log t}{\sigma(\alpha-1)}\right)$ in (20). By (18), we have

$$\begin{aligned}\Phi\left(\frac{\log t}{\sigma(\alpha-1)}\right) &\geq \frac{1}{2} + \frac{1}{\sqrt{2\pi}} \left(\frac{\log t}{\sigma(\alpha-1)} - \frac{2}{3} \left(\frac{\log t}{\sigma(\alpha-1)} \right)^3 \right) \\ &= \frac{1}{2} + \frac{1}{\sqrt{2\pi}} \left(\frac{-2\sqrt{2\pi}C\alpha(1-\alpha)}{\sigma(\alpha-1)} - \frac{2}{3} \left(\frac{-2\sqrt{2\pi}C\alpha(1-\alpha)}{\sigma(\alpha-1)} \right)^3 \right)\end{aligned}$$

$$= \frac{1}{2} + \frac{2C\alpha}{\sigma} - \frac{32\pi C^3 \alpha^3}{3\sigma^3}$$

and similarly the term $-\Phi\left(\frac{\log t}{\sigma\alpha}\right)$ has lower bound

$$-\Phi\left(\frac{\log t}{\sigma\alpha}\right) = \Phi\left(-\frac{\log t}{\sigma\alpha}\right) - 1 \geq \frac{1}{2} + \frac{2C(1-\alpha)}{\sigma} - \frac{32\pi C^3(1-\alpha)^3}{3\sigma^3} - 1.$$

Therefore,

$$t\left[\Phi\left(\frac{\log t}{(\alpha-1)\sigma}\right) - \Phi\left(\frac{\log t}{\alpha\sigma}\right)\right] - \frac{2tC}{\sigma} \geq \frac{2tC}{\sigma} - \frac{32\pi C^3 \alpha^3}{3\sigma^3} - \frac{32\pi C^3(1-\alpha)^3}{3\sigma^3} - \frac{2tC}{\sigma} \geq -\frac{32\pi C^3}{3\sigma^3}.$$

Assuming $\sigma\alpha(1-\alpha) \geq 2C^{3/2} \exp(\sqrt{2\pi}C)$, then $\sigma^2\alpha^2(1-\alpha)^2 \geq 4C^3 \exp(2\sqrt{2\pi}C)$, so we have

$$\frac{32\pi C^3}{3\sigma^3} \leq \frac{32\pi C^3 \alpha^3(1-\alpha)^3}{3\sigma^3 \alpha^3(1-\alpha)^3} \leq \frac{\pi C^3}{6\sigma^3 \alpha^3(1-\alpha)^3} \leq \frac{\pi \exp(-2\sqrt{2\pi}C)}{24\sigma\alpha(1-\alpha)} \leq \frac{\exp(-2\sqrt{2\pi}C)}{6\sigma\alpha(1-\alpha)}. \quad (22)$$

We combine (21) and (22) and have,

$$\mathbb{E}\left[g_\alpha\left(\log \frac{\varphi_0(Y)}{\varphi_1(Y)}\right)\right] \geq \frac{\exp(-2\sqrt{2\pi}C)}{5\sigma\alpha(1-\alpha)} - \frac{\exp(-2\sqrt{2\pi}C)}{6\sigma\alpha(1-\alpha)} = \frac{\exp(-2\sqrt{2\pi}C)}{30\sigma\alpha(1-\alpha)}.$$

C. Proof of Corollary 1

It is easy to check that

$$D_\alpha(\tilde{\varphi}_0^{\otimes n} \|\tilde{\varphi}_1^{\otimes n}) = -\log \int (\tilde{\varphi}_0^{\otimes n})^{1-\alpha} (\tilde{\varphi}_1^{\otimes n})^{\alpha} = -\log \prod_{j=1}^n \int \tilde{\varphi}_0^{1-\alpha} \tilde{\varphi}_1^{\alpha} = -n \log \int \tilde{\varphi}_0^{1-\alpha} \tilde{\varphi}_1^{\alpha} = n D_\alpha(\tilde{\varphi}_0 \|\tilde{\varphi}_1).$$

Now it is sufficient to check the conditions of Theorem 1 and apply the theorem. For fixed $\tilde{\varphi}_0$ and $\tilde{\varphi}_1$, $D''_{\alpha^*}(\tilde{\varphi}_0 \|\tilde{\varphi}_1)$, $\mathbb{E}[|Z_j|^3]$ and α^* in Theorem 1 are constants, so $\alpha^*(1-\alpha^*) \asymp 1$

$$\sum_{j=1}^n \mathbb{E}[|Z_j|^3] = n \mathbb{E}[|Z_1|^3] \lesssim -n D''_{\alpha^*}(\tilde{\varphi}_0 \|\tilde{\varphi}_1) = -D''_{\alpha^*}(\tilde{\varphi}_0^{\otimes n} \|\tilde{\varphi}_1^{\otimes n}).$$

Finally, $-n D''_{\alpha^*}(\tilde{\varphi}_0 \|\tilde{\varphi}_1)$ is large enough as n increases. Therefore, all assumptions in Theorem 1 are satisfied and the proof is complete.

D. Proof of (12)

We have

$$\begin{aligned} \frac{d^2}{d\alpha^2} \log \int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^{\alpha} d\mu &= \frac{d}{d\alpha} \frac{\int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^{\alpha} \log \frac{\varphi_1}{\varphi_0} d\mu}{\int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^{\alpha} d\mu} \\ &= \frac{e^{-D_\alpha(\varphi_0 \|\varphi_1)} \int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^{\alpha} \left(\log \frac{\varphi_1}{\varphi_0}\right)^2 d\mu - \left(\int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^{\alpha} \log \frac{\varphi_1}{\varphi_0} d\mu\right)^2}{e^{-2D_\alpha(\varphi_0 \|\varphi_1)}}. \end{aligned}$$

By Lemma 3, evaluating at α^* , we have $\int_{\Omega} \varphi_0^{1-\alpha^*} \varphi_1^{\alpha^*} \log \frac{\varphi_1}{\varphi_0} d\mu = 0$. Thus

$$\frac{d^2}{d\alpha^2} \log \int_{\Omega} \varphi_0^{1-\alpha} \varphi_1^{\alpha} d\mu \Big|_{\alpha=\alpha^*} = e^{D_{\alpha^*}(\varphi_0 \|\varphi_1)} \int_{\Omega} \varphi_0^{1-\alpha^*} \varphi_1^{\alpha^*} \left(\log \frac{\varphi_1}{\varphi_0}\right)^2 d\mu = \sum_{j=1}^n \mathbb{E}[Z_j^2].$$

By Lemma 3, $\sum_{j=1}^n \mathbb{E}[Z_j] = 0$, so $\sum_{j=1}^n \mathbb{E}[Z_j^2] = \sum_{j=1}^n \text{var}[Z_j]$.

APPENDIX B PROOF OF SECTION V

A. Proof of Lemma 1

With help of this representation, we have for any projection operator E_1 ,

$$\text{Tr}(\rho_0 E_1) = \text{Tr}\left(\left(\sum_{i=1}^n \lambda_i |x_i\rangle\langle x_i|\right) E_1\right) = \sum_{i=1}^d \lambda_i \langle x_i | E_1 x_i \rangle = \sum_{i=1}^d \lambda_i \|E_1 x_i\|^2 = \sum_{i=1}^d \lambda_i \sum_{j=1}^d |\langle E_1 x_i | y_j \rangle|^2.$$

where the last equality is by Parseval's identity. In the same manner,

$$\text{Tr}(\rho_1(I - E_1)) = \sum_{i=1}^d \gamma_i \sum_{j=1}^d |\langle (I - E_1)x_i | y_j \rangle|^2.$$

Thus, we have

$$\begin{aligned} \frac{1}{2} (\text{Tr}(\rho_0 E_1) + \text{Tr}(\rho_1(1 - E_1))) &= \frac{1}{2} \sum_{i,j=1}^d (\lambda_i |E_1 \langle x_i | y_j \rangle|^2 + \gamma_j |(I - E_1) \langle x_i | y_j \rangle|^2) \\ &\geq \frac{1}{2} \sum_{i,j=1}^d \min(\lambda_i, \gamma_j) \left(\frac{|E_1 \langle x_i | y_j \rangle| + |(I - E_1) \langle x_i | y_j \rangle|}{2} \right)^2 \\ &\geq \frac{1}{4} \sum_{i,j=1}^d \min\{\lambda_i, \gamma_j\} |\langle x_i | y_j \rangle|^2. \end{aligned}$$

B. Proof of Lemma 2

By independence of X_k 's, we have

$$\mathbb{P}((X_1, \dots, X_n) = ((i_1, j_1), \dots, (i_n, j_n))) = \prod_{k=1}^n A_{i_k j_k}^{(k)}.$$

Hence the lemma follows from the definition of tensor product.

C. Proof of Corollary 2

We observe that

$$\text{Tr}((\tilde{\rho}_0^{\otimes n})^{1-\alpha} (\tilde{\rho}_1^{\otimes n})^\alpha) = \text{Tr}((\tilde{\rho}_0^{1-\alpha} \tilde{\rho}_1^\alpha)^{\otimes n}) = \text{Tr}(\rho_0^{1-\alpha} \rho_1^\alpha)^n.$$

Now it is sufficient to check the conditions of Theorem 2. For fixed $\tilde{\rho}_0$ and $\tilde{\rho}_1$, $D''_{\alpha^*}(\tilde{\rho}_0 \| \tilde{\rho}_1)$, $\mathbb{E}[|Z_j|^3]$ and α^* in Theorem 2 are constants, so $\alpha^*(1 - \alpha^*) \asymp 1$

$$\sum_{j=1}^n \mathbb{E}[|Z_j|^3] = n \mathbb{E}[|Z_1|^3] \lesssim -n D''_{\alpha^*}(\tilde{\varphi}_0 \| \tilde{\varphi}_1) = -D''_{\alpha^*}(\tilde{\varphi}_0^{\otimes n} \| \tilde{\varphi}_1^{\otimes n}).$$

Finally, $-n D''_{\alpha^*}(\tilde{\varphi}_0 \| \tilde{\varphi}_1)$ is large enough as n increases. Therefore, all assumptions in Theorem 1 are satisfied and the proof is complete.