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Supplementary material for ‘On the random- versus systematic-scan sampler choice’

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SUMMARY

This supplement contains proofs of Proposition 1 and Lemma 2, as well as three lemmas.

Proof of Proposition 1. Let $f \in L_0^2(\mathbf{X}, \pi)$. Then for $n \geq 1$,

$$E_\pi \left\{ \sum_{i=0}^{n-1} f(X_i) \sum_{j=0}^{n-1} f(X_j) \right\} = n E_\pi \{ f^2(X_0) \} + 2 \sum_{0 \leq i < j \leq n-1} E_\pi \{ f(X_i) f(X_j) \},$$

and we rewrite the second term as

$$\begin{aligned} \sum_{0 \leq i < j \leq n-1} E_\pi \{ f(X_i) f(X_j) \} &= \sum_{0 \leq i < j \leq n-1} E_\pi \{ f(X_i) \Pi_{\sigma^{i:j-1}(1)} f(X_i) \} \\ &= \sum_{0 \leq i < j \leq n-1} \langle f, \Pi_{\sigma^{i:j-1}(1)} f \rangle_\pi \\ &= \sum_{0 \leq i < j \leq n-1} \langle f, \Pi_{\sigma^{i:j-i-1+i}(1)} f \rangle_\pi \\ &= \sum_{0 \leq i < n-1} \sum_{m=1}^{n-1-i} \langle f, \Pi_{\sigma^{i:m-1+i}(1)} f \rangle_\pi. \end{aligned} \tag{10}$$

Now we have

$$\begin{aligned} n^{-1} \sum_{0 < i < n-1} \sum_{m=1}^{n-1-i} \langle f, \Pi_{\sigma^{i:m-1+i}(1)} f \rangle_\pi \\ = \sum_{q=1}^k \frac{\lfloor (n-2-q)/k \rfloor}{n} \frac{1}{\lfloor (n-2-q)/k \rfloor} \sum_{p \in P_{k,q}} \sum_{m=1}^{n-1-pk-q} \langle f, \Pi_{\sigma^{0:m-1}(q)} f \rangle_\pi, \end{aligned} \tag{15}$$

where $P_{k,q} = \{m \in \mathbb{N} : 0 < mk + q < n-1\}$. The term $i = 0$ is treated similarly. We conclude by letting $n \rightarrow \infty$ and using a Cesàro sum argument for each $q \in \{1, \dots, k\}$. $\square$

We let $\bar{\text{pr}}(\cdot)$ denote the probability distribution of the Markov chain defined by $\mathcal{T}$.

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LEMMA 4. For $m \geq 0$ and $A_1, \dots, A_m \in \mathcal{X}^m$,

$$\bar{\text{pr}} \left(X_0^{\sigma^0(1)} \in A_0, X_1^{\sigma^1(1)} \in A_1, \dots, X_m^{\sigma^m(1)} \in A_m \right) = \text{pr} (X_0 \in A_0, X_1 \in A_1, \dots, X_m \in A_m) .$$

Proof. By construction for $A \in \mathcal{X}$, $\bar{\text{pr}}(X_0^{\sigma^0(1)} \in A) = \pi(A) = \text{pr}(X_0 \in A)$ and for $i \geq 1$, component $\sigma^i(1)$ is generated by kernel $\Pi_{\sigma^{i-1}(1)}$ and with

$$\mathcal{G}_i = \sigma \left\{ (X_m^{(1)}, \dots, X_m^{(k)}), 0 \leq m \leq i \right\}$$

we have

$$\bar{\text{pr}}(X_i^{\sigma^i(1)} \in A \mid \mathcal{G}_{i-1}) = \Pi_{\sigma^{i-1}(1)}(X_{i-1}^{\sigma^{i-1}(1)}, A)$$

25 from which the result follows. □

Proof of Lemma 2. We have

$$\begin{aligned} \langle f, [I - \lambda \Xi]^{-1} f \rangle &= \langle (I - \lambda \Xi)(I - \lambda \Xi)^{-1} f, (I - \lambda \Xi)^{-1} f \rangle \\ &= \langle (I - \lambda S)(I - \lambda \Xi)^{-1} f, (I - \lambda \Xi)^{-1} f \rangle \\ &= \sup_{h \in \mathcal{H}} 2 \langle (I - \lambda \Xi)^{-1} f, h \rangle - \langle h, (I - \lambda S)^{-1} h \rangle \\ 30 \quad &= \sup_{h \in \mathcal{H}} 2 \langle f, (I - \lambda \Xi^*)^{-1} h \rangle - \langle h, (I - \lambda S)^{-1} h \rangle \\ &= \sup_{g \in \mathcal{H}} 2 \langle f, g \rangle - \langle (I - \lambda \Xi^*) g, (I - \lambda S)^{-1} (I - \lambda \Xi^*) g \rangle \\ &= \sup_{g \in \mathcal{H}} 2 \langle f, g \rangle - \langle g, (I - \lambda S) g \rangle - \langle \lambda A g, (I - \lambda S)^{-1} \lambda A g \rangle \end{aligned}$$

where we have used that $\Xi = S + A$, $\Xi^* = S - A$, that for any $g \in \mathcal{H}$, $\langle g, A g \rangle = -\langle A g, g \rangle = 0$, Lemma 5 for the self-adjoint operator $(I - \lambda S)$, $\{(I - \lambda \Xi)^{-1}\}^* = (I - \lambda \Xi^*)^{-1}$, set $g = (I - \lambda \Xi^*)^{-1} h$ and again used the property $\langle g, A g \rangle = 0$. From Lemma 5 the supremum on the
35 third line is attained for $\hat{h} = (I - \lambda S)(I - \lambda \Xi)^{-1} f$, which translates into $\hat{g} = (I - \lambda \Xi^*)^{-1} \hat{h}$ on the last line. Consequently using again Lemma 5 for the operator $I - \lambda S$ we deduce that

$$\langle f, (I - \lambda \Xi)^{-1} f \rangle \leq \langle f, (I - \lambda S)^{-1} f \rangle - \lambda^2 \langle A \hat{g}, (I - \lambda S)^{-1} A \hat{g} \rangle.$$

□

The following provides a useful variational representation of the quadratic form of the inverse
40 of a positive self-adjoint operators, attributed to Bellman, and used for example by Caracciolo et al. (1990).

LEMMA 5. Let Ξ be a self-adjoint operator on a Hilbert space $\mathcal{H}$, satisfying $\langle f, \Xi f \rangle \geq 0$ for all $f \in \mathcal{H}$ and such that the inverse Ξ^{-1} exists. Then

$$\langle f, \Xi^{-1} f \rangle = \sup_{g \in \mathcal{H}} (2 \langle f, g \rangle - \langle g, \Xi g \rangle),$$

where the supremum is attained with $g = \Xi^{-1} f$.

45 The operator T is not self-adjoint, but one can easily determine the expression for its adjoint T^* in terms of $\mathfrak{S}$ and Δ or T . Visualizing T as a block diagonal matrix may be helpful.

LEMMA 6. Let T , Δ and $\mathfrak{S}$ be as defined in §2 and §3. Then

1. the adjoint of $\mathfrak{S}$ is $\mathfrak{S}^* = \mathfrak{S}^{-1}$,

2. $\Delta^* = \Delta$, that is Δ is self-adjoint,
3. $T = \Delta \circ \mathfrak{S}$ and the adjoint of T is $T^* = \mathfrak{S}^{-1} \circ \Delta = \mathfrak{S}^{-1} \circ T \circ \mathfrak{S}^{-1}$.

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Proof. The first statement follows from

$$\langle \varphi, \mathfrak{S}\varphi \rangle = \sum_{i=1}^k \langle \varphi_i, \varphi_{\sigma(i)} \rangle_{\pi} = \sum_{j=1}^k \langle \varphi_{\sigma^{-1}(j)}, \varphi_j \rangle_{\pi} = \langle \mathfrak{S}^{-1}\varphi, \varphi \rangle.$$

The second statement is direct while the third statement follows from the general fact that $T^* = \mathfrak{S}^* \circ \Delta^*$ followed by an application of the first two statements of the lemma. $\square$

BIBLIOGRAPHY

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CARACCILO, S., PELISSETTO, A. & SOKAL, A. D. (1990). Nonlocal Monte Carlo algorithm for self-avoiding walks with fixed endpoints. *J. Stat. Phys.* **60**, 1–53.