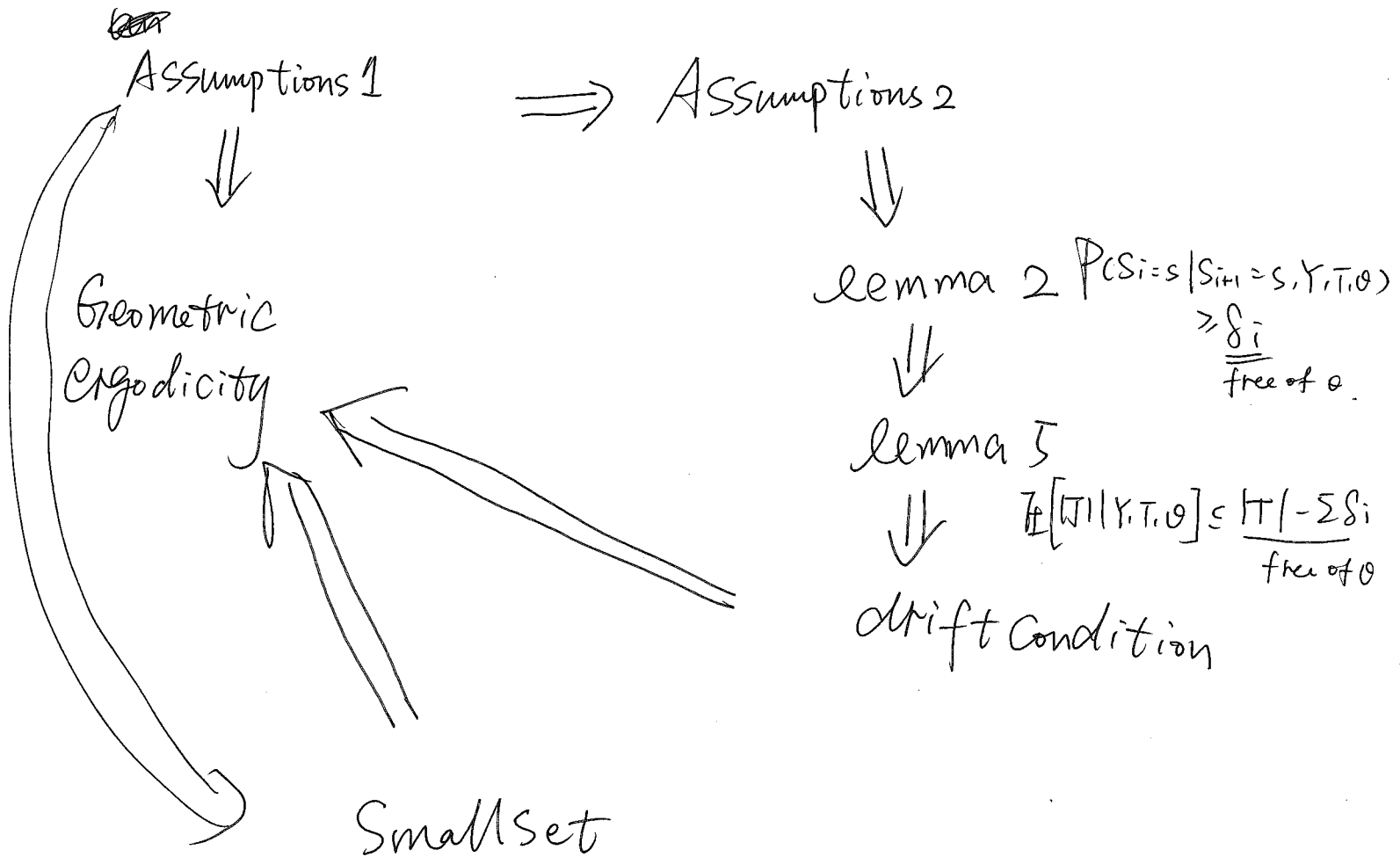


# Geometric Ergodicity Proof Structure



$$P(S|T, Y, \theta) \propto P(S, T|\theta) \cdot L(Y|T, S) \cdot P_0(\theta)$$

$$\propto V_0(S_0) g_0(S_0|\theta) \prod_{i=1}^N P(S_{i-1}, S_i|\theta) \cdot g_i(S_i|\theta) \cdot P_0(\theta)$$

$$g_i(S_i|\theta) = R(\theta) e^{-(T_{i+1}-T_i)R(\theta)} \times \prod_{j: T_i < \tau_j^{obs} < T_{i+1}} L_j(Y_j|S)$$

$$g_N(S|\theta) = e^{-(t_{max} - t_N)R(\theta)} \times \prod_{t_N < \tau_j^{obs} < t_{max}} L_j(Y_j|S)$$

Stochastic matrix  $P(\theta)$ :

$$P(\theta) = I + \frac{\alpha(\theta)}{R(\theta)}$$

$$P_{i-n_0=i}(S_{i-n_0}, S|\theta) = \sum_{\{S_{i-n_0+1} \dots S_{i-1}\}} \left( \prod_{\ell=i-n_0+1}^{i-1} P(S_{\ell-1}, S_{\ell}|\theta) \right) \cdot P(S_{i-1}, S_i|\theta)$$

$$P_{i-n_0=i}^g(S_{i-n_0}, S|\theta) = \sum_{\{S_{i-n_0+1} \dots S_{i-1}\}} \prod_{\ell=i-n_0+1}^{i-1} P(S_{\ell-1}, S_{\ell}|\theta) g_{\ell}(S_{\ell}|\theta)$$

~~$$P(S_{i-n_0}=S_{i-n_0}, S_i=S|Y, T, \theta) = C \cdot \sum_{S_0, S_1, \dots} V_0(S_0) g_0(S_0|\theta) \prod_{\ell=1}^{i-n_0-1} P(S_{\ell-1}, S_{\ell}|\theta) g_{\ell}(S_{\ell}|\theta)$$~~

$$P(S_{i-n_0}=S_{i-n_0}, S_i=S|Y, T, \theta) = C \cdot \sum_{\substack{S_0, S_1, \dots \\ \text{except for} \\ S_{i-n_0}, S_i}} V_0(S_0) g_0(S_0|\theta) \prod_{\ell=1}^{i-n_0-1} P(S_{\ell-1}, S_{\ell}|\theta) g_{\ell}(S_{\ell}|\theta) \cdot P_{i-n_0=i}^g(S_{i-n_0}, S|\theta) \cdot \left( \prod_{\ell=i}^N \dots \right)$$

# Assumptions 1.

i)  $\exists$  an irreducible matrix  $Q^{\min}$  s.t.

$$Q(\theta; s, s') \geq Q^{\min}(s, s') \quad \forall s \neq s' \in S$$

ii)  $\exists \gamma > 0$  s.t.

$$\sup_{\theta, s} \frac{Q(\theta; s)}{R(\theta)} \leq 1 - \gamma \quad \forall s \in S$$

iii)  $\exists \mu^{\max} < +\infty$  s.t.  $\sup_{\theta} R(\theta) \leq \mu^{\max}$

iv)  $q(\theta|\theta') \geq q_0(\theta)$ .  $q_0$  is the prior of  $\theta$ .

~~XXXXXXXXXX~~

~~XXXXXXXXXX~~

e.g.

$$\begin{pmatrix} -\alpha & \alpha & 0 & \dots \\ \beta & -\beta & \alpha & 0 & \dots \\ 0 & \beta & -\alpha & \beta & \alpha & 0 & \dots \end{pmatrix}$$

$$\lambda \begin{bmatrix} \alpha & \beta \end{bmatrix}$$

$\Leftrightarrow$

Assumption 1

# Assumptions 2.

for a fixed  $n_0, i$ , s.t.  $1 \leq n_0 \leq i \leq n-1$

i) for some  $\xi > 0$   $P_{i-n_0:i}(s_{i-n_0}, s; \theta) \geq \xi$  for  $\forall s_{i-n_0}, s \in S$

ii) for some  $\eta > 0$   $P_{i-n_0:i}(s, s; \theta) \geq \eta \quad \forall s \in S$

iii) for some  $g_{\ell}^{\min}$  and  $g_{\ell}^{\max}$ , we have  $g_{\ell}^{\min} \leq g_{\ell}(s; \theta) \leq g_{\ell}^{\max}$

$\forall s \in S \quad \ell \in [i-n_0+1, i]$

Assumption 1  $\Rightarrow$  Assumptions 2.

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$$i) \quad p^{\min} \triangleq I + \frac{Q^{\min}}{\mu^{\max}}$$

$\therefore Q^{\min}$  is irreducible and aperiodic

$\therefore p^{\min}$  is also irreducible and aperiodic

$\therefore \exists n_0 \in \mathbb{Z}^+ \quad \epsilon > 0 \quad \text{s.t.} \quad (p^{\min})^{n_0}(s, s') \geq \epsilon \quad \forall s, s' \in \mathcal{S}$

$$ii) \quad P(s, s; \theta) = 1 + \frac{Q(s, s; \theta)}{R(\theta)} \geq \eta$$

$$iii) \quad g_e^{\max} = g^{\max} = \mu^{\max}$$

For lower bound:

for  $g_e(s; \theta)$  including likelihood factors, there are at most  $k$ . We use 0 as lower bound.

$$\begin{aligned} \text{For the remaining, we have } g_e^{\min} &= g^{\min} \\ &= g^{\min} e^{-\mu^{\max}(\tau^{\max} - \tau^{\min})} \end{aligned}$$

$$g^{\min} \triangleq \min_S Q^{\min}(s) = \min_S \sum_{s' \neq s} Q^{\min}(s, s')$$

$$\Rightarrow R(\theta) \geq g^{\min} > 0. \quad \square$$

Lemma 2.

Under Assumptions 2

$$P(S_i = s | S_{i+1} = s, Y, T, \theta) \geq \delta_i \text{ and } \delta_i = \frac{1}{2} \gamma \cdot \prod_{l=i-n_0+1}^i \frac{g_l^{\min}}{g_l^{\max}}$$

proof:  $(\frac{a_i}{b_i} > c \Rightarrow \frac{\sum a_i}{\sum b_i} > c)$

We additionally Condition on  $S_{i-n_0} = S_{i-n_0}$

$S_i$  is free of  $\theta$ .

$$P(S_i = s | S_{i-n_0} = S_{i-n_0}, S_{i+1} = s, T, Y, \theta)$$

$$= \frac{P(S_i = s, S_{i+1} = s, S_{i-n_0} = S_{i-n_0} | T, Y, \theta)}{P(S_{i+1} = s, S_{i-n_0} = S_{i-n_0} | T, Y, \theta)}$$

$$= \frac{P_{i-n_0:i}^g(S_{i-n_0}, s; \theta) g_i(s; \theta) P_{\theta}(s, s; \theta)}{\sum_{s'} P_{i-n_0:i}^g(S_{i-n_0}, s'; \theta) \cdot g_i(s'; \theta) \cdot P(s', s; \theta)}$$

$$\geq \frac{\prod_{l=i-n_0+1}^i g_l^{\min}}{\prod_{l=i-n_0+1}^i g_l^{\max}} \cdot \frac{P_{i-n_0:i}(S_{i-n_0}, s; \theta) P(s, s; \theta)}{\sum_{s'} P_{i-n_0:i}(S_{i-n_0}, s'; \theta) P(s', s; \theta)} \geq \frac{1}{2} \gamma \prod_{l=i-n_0+1}^i \frac{g_l^{\min}}{g_l^{\max}} \quad \square$$

Notations.  $P(S | T, Y, \theta) \propto v_0(s_0) g_0(s_0) \prod_{i=1}^N P(S_{i-1}, S_i; \theta) \cdot g(S_i; \theta) \cdot P(\theta)$

$$P(\theta) = I + \frac{Q(\theta)}{R(\theta)}$$

Lemma 5  $|T| = n+1$

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if Assumptions 2 are true for  $i \in [n_0, n-1]$  then

$$\mathbb{E}[|J| \mid Y, T, \theta] \leq n+1 - \sum_{i=n_0}^{n-1} \delta_i$$

$$J = \{i \in [1, n] : S_{i-1} \neq S_i\} \cup \{0\}$$

Proof:  $|J| = 1 + \sum_{i=0}^{n-1} \mathbb{1}_{\{S_i \neq S_{i+1}\}}$

Apply Lemma 2 for  $\forall i \in [n_0, n-1]$

$$P(S_i = s \mid S_{i+1}, Y, T, \theta) \geq \delta_i$$

$$\mathbb{E}[\mathbb{1}_{\{S_i \neq S_{i+1}\}} \mid Y, T, \theta]$$

$$= \mathbb{E}[P(S_i \neq S_{i+1} \mid S_{i+1}, Y, T, \theta) \mid Y, T, \theta]$$

$$\leq 1 - \delta_i$$

for  $i < n_0$  we use upper bound 1.

$$\mathbb{E}[|J| \mid T, Y, \theta] \leq n+1 - \sum_{i=n_0}^{n-1} \delta_i = |T| - \sum_{i=n_0}^{n-1} \delta_i \quad \square$$

Drift Condition:

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Under Assumptions 1,  $\exists \delta > 0$  and  $C < +\infty$

$$\text{s.t. } \mathbb{E}[|J(x')| | X] \leq (1-\delta)|J(x)| + C$$

$$X \triangleq (\text{path}, 0)$$

$X$   
↓ step 1

$(S, T', 0)$

↓ step 2

$(S, T', 0')$

↓ step 3

$(S', T', 0')$

↓  
 $X'$

Proof:

in step 4, we add potential jumps  $V$ .

$$|V| \text{ has rate } \lambda(t^{\max} - t^{\min}, R(0)) - \int_{t^{\min}}^{t^{\max}} \lambda(s, 0) ds$$

$$\leq \mu^{\max} (t^{\max} - t^{\min}) \triangleq \mathcal{M}$$

$$\therefore \mathbb{E}[|T'| | X] \leq |J(x)| + \mathcal{M}$$

$$|T'| = n+1$$

use lemma 2

There are at most  $(k+1)n_0$  s.t.  $\delta_i = 0$

at least  $n - (k+1)n_0$   $\delta_i = \frac{1}{2} \eta \left( \frac{g^{\min}}{g^{\max}} \right)^{n_0} \triangleq \delta > 0$

$$\therefore \text{lemma 5} \Rightarrow \mathbb{E}[|J'| | T'] \leq (1-\delta)|T'| + (k+1)n_0\delta + \delta$$

$$\mathbb{E}[|J(x')| | X] = \mathbb{E}[\mathbb{E}[|J(x')| | T'] | X]$$

$$\leq (1-\delta)(|J(x)| + \mathcal{M}) + (k+1)n_0\delta + \delta \quad \square$$

# Small set condition.

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The small set  $\{x : |J(x)| \leq h\}$  is 1-small set for  $\forall h > 0$ .

Proof: First define a sequence

$$X \triangleq (\text{path}, \theta)$$

$$\text{path}^* = \begin{pmatrix} S^* \\ T^* \end{pmatrix} \begin{pmatrix} s_0^*, s_1^*, \dots, s_n^* \\ t_0^*, t_1^*, \dots, t_n^* \end{pmatrix} \quad n \geq k$$

Then the regeneration measure  $\tilde{P}(dx')$  is described as follows

$$T_i' \sim \text{unif}(t_{i-1}^*, t_i^*) \quad \text{for } i=1, 2, \dots, n$$

$$S_i' = S_i^* \quad \text{for } i=0, 1, 2, \dots, n$$

$$\theta' \sim P_\theta(\theta') \quad (\text{prior})$$

The random time vector  $(T_1', T_2', T_3', \dots, T_n')$  has the uniform distribution on the set  $T = \{(t_1, t_2, \dots, t_n) \mid t_{i-1}^* \leq t_i \leq t_i^*, \text{ for } i=1, 2, \dots, n\}$

The path is determined by  $(T', S')$

We first construct  $S^+ (S_1^+, S_2^+, S_3^+, \dots, S_k^+)$  s.t.  $\prod_{j=1}^k L_j(t_j | S_j^+) \triangleq L^+ > 0$

~~Under assumption 1~~

define  $S^* = (S_0^*, S_1^*, S_2^*, \dots, S_n^*)$  s.t.  $S^+$  is a ~~sub~~subsequence of  $S^*$   $(S_{i-1}^* \neq S_i^*)$

Also, ~~embed~~ embed  $t^{\text{obs}} = (t_1^{\text{obs}}, t_2^{\text{obs}}, \dots, t_k^{\text{obs}})$  in a longer sequence  $(t_1^*, t_2^*, \dots, t_n^*)$ .

$$\nu_0(S_0^*) \cdot \prod_{i=1}^{n-1} P(\theta; S_i^*, S_{i+1}^*) \geq \nu_0(S_0^*) \cdot \prod_{i=1}^{n-1} \frac{Q^{\min}(S_i^*, S_{i+1}^*)}{h^{\max}} \triangleq \beta_i^* > 0$$



fix  $X = (\text{path}, \theta)$  with  $|J(X)| \leq h$ .

There is another way in which the algorithm can be executed.

Note that  $\frac{Q(\theta; S)}{R(\theta)} \leq 1 - \eta \Rightarrow R(\theta) - Q(\theta; S) \geq \eta R(\theta) \geq \eta \underline{r}^{\min} \triangleq \varepsilon > 0$ .

in step 1. We independently sample two Poisson processes on the interval  $[t^{\min}, t^{\max}]$ , say  $V^0$  and  $V^{\text{rest}}$  with rate  $\varepsilon$  and  $R(\theta) - Q(\theta; S) - \varepsilon$

let  $V = V^0 \cup V^{\text{rest}}$

$$T' = J(X) \cup V$$

$$P(V^0 \in T) \triangleq \beta_0 > 0$$

$$P(V^{\text{rest}} = \emptyset) \geq e^{-r^{\max}(t^{\max} - t^{\min})} \triangleq \beta^{\text{rest}} > 0.$$

in step 2. We update  $\theta$  using MM. Assume  $q(\theta'|\theta) \geq \frac{K P_0(\theta')}{P_0(\theta)}$

i) propose  $\theta' \sim q(\theta'|\theta)$

ii) Accept  $\theta'$  with probability  $\alpha = \frac{q(\theta|\theta') P(\theta'|T'; Y)}{q(\theta'|\theta) P(\theta|T'; Y)} \wedge 1$   
 $\triangleq \underline{\alpha} \wedge 1$

$$P(\theta, T', Y) = \int_S P(\theta, S, T', Y) ds$$

$$= \int_S P_0(\theta) \cdot P(S, T'|\theta) \cdot P(Y|S, T') ds$$

$$\in [L_{\min} \cdot P_0(\theta') \cdot P(T'|\theta'), L_{\max} \cdot P_0(\theta) \cdot P(T'|\theta)]$$

$$\underline{\alpha} \geq \frac{q(\theta|\theta')}{q(\theta'|\theta)} \cdot \frac{L_{\min}}{L_{\max}} \cdot \frac{P(T'|\theta')}{P(T'|\theta)} \cdot \frac{P_0(\theta')}{P_0(\theta)} = (*)$$

$$\because P(T'|\theta) = R(\theta)^n (t^{\max} - t^{\min})^n e^{-R(\theta)(t^{\max} - t^{\min})}$$

$$\therefore (*) \frac{P_0(\theta) q(\theta|\theta')}{R(\theta) q(\theta'|\theta)} \cdot \frac{L_{\min}}{L_{\max}} \cdot \left(\frac{R(\theta')}{R(\theta)}\right)^n e^{-(R(\theta') - R(\theta))(t^{\max} - t^{\min})} \geq \frac{q(\theta|\theta')}{q(\theta'|\theta)} \cdot \frac{L_{\min}}{L_{\max}} \cdot \left(\frac{t^{\min}}{t^{\max}}\right)^n e^{-r^{\max}(t^{\max} - t^{\min})} \cdot \frac{P_0(\theta')}{P_0(\theta)}$$

$$q(\theta'|\theta) d\theta' \cdot \underline{\alpha} \geq q(\theta|\theta') \cdot \text{Const} \cdot \frac{P_0(\theta')}{P_0(\theta)} \geq K \cdot \text{Const} \cdot P_0(\theta') d\theta'$$

$$q(\theta'|\theta) d\theta' \cdot 1 \geq K P_0(\theta') d\theta'$$

Step 2:  $P(S' | T', \theta', \gamma)$

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$$\prod_{i=0}^{|T'|} g_i(S'_i) \leq (\mu_{\max})^{|T'|} L^{\max}$$

$$L^{\max} = \frac{k}{\prod_{j=1}^k} \max_s L_j(\gamma_j | s)$$

We can use rejection sampling ~~to~~ to sample  $S'$  from  $P(S' | T', \theta', \gamma)$

i) Simulate  $S'$  with transition matrix  $P(\theta'; \dots)$  and initial  $v_0$

ii) Accept  $S'$  with prob ~~prob~~ 
$$\frac{v_0(S'_0) \prod_{i=1}^{|T'|-2} P(\theta'; S'_i, S'_{i+1}) g_i(S'_i) \cdot g_0(S'_0) \prod_{i=1}^{|T'|-1} g_i(S'_{i+1})}{v_0(S'_0) \prod_{i=1}^{|T'|-2} P(\theta'; S'_i, S'_{i+1}) (\mu_{\max})^{|T'|} L^{\max}}$$

$$= \frac{\prod_{i=0}^{|T'|} g_i(S'_i)}{(\mu_{\max})^{|T'|} L^{\max}}$$

$$X = (\text{path}, \theta) \rightarrow X = (\text{path}', \theta')$$

(E<sub>1</sub>) in step 1, we obtain  $T' = J(X) \cup V^0$  and  $V^0 \in \gamma$

$$P(E_1) \geq \beta^0 \cdot \beta^{\text{nest}}$$

(E<sub>2</sub>) update  $\theta$  with  $\theta'$

$$P(E_2) \geq \text{Const} \cdot P_0(\theta') d\theta'$$

(E<sub>3</sub>) All points in  $J(X)$  ~~are~~ <sup>become</sup> virtual

$$P(E_3) \geq \beta_i^* \cdot \gamma(J(X)) = \beta_i^* \cdot \gamma^h \triangleq \beta_i$$

(E<sub>4</sub>) given E<sub>1</sub> - E<sub>3</sub>,  $S'$  is accepted

$$P(E_4) \geq \left( \frac{\gamma_{\min}}{\mu_{\max}} \right)^{h+n} \exp(-\mu_{\max}(t^{\max} - t^{\min})) \frac{L^+}{L^{\max}} = \beta_2$$

$$E = E_1 \cap E_2 \cap E_3 \cap E_4$$

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$$\begin{aligned} \therefore A(x, dx') &= \cancel{\Phi(dx')} \cdot \cancel{\text{Const}} \cdot P_0(\theta') d\theta'} \\ &= \underline{\Phi(dx')} \cdot \text{Const}. \end{aligned} \quad \left. \vphantom{\begin{aligned} \therefore A(x, dx') &= \cancel{\Phi(dx')} \cdot \cancel{\text{Const}} \cdot P_0(\theta') d\theta'} \right\} ?$$

if  $E$  happens, then a path' is generated and it's independent with  $X$ .  $\text{path} \sim \Phi \Big|_{\text{path}} (\cdot)$

$$\begin{aligned} \therefore A(x, dx') &= C \cdot P_0(\theta') d\theta' \cdot \Phi_{\text{path}}(\text{path}') \\ &= C \cdot \underline{\Phi(dx')} \quad \square \end{aligned}$$