

#### **D-ALPS:** Incorporating non-reversible temperature swapping to the Annealed Leap-Point Sampler

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# Warming Up

- Traditional MCMC algorithms with localized proposals tend to get trapped in a single modal region when facing a multimodal target  $\pi(x)$
- Parallel Tempering (PT) methods (Geyer, 1991) introduce an (inverse) temperature schedule  $\beta_1 = 1 > \beta_2 > \cdots > \beta_K > 0$  and an extended target of K replicas of x

$$\pi_{PT}(x_1, x_2, \dots, x_K) \propto \prod_{k=1}^K \pi(x_k)^{\beta_k}$$



• Each PT cycle contains two types of moves

- 1. Parallel within-level exploration (RWM, HMC, etc.)
- 2. Temperature Swap according to Metropolis ratio:

 $\frac{\pi_{PT}(x_1, \dots, x_{k+1}, x_k, \dots, x_K)}{\pi_{PT}(x_1, \dots, x_k, x_{k+1}, \dots, x_K)} = \exp\{(\beta_k - \beta_{k+1})[\log \pi(x_{k+1}) - \log \pi(x_k)]\}$ 

QuanTA Tawn and Roberts (2019) QuanTA transformation









ALPS targets are not direct powered-up versions of the original target but use mode-point information to approximately preserve regional weights (full details in paper):

 $\pi_{\beta}(x) \propto \begin{cases} \pi(x)^{\beta} \pi(\mu(x,\beta))^{1-\beta}, & \mu(x,\beta) = \mu(x,1) \\ G(x,\beta), & \text{a.o.c.} \end{cases}$ 

---> Switches Modes --> Mode 2 State type ->> Mode 1

# Now to the Cold

- Tawn et al (2021) introduced **ALPS**:
  - 1. Mode finder exploration component at a hot temperature  $\beta_H \ll 1$
  - 2. Parallel Annealing component with colder schedule  $\beta_1 = 1 < \beta_2 < \cdots < \beta_K$
- At colder temperatures ALPS targets<sup>1</sup> look increasingly Gaussian around the mode points  $\{\mu_1, \mu_2, \dots, \mu_m\}$ , so use **Independent**

## Non Rev. Swapping

- Consider 3 types of swapping:
  - 1. Naive- randomly select a single pair
  - 2. SEO- randomly select block of Even or Odd pairs
  - 3. **DEO-** deterministically alternate between Even and Odd pairs.

![](_page_0_Figure_31.jpeg)

- Syed et al (2022) proved that in PT, nonreversible DEO swapping of Okabe et al (2001) dominates its reversible SEO version
- Both SEO and DEO take advantage of parallelization, but DEO carries momentum

 $G(x,\beta) = \left(\frac{\tau}{\beta}\right)^{d/2} \pi(\mu(x,\beta)) |\Sigma(x,\beta)|^{1/2} \phi\left(x \mid \mu(x,\beta), \frac{\Sigma(x,\beta)}{\beta}\right)$ 

![](_page_0_Picture_37.jpeg)

#### Results

![](_page_0_Figure_39.jpeg)

### References

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