

Supporting Information for

A Generalized Transition State Theory Treatment of Water-Assisted Proton Transport Processes in Proteins

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1. Derivation of Transition State Rate Constant in Terms of an n-Dimensional PMF

The general expression of the rate constant when assuming the dividing surface of an N-dimensional PMF is defined as $\xi_1 = c$, is given by

$$k_{\text{TST}} = \frac{1}{Q^{\text{R}}} \int \frac{d\mathbf{q} d\mathbf{p}_{\text{q}}}{h^{3N}} \int d\boldsymbol{\xi} \int d\mathbf{p}_{\boldsymbol{\xi}} \delta(\xi_1 - c) e^{-\beta H} \Theta(\dot{\xi}_1) \dot{\xi}_1 \quad (\text{S1})$$

Here, $\boldsymbol{\xi} = (\xi)_i$ is the CV vector with dimension n , and ξ_1 is the PT CV related to the excess proton CEC., while $\mathbf{p}_{\boldsymbol{\xi}}$ is the conjugate momentum vector of $\boldsymbol{\xi}$.

The Hamiltonian of the system can be expressed as a function of the $\boldsymbol{\xi}$ and $\mathbf{p}_{\boldsymbol{\xi}}$ vectors as

$$H = \frac{1}{2} \mathbf{p}_{\boldsymbol{\xi}}^{\text{T}} \mathbf{Z}_{\boldsymbol{\xi}} \mathbf{p}_{\boldsymbol{\xi}} + \frac{1}{2} \mathbf{p}_{\mathbf{q}}^{\text{T}} \mathbf{Z}_{\mathbf{q}} \mathbf{p}_{\mathbf{q}} + V(\boldsymbol{\xi}, \mathbf{q}) \quad (\text{S2})$$

where $\mathbf{Z}_{\boldsymbol{\xi}}$ and $\mathbf{Z}_{\mathbf{q}}$ are the mass matrices for the CVs from the degrees of freedom. The definition of $\mathbf{Z}_{\boldsymbol{\xi}}$ is given by Eq. (3).

The rate expression now becomes

$$k_{\text{TST}} = \frac{1}{Q^{\text{R}}} \int \frac{d\mathbf{q} d\mathbf{p}_{\mathbf{q}}}{h^{3N}} \int d\boldsymbol{\xi} \delta(\xi_1 - c) e^{-\beta(\frac{1}{2} \mathbf{p}_{\mathbf{q}}^{\text{T}} \mathbf{Z}_{\mathbf{q}} \mathbf{p}_{\mathbf{q}} + V(\boldsymbol{\xi}, \mathbf{q}))} \int d\mathbf{p}_{\boldsymbol{\xi}} \Theta(\dot{\xi}_1) \dot{\xi}_1 e^{-\beta \frac{1}{2} \mathbf{p}_{\boldsymbol{\xi}}^{\text{T}} \mathbf{Z}_{\boldsymbol{\xi}} \mathbf{p}_{\boldsymbol{\xi}}} \quad (\text{S3})$$

Note that by incorporating

$$\dot{\boldsymbol{\xi}} = \frac{\partial H}{\partial \mathbf{p}_{\boldsymbol{\xi}}} = \mathbf{Z}_{\boldsymbol{\xi}} \mathbf{p}_{\boldsymbol{\xi}} \quad (\text{S4})$$

one can transform the representation from $\mathbf{p}_{\boldsymbol{\xi}}$ to $\dot{\boldsymbol{\xi}}$,

The transition matrix $\mathbf{A}_{\boldsymbol{\xi}}$ is the inverse matrix of $\mathbf{Z}_{\boldsymbol{\xi}}$, and is expressed in the following form

$$\mathbf{A}_n = \mathbf{Z}_{\boldsymbol{\xi}}^{-1} = \begin{pmatrix} \mathbf{a} & \mathbf{b}^{\text{T}} \\ \mathbf{b} & \mathbf{A}_{n-1} \end{pmatrix} \quad (\text{S5})$$

The kinetic term in the exponential now becomes

$$\frac{1}{2} \mathbf{p}_{\boldsymbol{\xi}}^{\text{T}} \mathbf{Z}_{\boldsymbol{\xi}} \mathbf{p}_{\boldsymbol{\xi}} = \frac{1}{2} \dot{\boldsymbol{\xi}}^{\text{T}} \mathbf{Z}_{\boldsymbol{\xi}}^{-1} \dot{\boldsymbol{\xi}} \quad (\text{S6})$$

From Eq. (S4), (S5) and (S6), we have

$$\int d\mathbf{p}_{\boldsymbol{\xi}} \Theta(\dot{\xi}_1) \dot{\xi}_1 e^{-\beta \frac{1}{2} \mathbf{p}_{\boldsymbol{\xi}}^{\text{T}} \mathbf{Z}_{\boldsymbol{\xi}} \mathbf{p}_{\boldsymbol{\xi}}} = |\mathbf{A}_{\boldsymbol{\xi}}| \int d\dot{\xi}_n \dots \int d\dot{\xi}_2 \int_0^{\infty} d\dot{\xi}_1 \dot{\xi}_1 e^{-\frac{\beta}{2} \dot{\boldsymbol{\xi}}^{\text{T}} \mathbf{Z}_{\boldsymbol{\xi}}^{-1} \dot{\boldsymbol{\xi}}} \quad (\text{S7})$$

Using the Gaussian integral formula,

$$\int dx_n \dots \int dx_2 \int_0^{\infty} dx_1 x_1 \exp -\frac{1}{2} (\mathbf{x}^{\text{T}} \mathbf{A} \mathbf{x}) = (\sqrt{2\pi})^{n-1} \frac{\sqrt{|\mathbf{A}_{n-1}|}}{|\mathbf{A}_n|} \quad (\text{S8})$$

the integration over the velocity representation in Eq. (S7) becomes

$$\int d\mathbf{p}_{\boldsymbol{\xi}} \Theta(\dot{\xi}_1) \dot{\xi}_1 e^{-\beta \frac{1}{2} \mathbf{p}_{\boldsymbol{\xi}}^{\text{T}} \mathbf{Z}_{\boldsymbol{\xi}} \mathbf{p}_{\boldsymbol{\xi}}} = |\mathbf{A}_n| (\sqrt{2\pi})^{n-1} \frac{\sqrt{\beta^{n-1} |\mathbf{A}_{n-1}|}}{\beta^n |\mathbf{A}_n|} \quad (\text{S9})$$

$$= \left(\sqrt{\frac{2\pi}{\beta}} \right)^{n-1} \frac{1}{\beta} \sqrt{|\mathbf{A}_{n-1}|}$$

By then incorporating the Gaussian formula

$$\int d\mathbf{p}_\xi e^{-\beta \frac{1}{2} \mathbf{p}_\xi^T \mathbf{z}_\xi \mathbf{p}_\xi} = \sqrt{\frac{(2\pi)^n |\mathbf{A}_n|}{\beta^n}} \quad (\text{S10})$$

the integral over the momentum on the left-hand side of Eq. (S8) can be expressed as

$$\begin{aligned} \int d\mathbf{p}_\xi \Theta(\xi_1) \dot{\xi}_1 e^{-\beta \frac{1}{2} \mathbf{p}_\xi^T \mathbf{z}_\xi \mathbf{p}_\xi} &= \frac{\left(\sqrt{\frac{2\pi}{\beta}} \right)^{n-1} \frac{1}{\beta} \sqrt{\beta^n \sqrt{|\mathbf{A}_{n-1}|}}}{(\sqrt{2\pi})^n \sqrt{|\mathbf{A}_n|}} \int d\mathbf{p}_\xi e^{-\beta \frac{1}{2} \mathbf{p}_\xi^T \mathbf{z}_\xi \mathbf{p}_\xi} \\ &= \frac{1}{\sqrt{2\pi\beta}} \frac{\sqrt{|\mathbf{A}_{n-1}|}}{\sqrt{|\mathbf{A}_n|}} \int d\mathbf{p}_\xi e^{-\beta \frac{1}{2} \mathbf{p}_\xi^T \mathbf{z}_\xi \mathbf{p}_\xi} \end{aligned} \quad (\text{S11})$$

By applying the inverse operator on both sides of Eq. (S5)

$$\mathbf{z}_\xi = \mathbf{A}_n^{-1} \quad (\text{S12})$$

and due to the fact that the specific element in an inverse matrix can be explicitly expressed with the matrix

$$(\mathbf{z}_\xi)_{11} = \mathbf{z}_{11} = \frac{|\mathbf{A}_{n-1}|}{|\mathbf{A}_n|} \quad (\text{S13})$$

we can substitute the integrand to obtain the same integration results in the rate expression as

$$\begin{aligned} k_{\text{TST}} &= \frac{1}{Q^R} \int \frac{d\mathbf{q} d\mathbf{p}_q}{h^{3N}} \int d\xi \delta(\xi_1 - c) e^{-\beta \left(\frac{1}{2} \mathbf{p}_\xi^T \mathbf{z}_\xi \mathbf{p}_\xi + V(\xi, \mathbf{q}) \right)} \frac{1}{\sqrt{2\pi\beta}} \sqrt{Z_{11}} \int d\mathbf{p}_\xi e^{-\beta \frac{1}{2} \mathbf{p}_\xi^T \mathbf{z}_\xi \mathbf{p}_\xi} \\ &= \frac{1}{Q^R \sqrt{2\pi\beta}} \int \frac{d\mathbf{q} d\mathbf{p}_q d\xi d\mathbf{p}_\xi}{h^{3N}} \delta(\xi_1 - c) \sqrt{Z_{11}} e^{-\beta H} \\ &= \frac{1}{Q^R \sqrt{2\pi\beta}} \frac{\int \frac{d\mathbf{q} d\mathbf{p}_q d\xi d\mathbf{p}_\xi}{h^{3N}} \delta(\xi_1 - c) \sqrt{Z_{11}} e^{-\beta H}}{\int \frac{d\mathbf{q} d\mathbf{p}_q d\xi d\mathbf{p}_\xi}{h^{3N}} \delta(\xi_1 - c) e^{-\beta H}} \\ &\quad \times \int \frac{d\mathbf{q} d\mathbf{p}_q d\xi d\mathbf{p}_\xi}{h^{3N}} \delta(\xi_1 - c) e^{-\beta H} \\ &= \frac{1}{Q^R \sqrt{2\pi\beta}} \langle \sqrt{Z_{11}} \rangle_{\xi_1=c} \int \frac{d\mathbf{q} d\mathbf{p}_q d\xi d\mathbf{p}_\xi}{h^{3N}} \delta(\xi_1 - c) e^{-\beta H} \\ &= \frac{1}{\sqrt{2\pi\beta}} \langle \sqrt{Z_{11}} \rangle_{\xi_1=c} \frac{\int d\xi e^{-\beta W(\xi_1=c)}}{Q^R} \end{aligned} \quad (\text{S14})$$

where $Z_{11} = \sum_{k=1}^{3N} \frac{1}{M_k} \left(\frac{\partial \xi_1}{\partial r_k} \right)^2$ is the inverse value of the effective mass of the first CV.

2. Calculation of the ensemble average of $\langle \sqrt{Z_\xi} \rangle$ among transition states in terms of a 2D PMF

1) Review of WHAM scheme to obtain a 1D PMF for the transition ridge $\xi_1 = c$ in the 2D PMF

First, let us review WHAM. WHAM here attempts to maximize the likelihood of probability summing over window k and bin i of ξ_2 axis at transition ridge $\xi_1 = c$.

Statement 1:

$$\text{Minimize } -\ln L_{WHAM} = -\sum_k \sum_i \ln(\pi_i^k)^{N_i^k} = -\sum_k \sum_i N_i^k \ln(f^k \gamma_i^k \pi_i) \quad \text{so that } \sum_i \pi_i = 1.$$

Here, $\pi_i = P(\xi_2 = (\xi_2)_i | \xi_1 = c)$ is the conditional probability of ξ_2 falling into bin i out of whole transition ridge $\xi_1 = c$. π_i^k is the biased probability at simulation condition k that falls into bin i . π_i and π_i^k are related by the known umbrella potential added, shown as $\pi_i^k = f^k \gamma_i^k \pi_i$. Here, γ_i^k is the constant bias factor caused by adding umbrella window k . $\gamma_i^k = e^{-\kappa((\xi_2)_i - (\text{center}^k))^2}$. $f^k = \frac{1}{\sum_l \pi_l \gamma_l^k}$ is the normalization factor that ensures $\sum_i \pi_i^k = 1$.

Statement 2:

$$\max_v \left[\min_{\pi_i} \left(-\ln L_{WHAM} + v \cdot \left(\sum_i \pi_i - 1 \right) \right) \right]$$

Statement 1 is equal to statement 2 due to Lagrange Duality theory.

To solve $\min_{\pi_i} (-\ln L_{WHAM} + v \cdot (\sum_i \pi_i - 1))$, we do derivative over π_i for all possible i .

$$\frac{\partial}{\partial \pi_i} \left(-\sum_k \sum_j N_j^k \ln \left(\frac{\gamma_j^k \pi_j}{\sum_l \pi_l \gamma_l^k} \right) + v \cdot \left(\sum_j \pi_j - 1 \right) \right) = 0 \quad \forall i \quad (S15)$$

$$\frac{\partial}{\partial \pi_i} \left(-\sum_k \sum_j \left[N_j^k \cdot \ln \gamma_j^k \pi_j - N_j^k \cdot \ln \sum_l \pi_l \gamma_l^k \right] + v \cdot \sum_j \pi_j - v \right) = 0 \quad \forall i \quad (S16)$$

$$-\sum_k \frac{N_i^k}{\pi_i} + \sum_k \left(\left(\sum_j N_j^k \right) \cdot \frac{\gamma_i^k}{\sum_l \pi_l \gamma_l^k} \right) + v = 0 \quad \forall i \quad (S17)$$

$$v = \frac{1}{\pi_i} \sum_k N_i^k - \sum_k \left(\left(\sum_j N_j^k \right) \cdot \gamma_i^k f^k \right) \quad (S18)$$

This equation stands for all π_i . Then we need to get v that maximize the following expression

$$\max_v \left[- \sum_k \sum_j N_j^k \ln \left(\frac{\gamma_j^k \pi_j}{\sum_l \pi_l \gamma_l^k} \right) + v \cdot \left(\sum_j \pi_j - 1 \right) \right] \quad (\text{S19})$$

It's obvious to get that $v = 0$. The solution is shown below,

$$\pi_i = \frac{\sum_k N_i^k}{\sum_k \left((\sum_j N_j^k) \cdot \gamma_i^k f^k \right)} \quad (\text{S20})$$

$$f^k = \frac{1}{\sum_l \pi_l \gamma_l^k}$$

2) Calculation of $\langle A \rangle_{\xi_1=c}$ on the 2D-PMF

Now we want to calculate the ensemble average for $\langle A \rangle_{\xi_1=c}$, where the physical quantity happens to be $\sqrt{Z_{\xi_{11}}}$.

$$\langle A \rangle_{\xi_1=c} = \sum_i \left(P(\xi_2 = (\xi_2)_i \mid \xi_1 = c) \cdot \langle A \rangle_{i, \xi_1=c} \right) \quad (\text{S21})$$

$$\text{Here, } P(\xi_2 = (\xi_2)_i \mid \xi_1 = c) = \pi_i = \frac{e^{-\beta E_i}}{\sum_j e^{-\beta E_j}} = \frac{\sum_k N_i^k}{\sum_k \left((\sum_j N_j^k) \cdot \gamma_i^k f^k \right)}.$$

Notice that $\langle A \rangle_i^l(\text{unbiased})$ is the unbiased, reweighted ensemble average derived from quantity A sampled at window l and falls into bin i , $\langle A \rangle_i^l(\text{biased})$ is the biased ensemble average of quantity A directly sampled at window l and falls into bin i

$$\langle A \rangle_{i, \xi_1=c} = \sum_l \left(w_{il} \cdot \langle A \rangle_i^l(\text{unbiased}) \right) \quad (\text{S22})$$

$$\langle A \rangle_i^l(\text{unbiased}) = \frac{\sum_t A_i^j(t) e^{+\beta V_i^j(t)}}{\sum_t e^{+\beta V_i^j(t)}} = \frac{\int d\xi_1 \int d\xi_2 \int d\mathbf{r}^{3N-2} A_i^j(t) e^{-\beta(H+V_i^j)} e^{+\beta V_i^j}}{\int d\xi_1 \int d\xi_2 \int d\mathbf{r}^{3N-2} e^{-\beta(H+V_i^j)} e^{+\beta V_i^j}} \quad (\text{S23})$$

$$\langle A \rangle_i^l(\text{biased}) = \frac{1}{T} \sum_t A_i^j(t) = \frac{\int d\xi_1 \int d\xi_2 \int d\mathbf{r}^{3N-2} A_i^j(t) e^{-\beta(H+V_i^j)}}{\int d\xi_1 \int d\xi_2 \int d\mathbf{r}^{3N-2} e^{-\beta(H+V_i^j)}} \quad (\text{S24})$$

Since $V_i^j(t)$ is a time-independent potential, we have

$$\langle A \rangle_i^l(\text{unbiased}) = \langle A \rangle_i^l(\text{biased}) \quad (\text{S25})$$

Hence,

$$\langle A \rangle_{i, \xi_1=c} = \sum_l \left(w_{il} \cdot \langle A \rangle_i^l(\text{unbiased}) \right) = \sum_l \left(w_{il} \cdot \frac{1}{T} \sum_t A_i^l(t) \right) \quad (\text{S26})$$

To solve w_l , we look at the probability of bin i and its decomposition to the probability of this bin contributing from window l .

$$\pi_i = \sum_l \left(w_{il} \cdot \pi_i^l(\text{unbiased}) \right) = \sum_l \left(w_{il} \cdot \frac{\frac{N_i^l}{\gamma_i^l}}{\sum_j \frac{N_j^l}{\gamma_j^l}} \right) \quad (\text{S27})$$

From WHAM derivation in the first section, we know that

$$\pi_i = \frac{\sum_l N_i^l}{\sum_k \left((\sum_j N_j^k) \cdot \gamma_i^k f^k \right)} = \sum_l \frac{N_i^l}{\sum_k \left((\sum_j N_j^k) \cdot \gamma_i^k f^k \right)} \quad (\text{S28})$$

Hence,

$$w_{il} \cdot \frac{\frac{N_i^l}{\gamma_i^l}}{\sum_j \frac{N_j^l}{\gamma_j^l}} = \frac{N_i^l}{\sum_k \left((\sum_j N_j^k) \cdot \gamma_i^k f^k \right)} \quad (\text{S29})$$

$$w_{il} = \frac{\gamma_i^l \cdot \sum_j \frac{N_j^l}{\gamma_j^l}}{\sum_k \left((\sum_j N_j^k) \cdot \gamma_i^k f^k \right)} = \frac{\pi_i}{\sum_l N_i^l} \cdot \gamma_i^l \cdot \sum_j \frac{N_j^l}{\gamma_j^l} \quad (\text{S30})$$

To sum up,

$$\begin{aligned} \left\langle \sqrt{Z_\xi} \right\rangle_{\xi_1=c} &= \sum_i \left(P(\xi_2 = (\xi_2)_i \mid \xi_1 = c) \cdot \langle A \rangle_{i, \xi_1=c} \right) \\ &= \sum_i \left(\pi_i \cdot \sum_l \left(w_{il} \cdot \frac{1}{T} \sum_t A_i^l(t) \right) \right) \\ &= \sum_i \left[\frac{\pi_i^2}{\sum_l N_i^l} \cdot \sum_l \left(\gamma_i^l \cdot \left(\sum_j \frac{N_j^l}{\gamma_j^l} \right) \cdot \frac{1}{T} \sum_t \left(\sqrt{Z_\xi} \right)_i^l(t) \right) \right] \end{aligned} \quad (\text{S31})$$

3) Calculation of $\sqrt{Z_\xi}$ on a Given Frame of the Trajectory

The inverse value of effective mass of ξ is defined as

$$Z_\xi := \sum_{i=1}^{3N} \frac{1}{M_i} \left(\frac{\partial \xi}{\partial r_i} \right)^2 \quad (\text{S32})$$

where N is the number of atoms in the system, M_i is the mass of atom i , r_i is the Cartesian coordinate (either x , or y , or z dimension) of atom i .

$$\frac{\partial \xi}{\partial r_i} = \left(\frac{\partial \xi}{\partial V_{\text{bias}}} \right) \left(\frac{\partial V_{\text{bias}}}{\partial r_i} \right) \quad (\text{S33})$$

Since

$$\frac{\partial \xi}{\partial V_{\text{bias}}} = - \left(\frac{\partial V_{\text{bias}}(\xi)}{\partial \xi} \right)^{-1} = - \left(\frac{\partial \left[\frac{1}{2} \kappa (\xi - s)^2 \right]}{\partial \xi} \right)^{-1} = - \frac{1}{\kappa (\xi - s)} \quad (\text{S34})$$

$$\frac{\partial V_{\text{bias}}}{\partial r_i} = -F_{r_i}^{\text{bias}} \quad (\text{S35})$$

$F_{r_i}^{\text{bias}}$, the bias force on each atom along r_i direction, is provided by PUMED package, dump-force command.

So, we have

$$\frac{\partial \xi}{\partial r_i} = \left(\frac{\partial \xi}{\partial V_{\text{bias}}} \right) \left(\frac{\partial V_{\text{bias}}}{\partial r_i} \right) = \frac{F_{r_i}^{\text{bias}}}{\kappa(\xi - s)} \quad (\text{S36})$$

$$Z_\xi = \sum_{i=1}^{3N} \frac{1}{M_i} \left(\frac{\partial \xi}{\partial r_i} \right)^2 = \sum_{i=1}^{3N} \frac{1}{M_i} \left(\frac{F_{r_i}^{\text{bias}}}{\kappa(\xi - s)} \right)^2 \quad (\text{S37})$$

$$\sqrt{Z_\xi} = \sqrt{\sum_{i=1}^{3N} \frac{1}{M_i} \left(\frac{F_{r_i}^{\text{bias}}}{\kappa(\xi - s)} \right)^2} \quad (\text{S38})$$

3. The Proof of equipartition theorem $\langle \mathbf{Z}_{11} \rangle = \beta \langle \xi_1^2 \rangle$

$$\begin{aligned} \beta \langle \xi_1^2 \rangle &= \int \frac{d\mathbf{q} d\mathbf{p}_q}{h^{3N}} \int d\xi \beta e^{-\beta \left(\frac{1}{2} \mathbf{p}_q^T \mathbf{Z}_q \mathbf{p}_q + V(\xi, \mathbf{q}) \right)} \int d\mathbf{p}_\xi \xi_1^2 e^{-\beta \frac{1}{2} \xi^T \mathbf{Z}_\xi^{-1} \xi} \\ &= \int \frac{d\mathbf{q} d\mathbf{p}_q}{h^{3N}} \int d\xi \beta e^{-\beta \left(\frac{1}{2} \mathbf{p}_q^T \mathbf{Z}_q \mathbf{p}_q + V(\xi, \mathbf{q}) \right)} \int d\xi |\mathbf{A}| \xi_1^2 e^{-\beta \frac{1}{2} \xi^T \mathbf{Z}_\xi^{-1} \xi} \\ &= \int \frac{d\mathbf{q}}{h^{3N}} \int d\xi e^{-\beta V(\xi, \mathbf{q})} \sqrt{\frac{(2\pi)^n}{\beta |\mathbf{A}|}} |\mathbf{A}_{n-1}| \end{aligned} \quad (\text{S39})$$

The last line of Eq. (S39) employs the integral formula

$$\int d\mathbf{x} \mathbf{x}_1^2 e^{-\frac{\beta}{2} \mathbf{x}^T \mathbf{A} \mathbf{x}} = \sqrt{\frac{(2\pi)^n}{\beta^3 |\mathbf{A}|^3}} |\mathbf{A}_{n-1}| \quad (\text{S40})$$

Plugging in Eq. (S13), we get

$$\beta \langle \xi_1^2 \rangle = \int \frac{d\mathbf{q}}{h^{3N}} \int d\xi e^{-\beta V(\xi, \mathbf{q})} \sqrt{\frac{(2\pi)^n |\mathbf{A}|}{\beta}} \mathbf{Z}_{11} \quad (\text{S41})$$

Since $\mathbf{A} = \mathbf{Z}^{-1}$ and $|\mathbf{Z}^{-1}| = |\mathbf{Z}|^{-1}$

$$\beta \langle \xi_1^2 \rangle = \int \frac{d\mathbf{q}}{h^{3N}} \int d\xi e^{-\beta V(\xi, \mathbf{q})} \sqrt{\frac{(2\pi)^n}{\beta |\mathbf{Z}_\xi|}} \mathbf{Z}_{11} \quad (\text{S42})$$

Now, we look at the derivation of $\langle \mathbf{Z}_{11} \rangle$

$$\begin{aligned} \langle \mathbf{Z}_{11} \rangle &= \int \frac{d\mathbf{q}}{h^{3N}} \int d\xi \mathbf{Z}_{11} e^{-\beta \left(\frac{1}{2} \mathbf{p}_q^T \mathbf{Z}_q \mathbf{p}_q + V(\xi, \mathbf{q}) \right)} \int d\mathbf{p}_\xi e^{-\beta \frac{1}{2} \mathbf{p}_\xi^T \mathbf{Z}_\xi \mathbf{p}_\xi} \\ &= \int \frac{d\mathbf{q}}{h^{3N}} \int d\xi \mathbf{Z}_{11} \left(\sqrt{\frac{(2\pi)^n}{\beta |\mathbf{Z}_\xi|}} \right) e^{-\beta V(\xi, \mathbf{q})} \end{aligned} \quad (\text{S43})$$

Hence,

$$\langle \mathbf{Z}_{11} \rangle = \beta \langle \xi_1^2 \rangle$$