

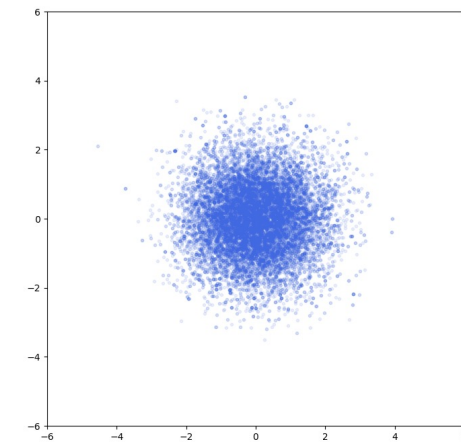
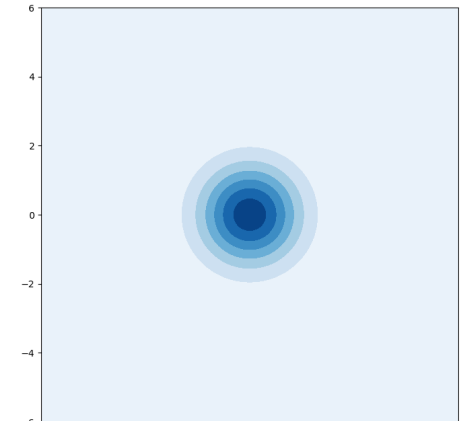
Fixed-distance Hamiltonian Monte Carlo

Hadi Mohasel Afshar, Sally Cripps

CSIRO's Data61 & The University of Sydney

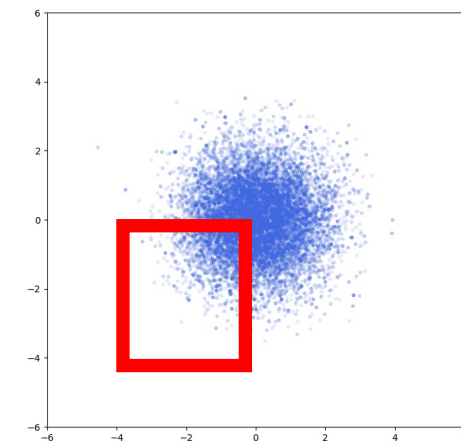
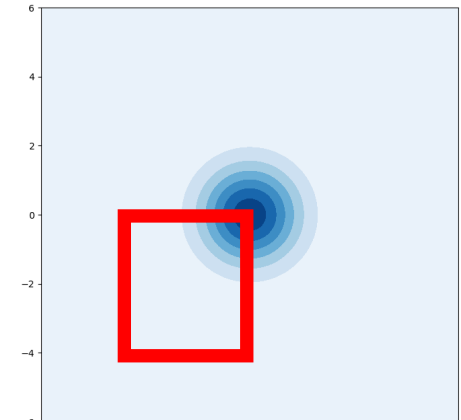
Markov Chain Monte Carlo (MCMC)

- MCMC are a class methods that approximate a probability distribution with a set of states a.k.a. samples.



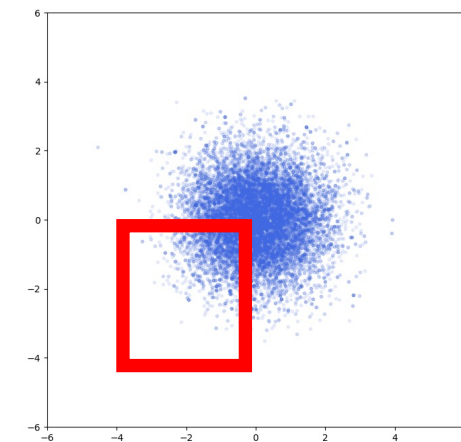
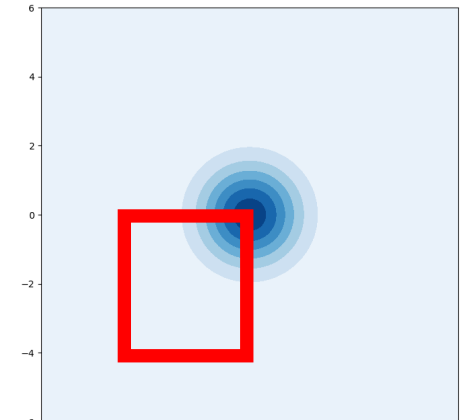
Markov Chain Monte Carlo (MCMC)

- MCMC are a class methods that approximate a probability distribution with a set of states a.k.a. samples.
- Sampling is unbiased if #draws from each region is proportional to the probability mass of that region.



Markov Chain Monte Carlo (MCMC)

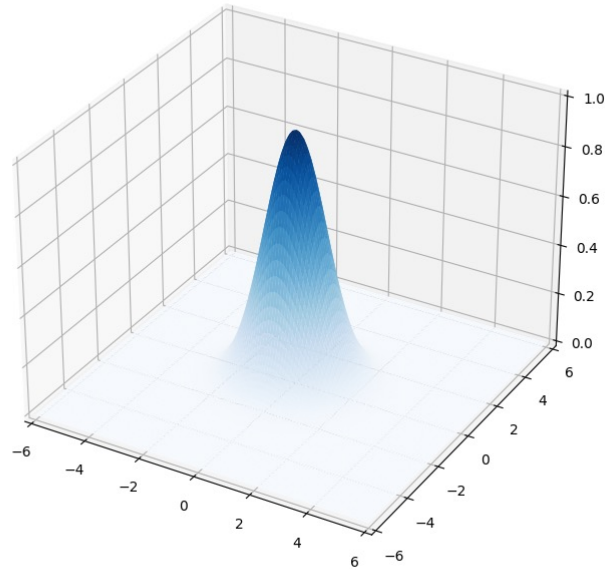
- MCMC are a class methods that approximate a probability distribution with a set of states a.k.a. samples.
- Sampling is unbiased if #draws from each region is proportional to the probability mass of that region.
- The state of the art MCMC methods (e.g. NUTS) are variations of the Hamiltonian Monte Carlo (HMC)



Hamiltonian Monte Carlo (HMC)

- In HMC, the potential energy is defined as the $-\log$ of PDF.

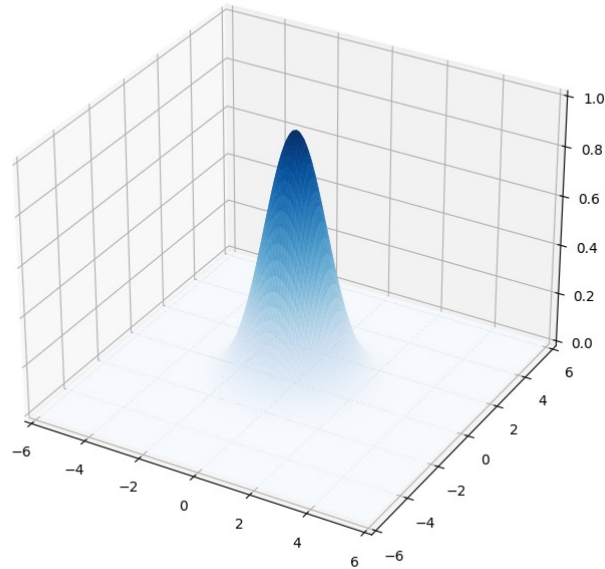
Target probability density: $\pi(\mathbf{q})$



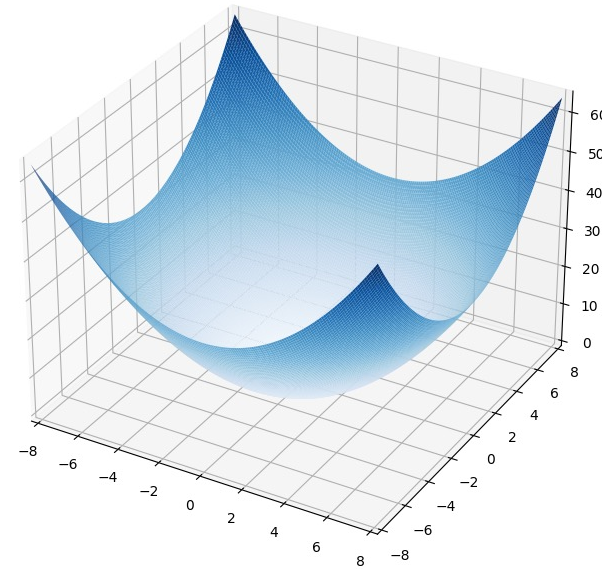
Hamiltonian Monte Carlo (HMC)

- In HMC, the potential energy is defined as the $-\log$ of the PDF.

Target probability density: $\pi(\mathbf{q})$

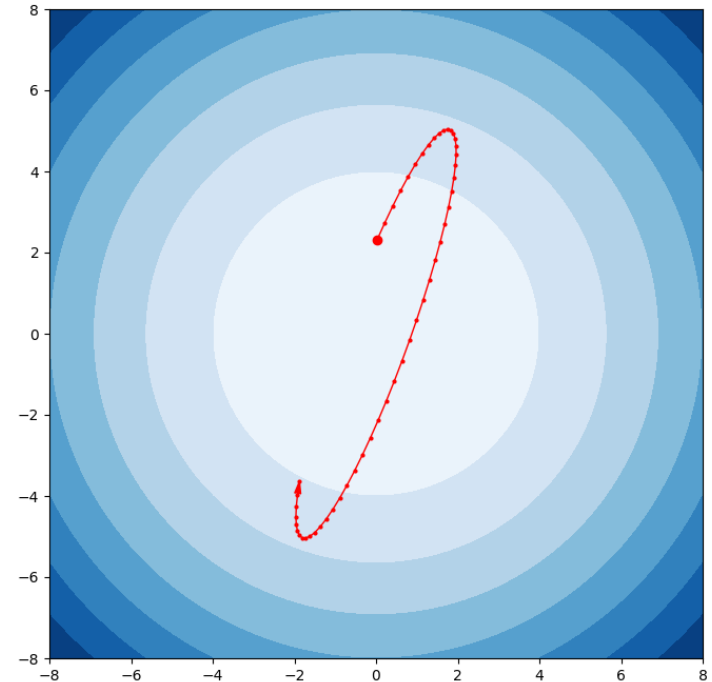


Potential energy: $U(\mathbf{q}) = -\log \pi(\mathbf{q})$



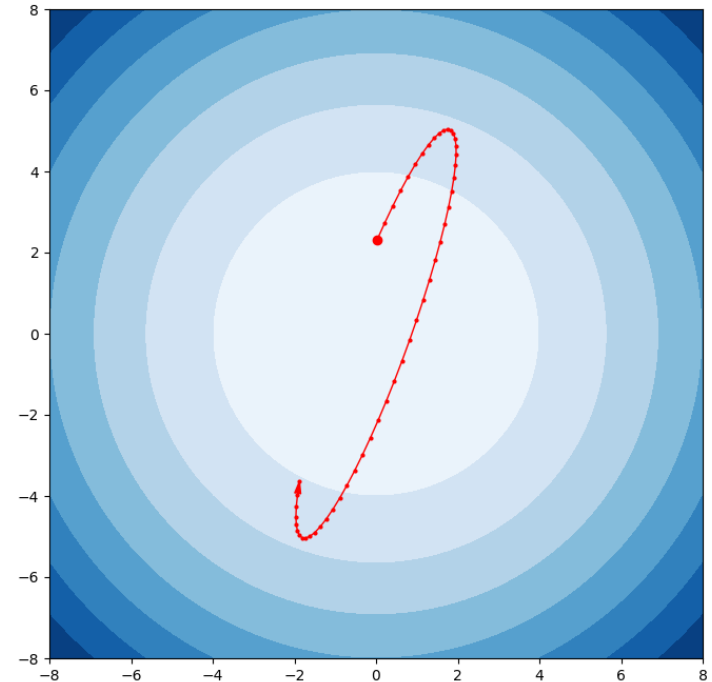
Hamiltonian Monte Carlo (HMC)

- In HMC, the potential energy is defined as the neg. log of the PDF.
- An auxiliary momentum $\mathbf{p} \sim N(0, I)$ is assigned to the current state. Then, it is evolved via the simulation of the equations of motion for a fixed time.

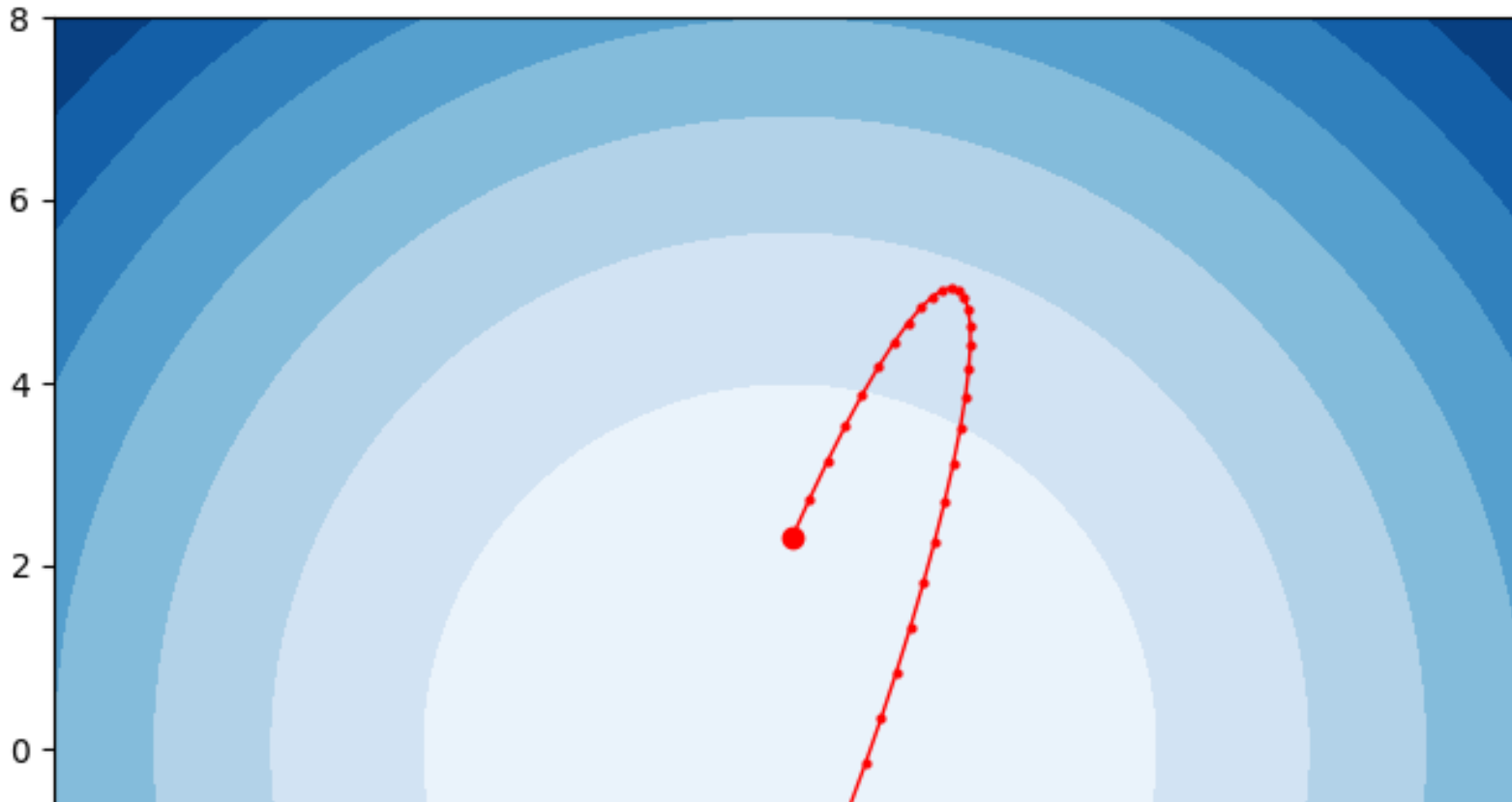


Hamiltonian Monte Carlo (HMC)

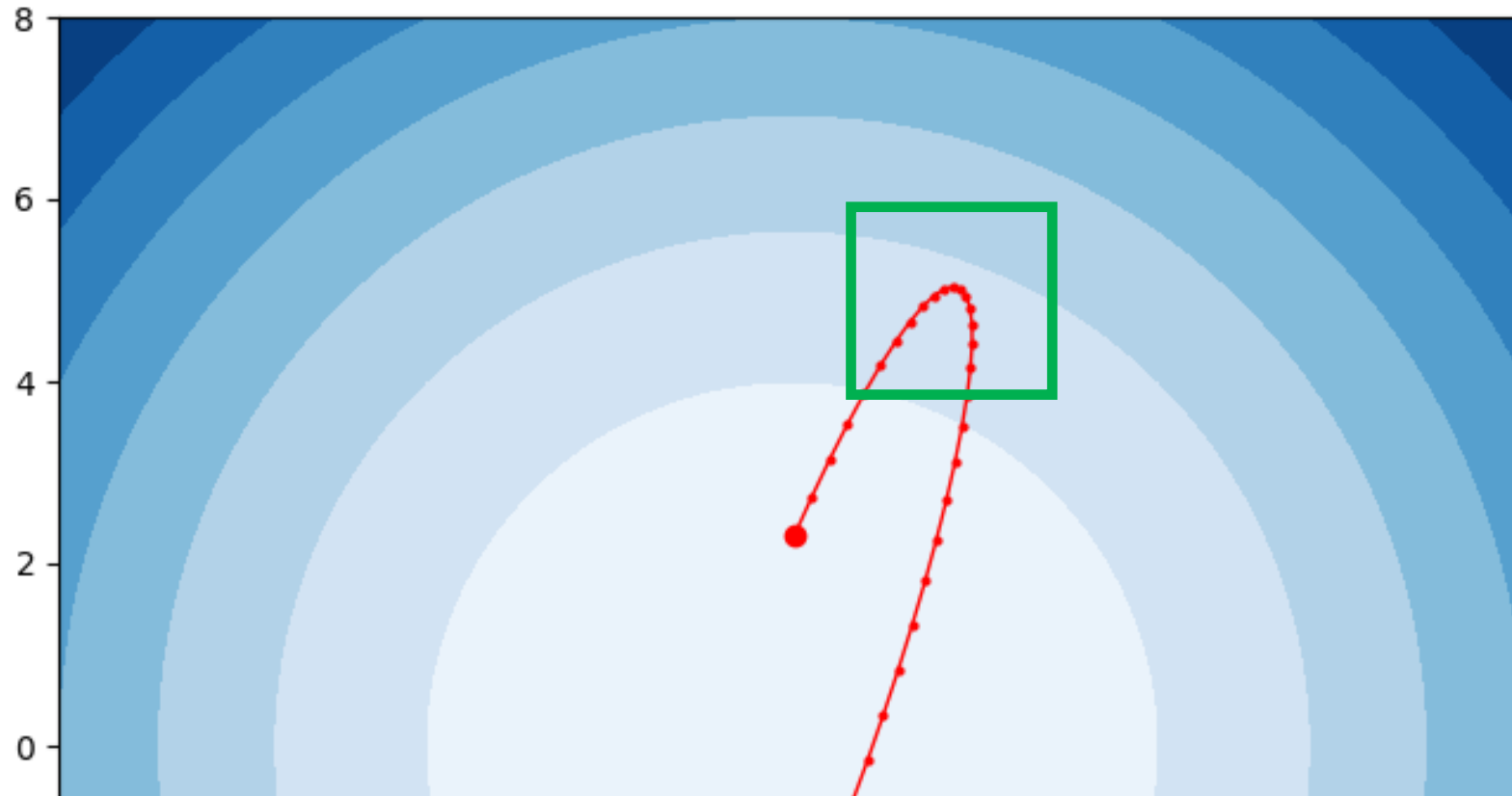
- In HMC, the potential energy is defined as the neg. log of PDF.
- An auxiliary momentum $\mathbf{p} \sim N(0, I)$ is assigned to the current state. Then, it is evolved via the simulation of the equations of motion for a fixed time.
- If the simulation is precise, the acceptance probability of the resulting proposal is high.



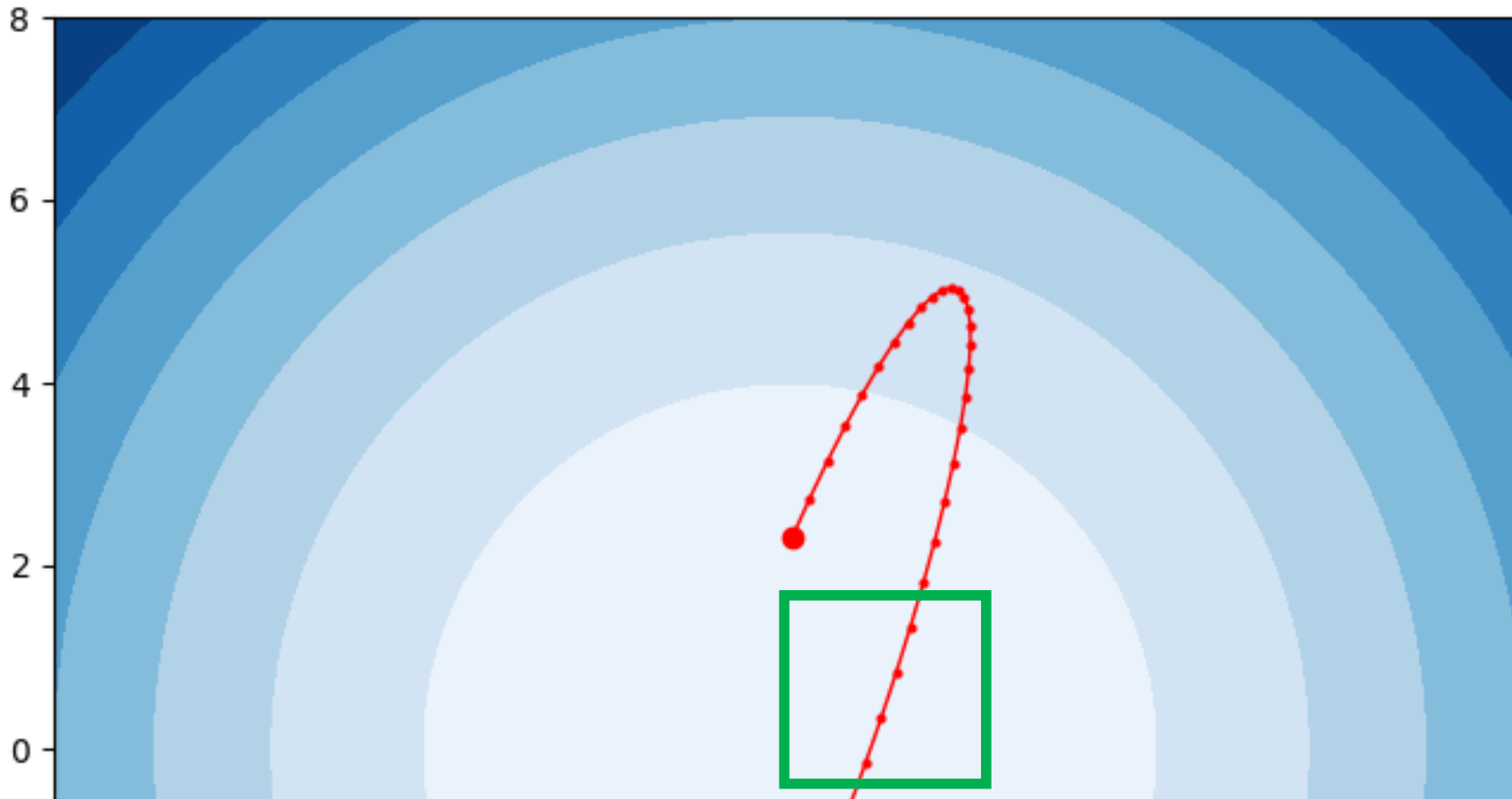
Note: In high potential (i.e. low probability) regions the velocity is low



More states are generated from these regions...

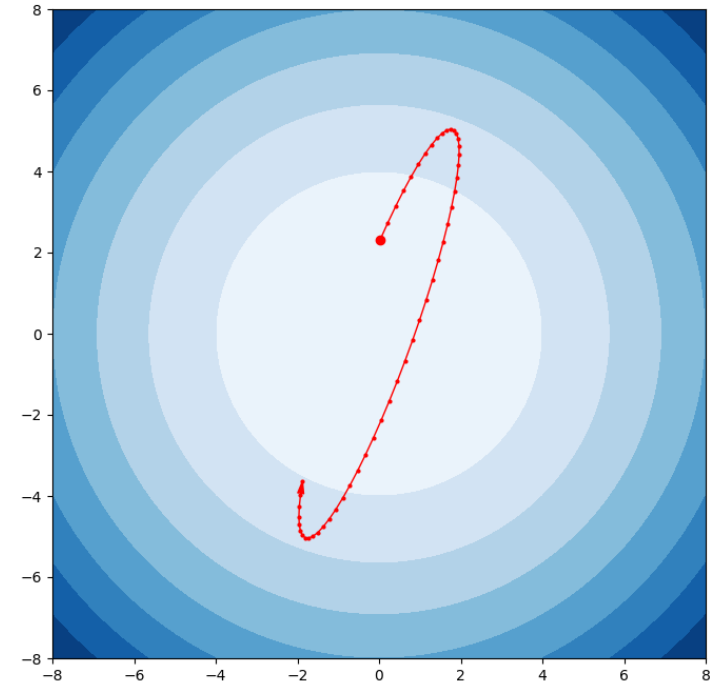


Conversely, less time is spent in the high-probability regions



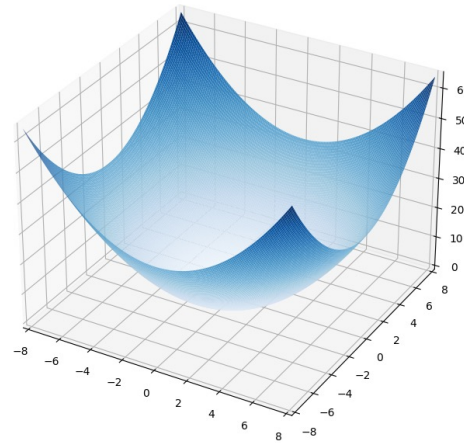
Evolution bias

- As such, there is bias towards proposing low-probability states which is counter-intuitive!



To verify ...

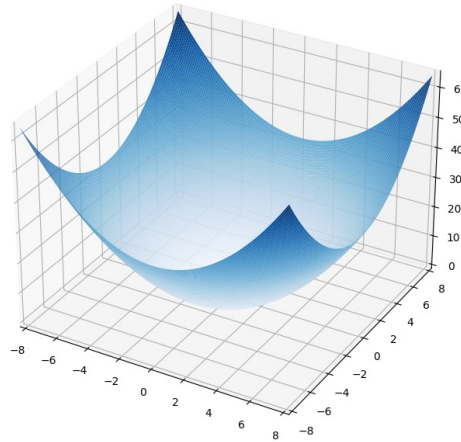
Consider a potential energy function:



To verify ...

Consider a potential energy function:

We let the current state be on the unit circle

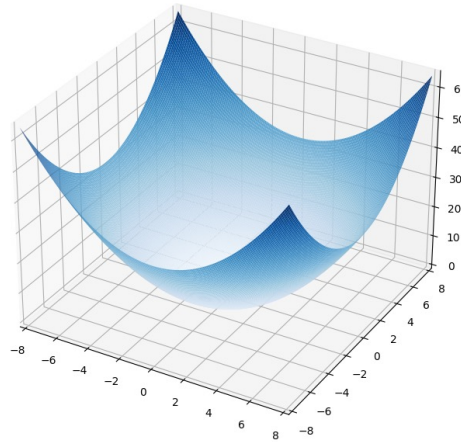


To verify ...

Consider a potential energy function:

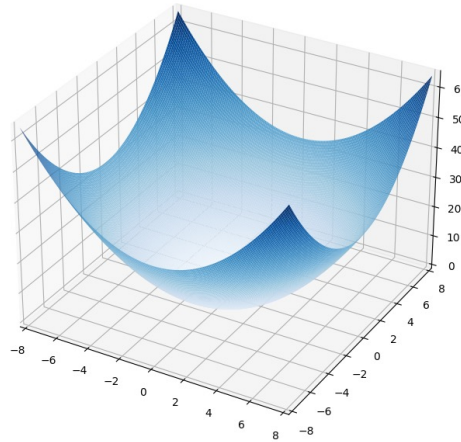
We let the current state be on the unit circle

We fix the momentum magnitude but randomly choose its direction



To verify ...

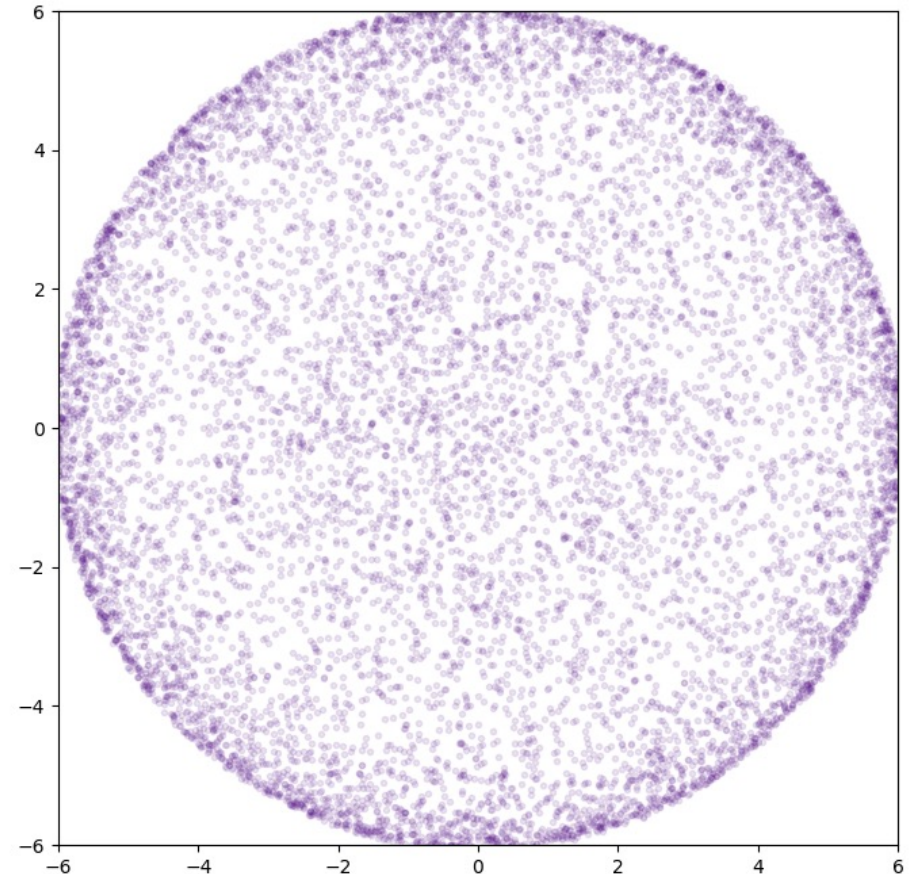
Consider a potential energy function:



We let the current state be on the unit circle

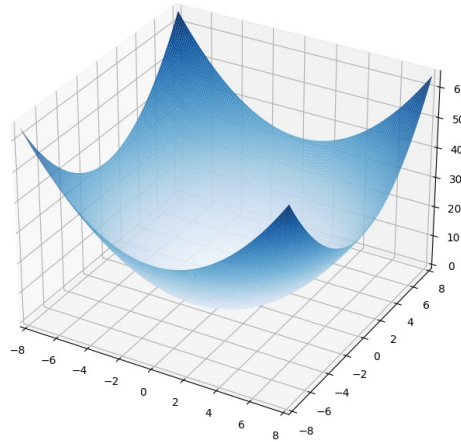
We fix the momentum magnitude but randomly choose its direction

... and plot the proposals:



To verify ...

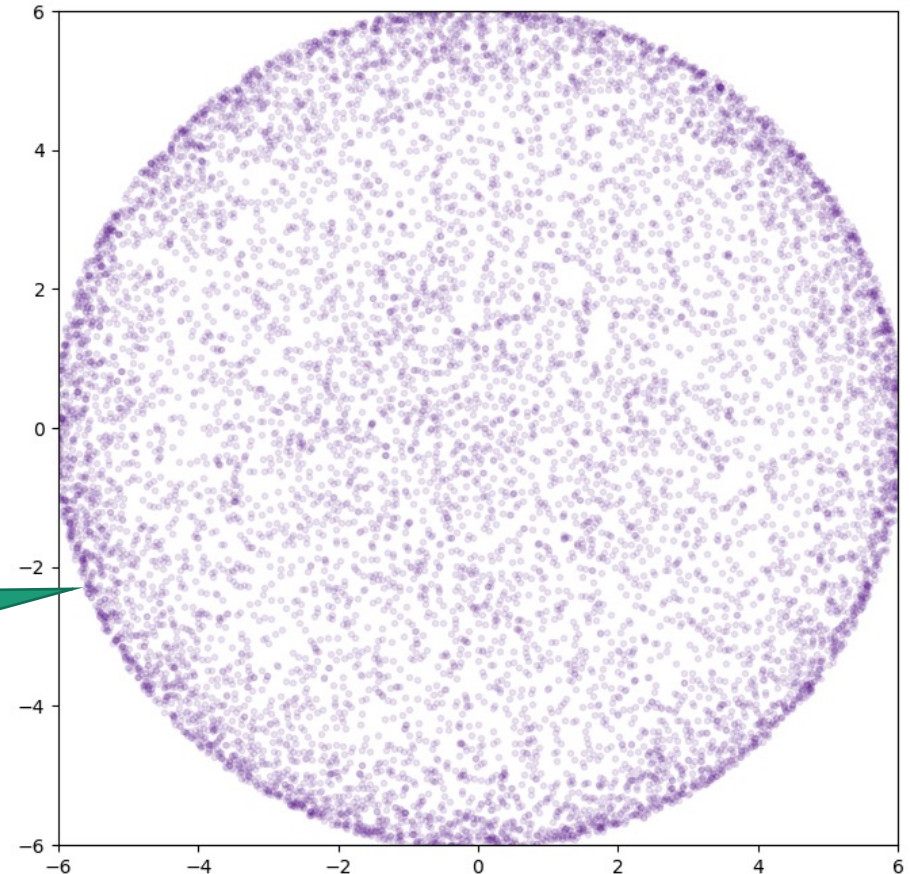
Consider a potential energy function:



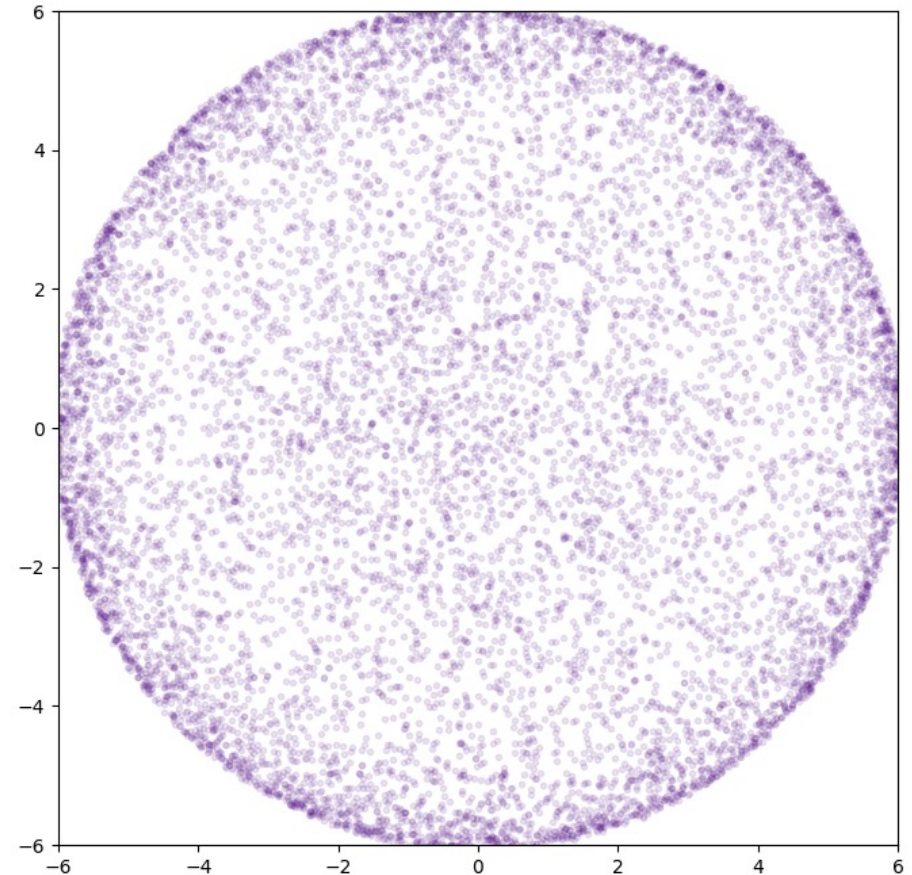
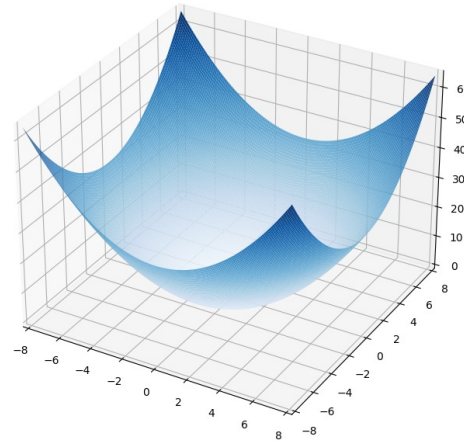
We let the current state be on the unit circle

We fix the m
but random
... and plot th

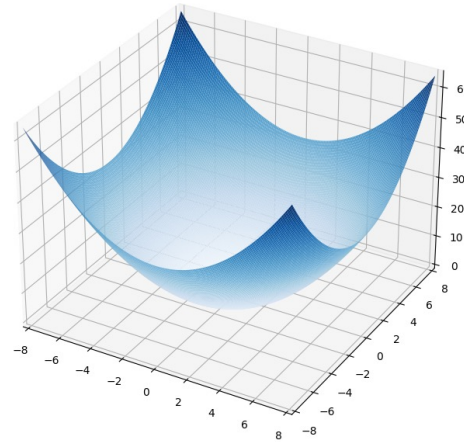
**Regions of low probability
are over-represented**



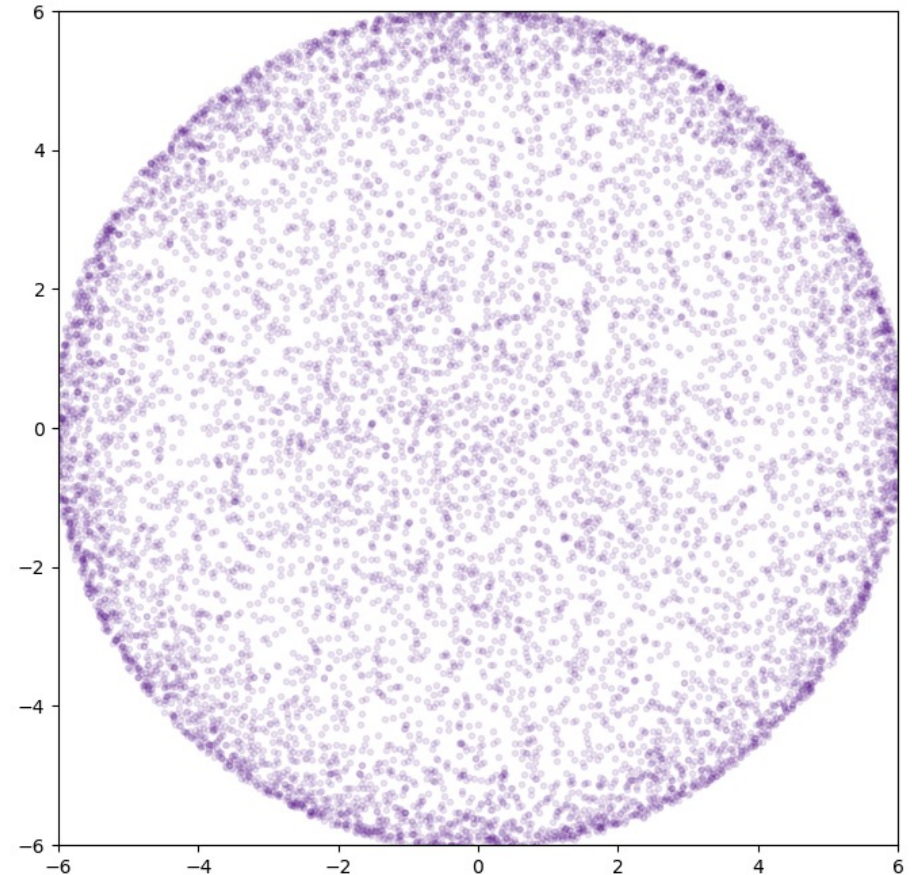
Then, what makes the total HMC sampling unbiased?



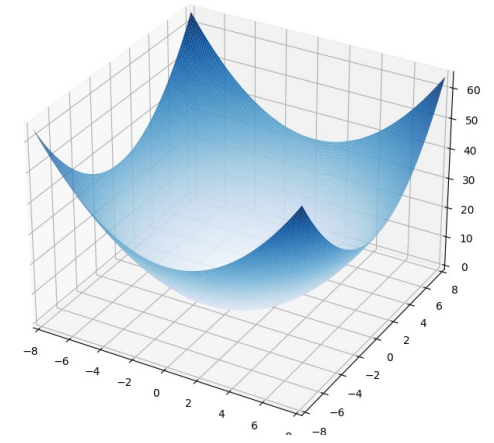
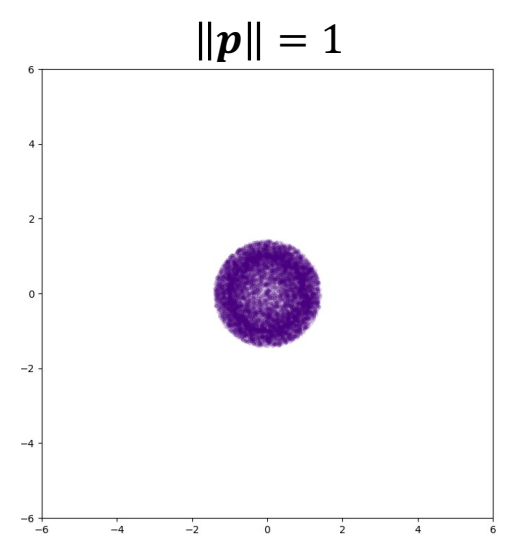
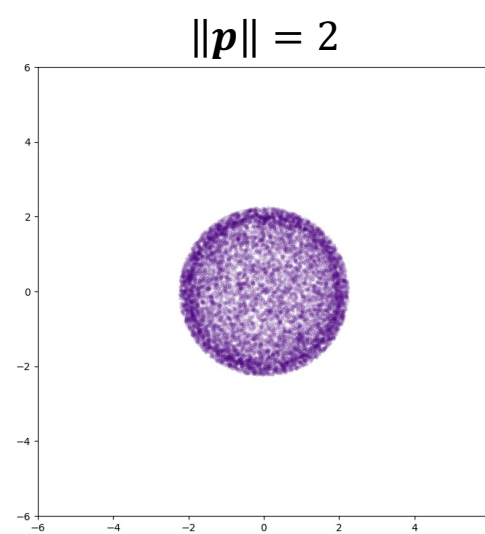
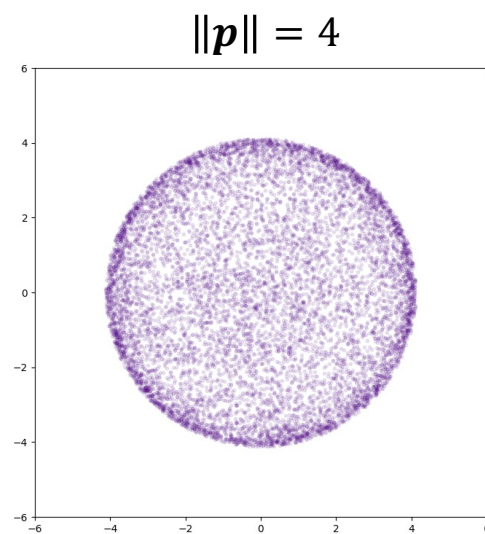
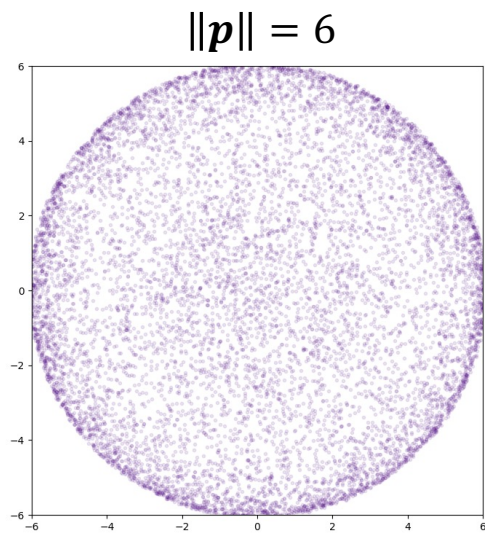
Then, what makes the total HMC sampling unbiased?



The momentum distribution...

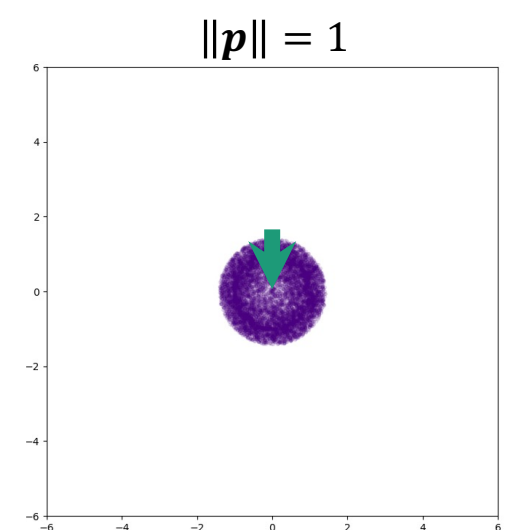
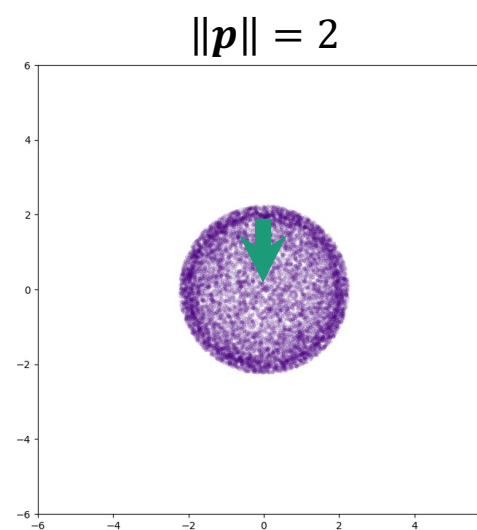
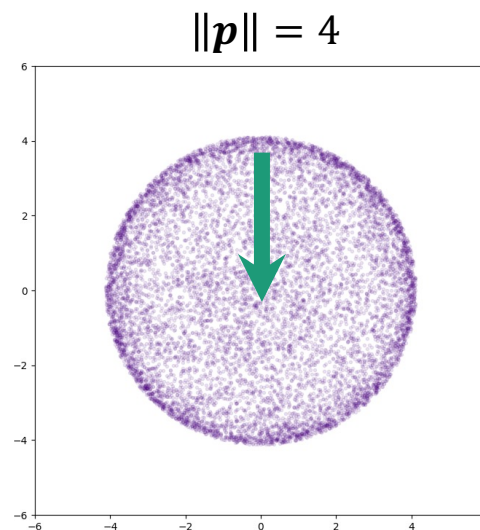
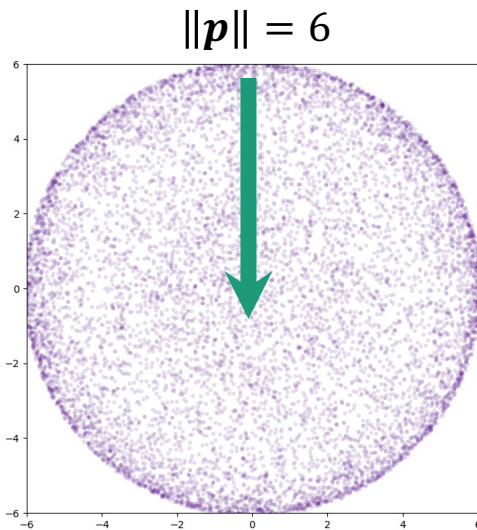
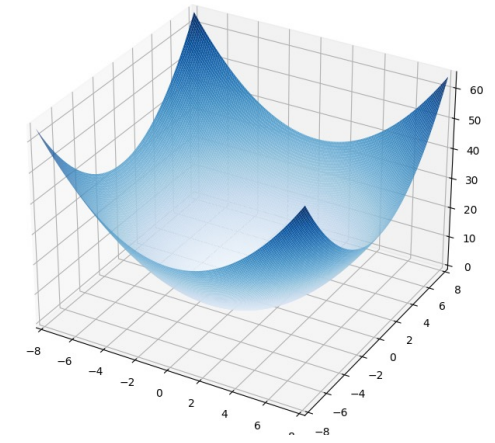


Previous example with different momentum magnitudes...



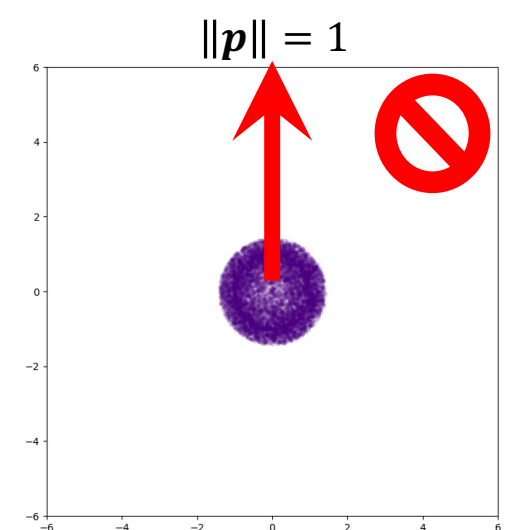
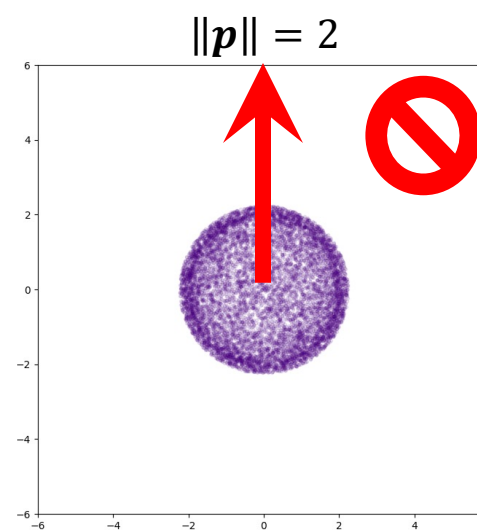
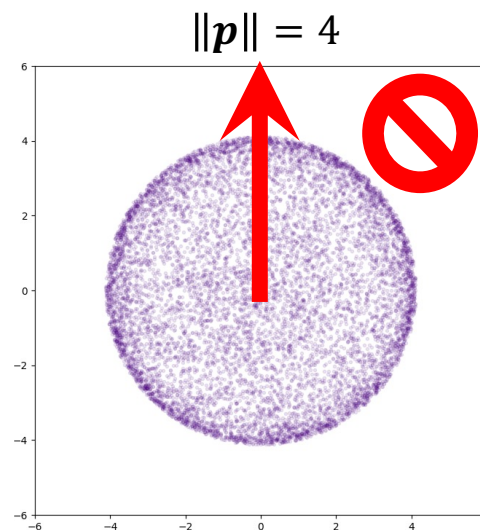
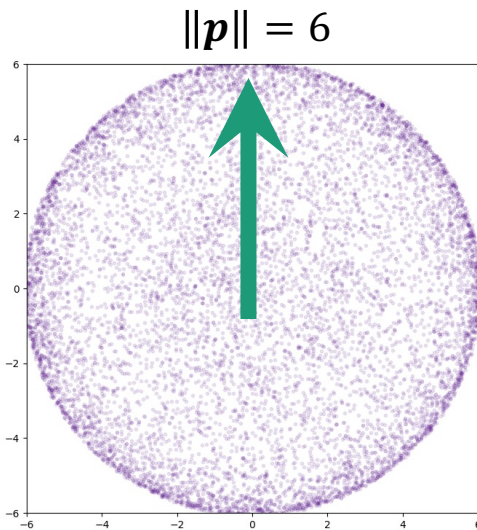
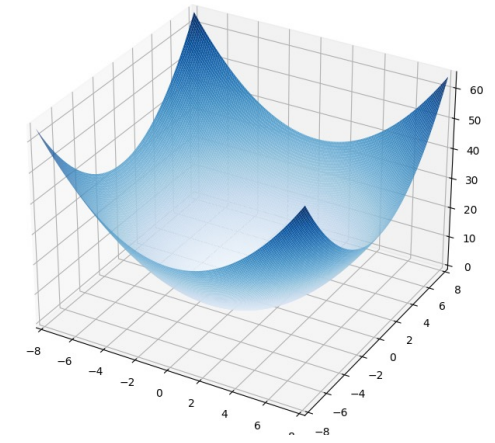
Previous example with different momentum magnitudes...

The transition from a low to high probability region is always possible



Previous example with different momentum magnitudes...

But if the momentum's magnitude is low, the transition from a high to low probability region may NOT be possible.



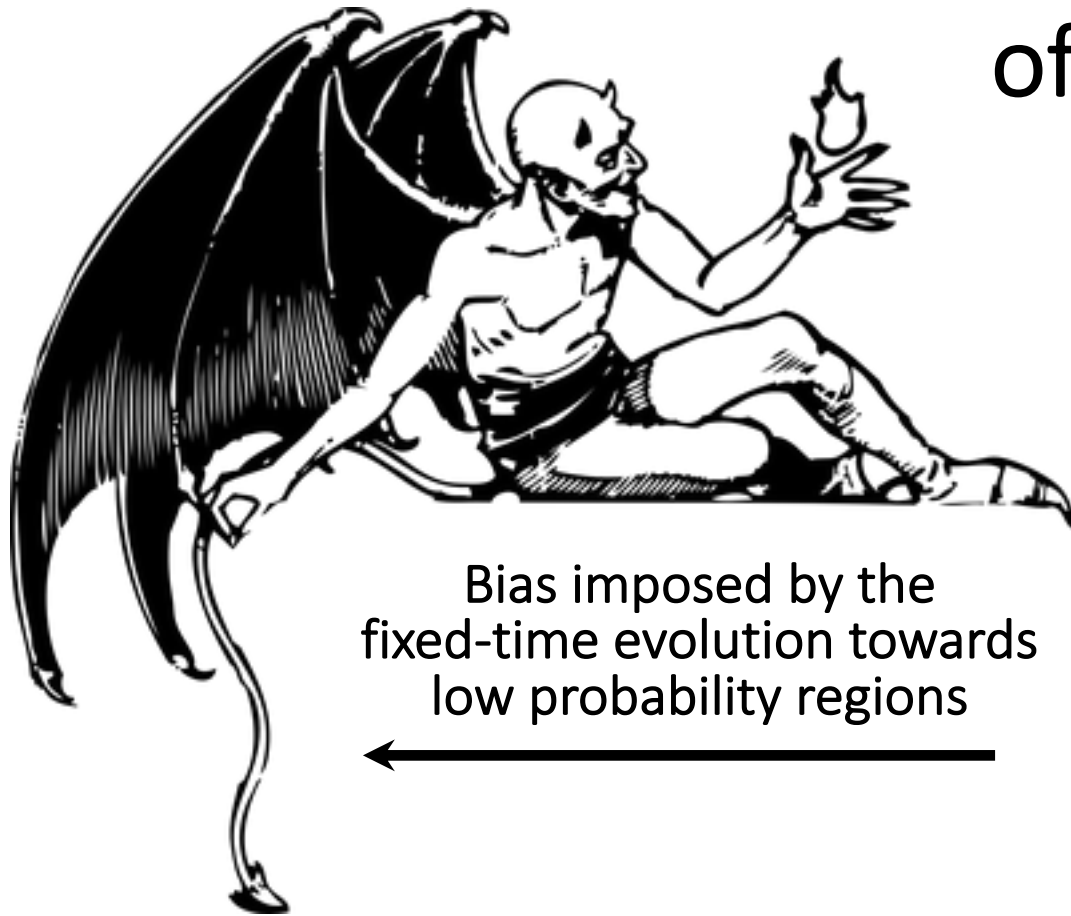
Two counter-balancing biases...

Convergence of HMC



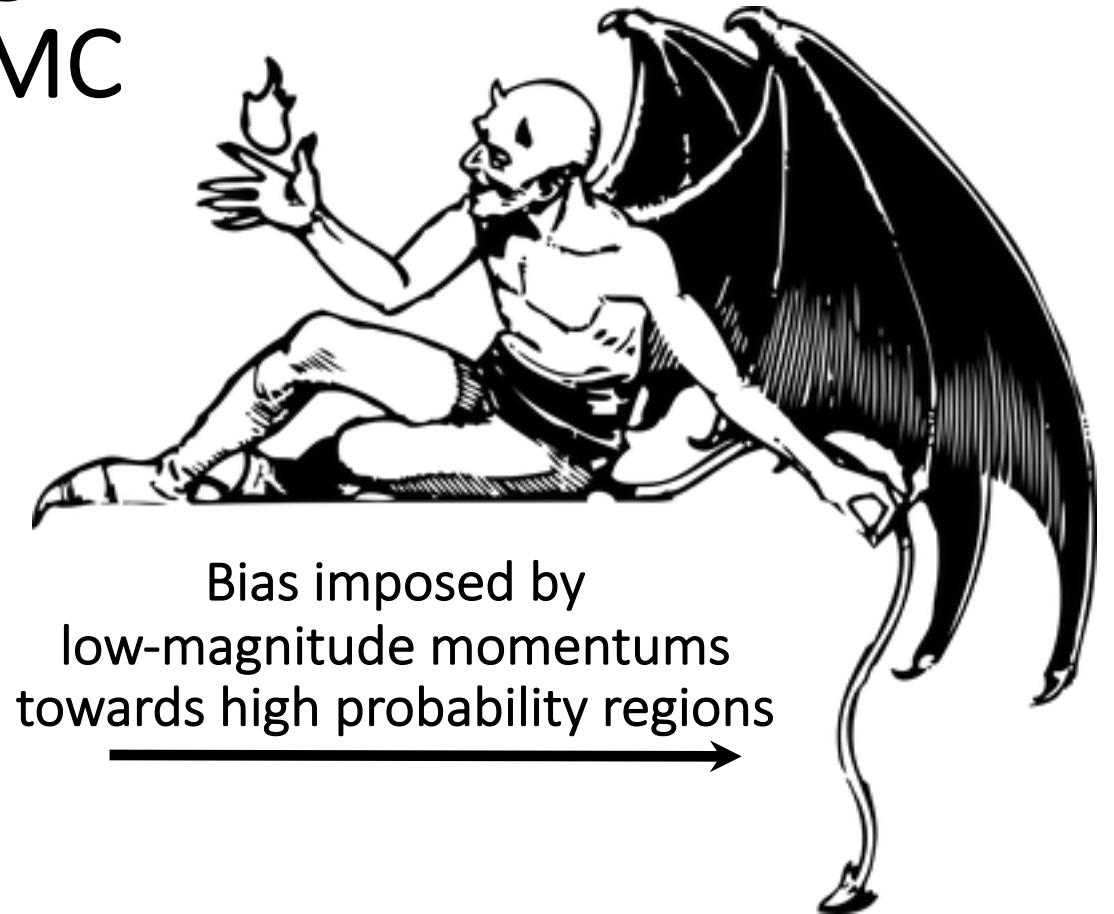
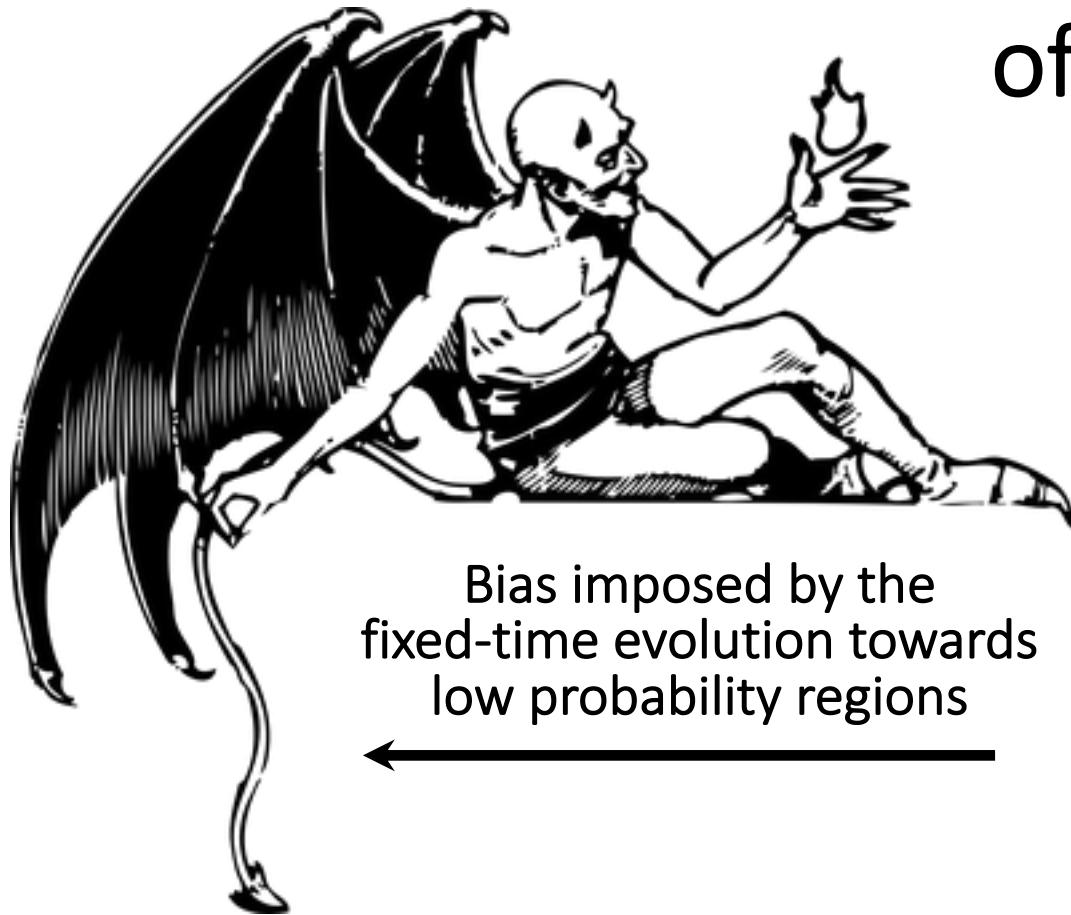
Two counter-balancing biases...

Convergence of HMC



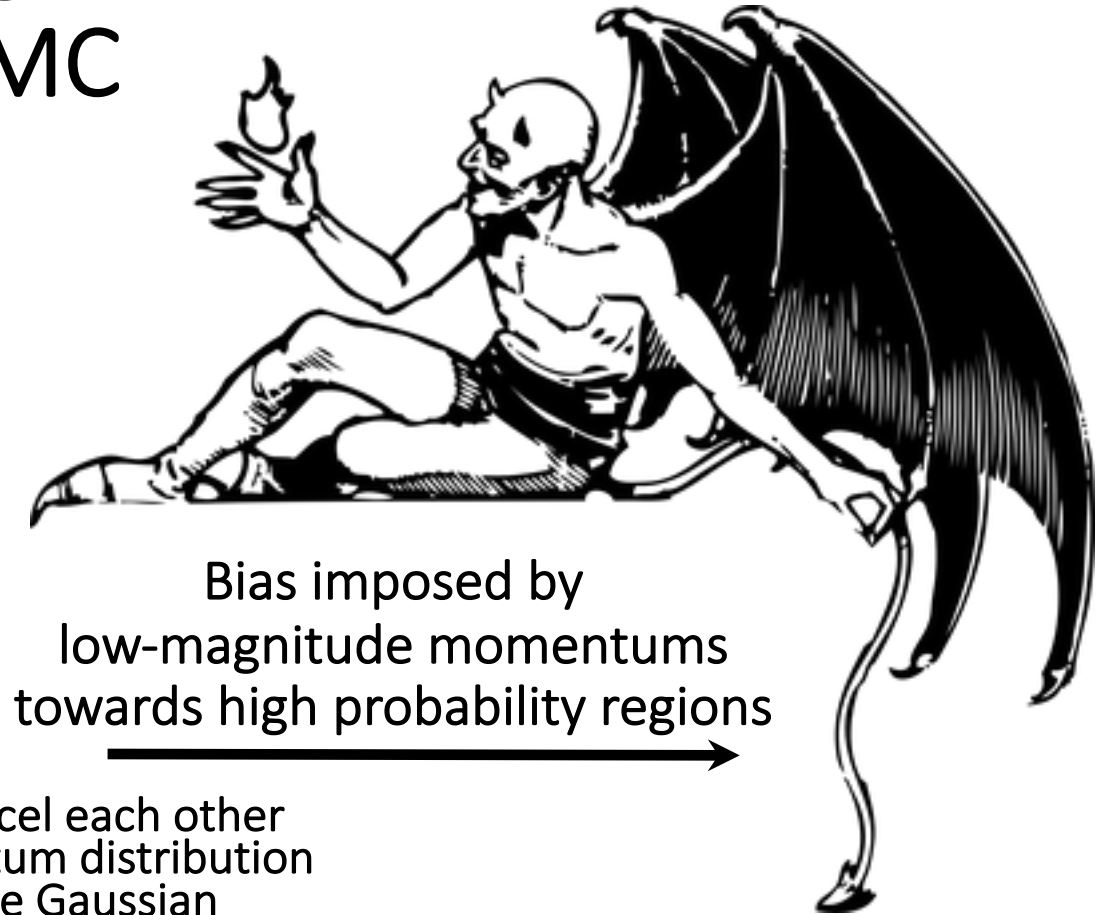
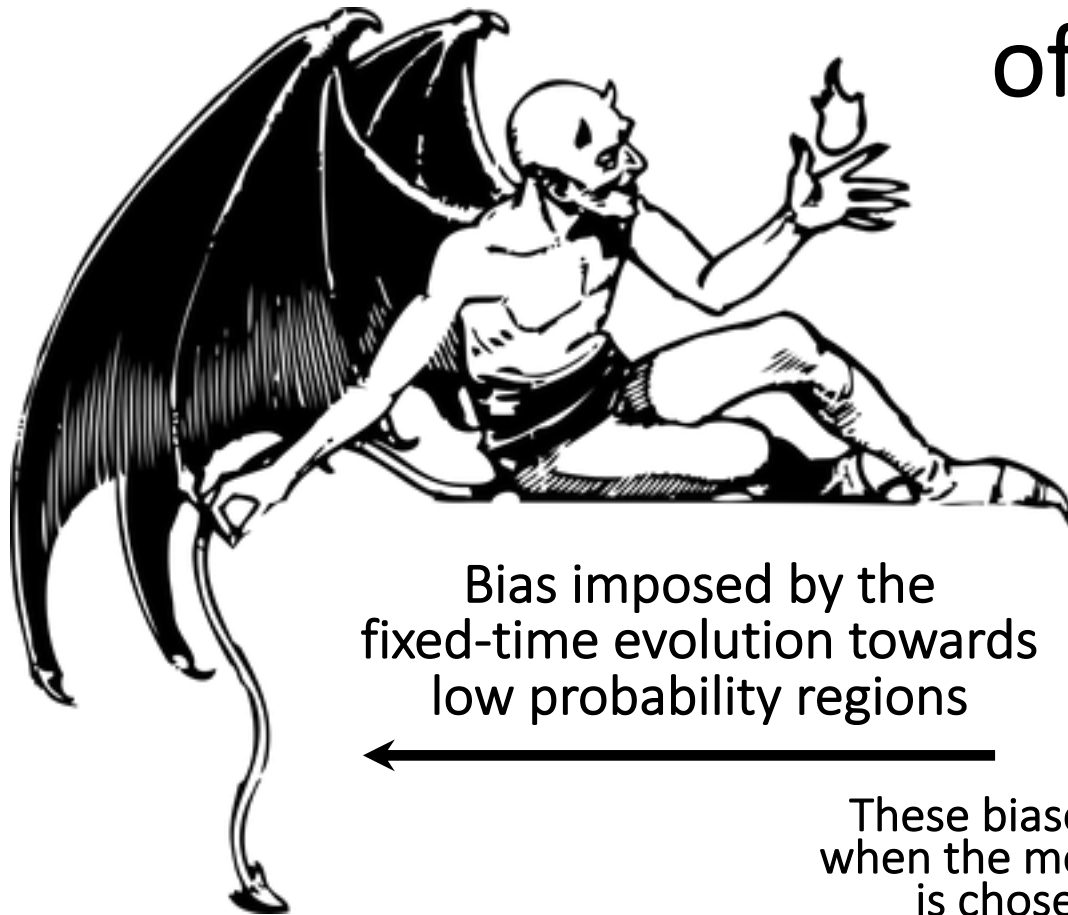
Two counter-balancing biases...

Convergence of HMC



Two counter-balancing biases...

Convergence of HMC

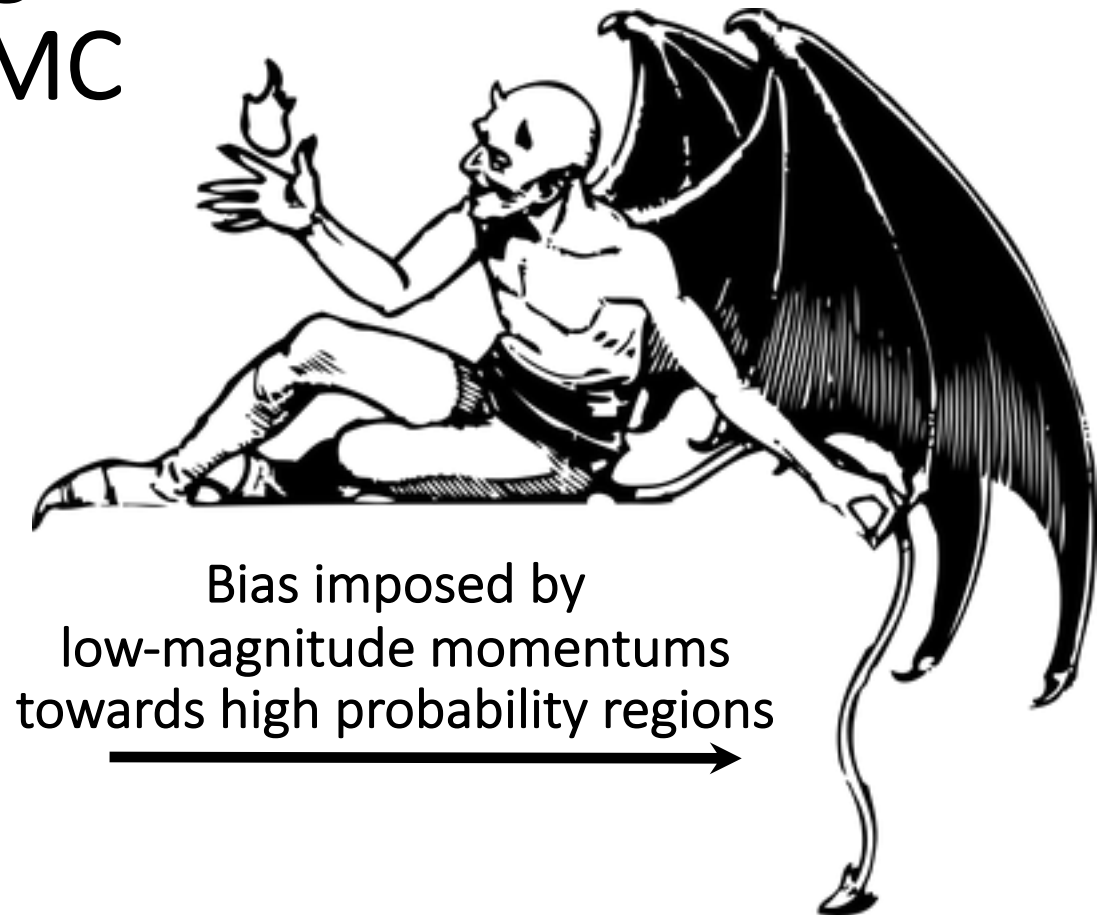


These biases cancel each other
when the momentum distribution
is chosen to be Gaussian

Resolving the first bias

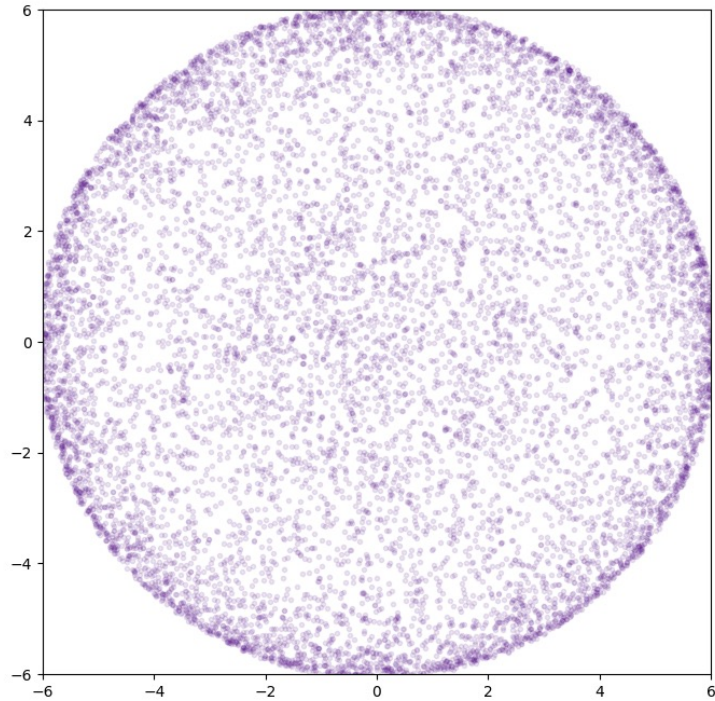
Convergence of HMC

Fixed distance
(rather than fixed-time)
evolution

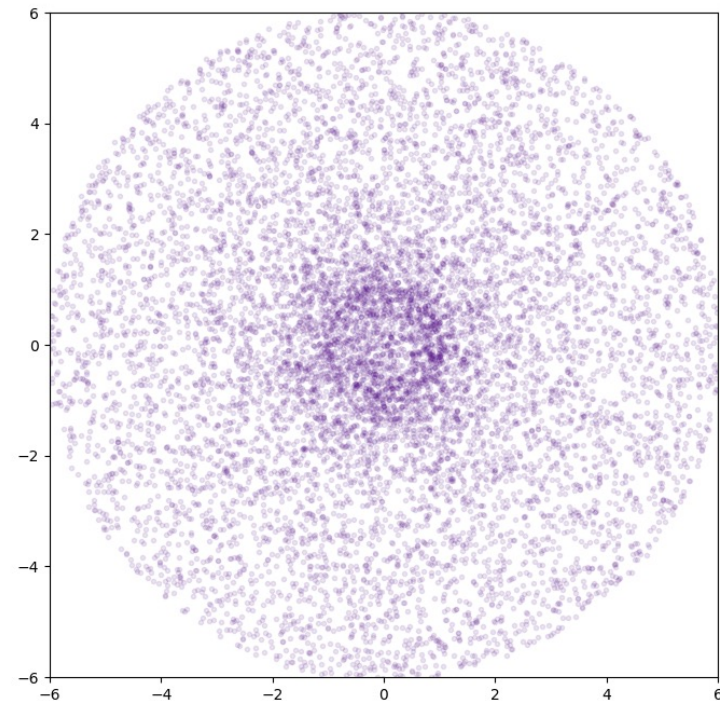


Bias imposed by
low-magnitude momentums
towards high probability regions

Fixed-distance simulation of the equations of motion

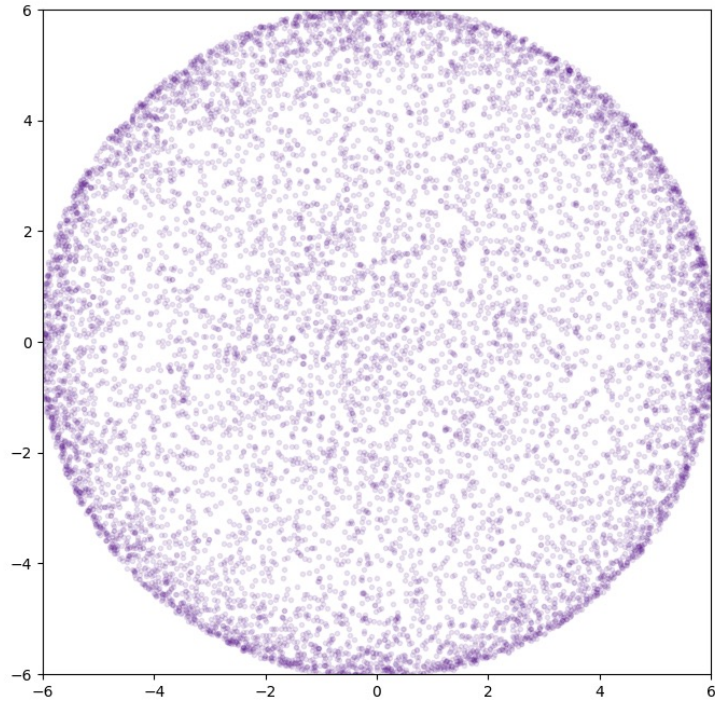


Fixed-time evolution

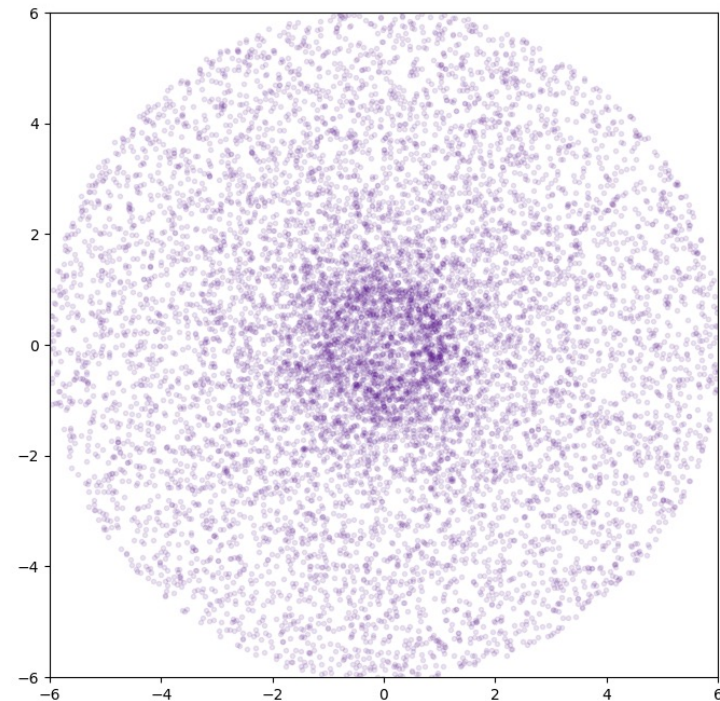


Fixed-distance evolution

Fixed-distance simulation of the equations of motion



Fixed-time evolution

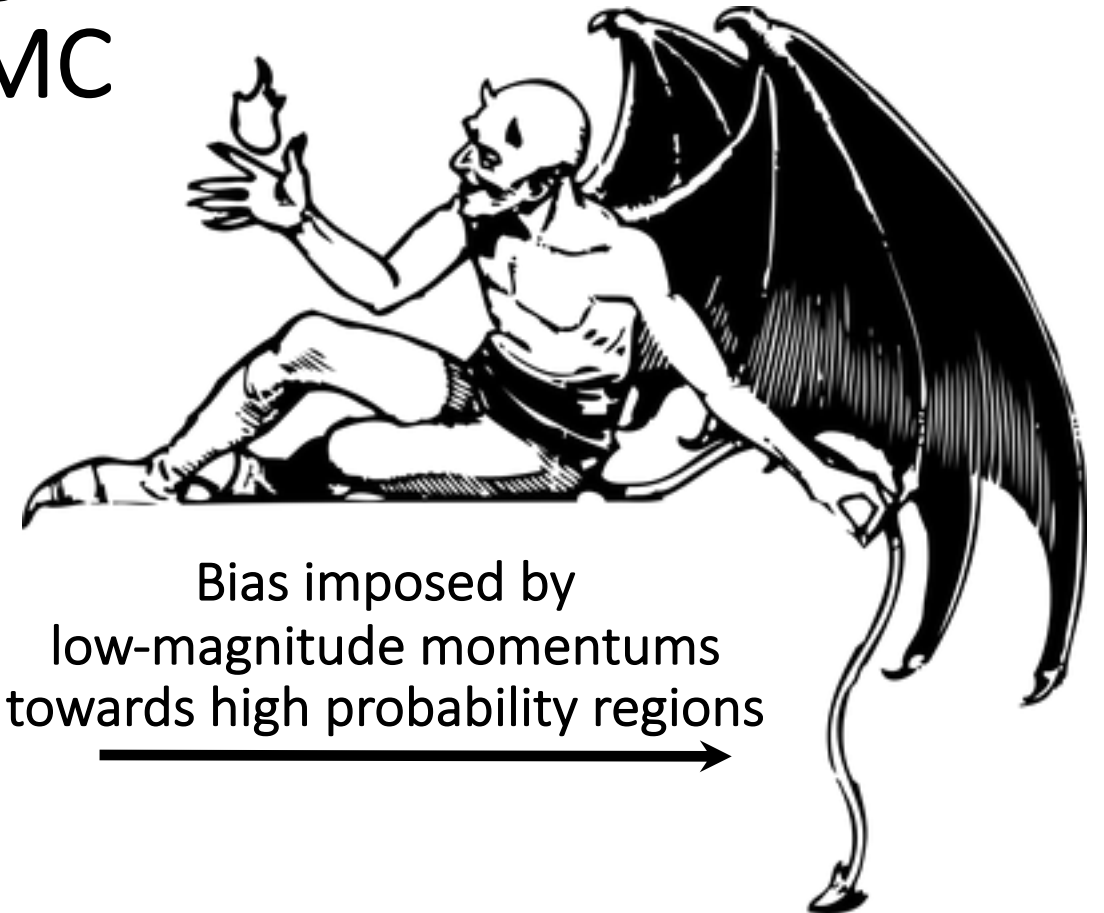


Fixed-distance evolution

Even though more time is spent in low-probability regions, more distance is not traversed there

To resolve the second bias...

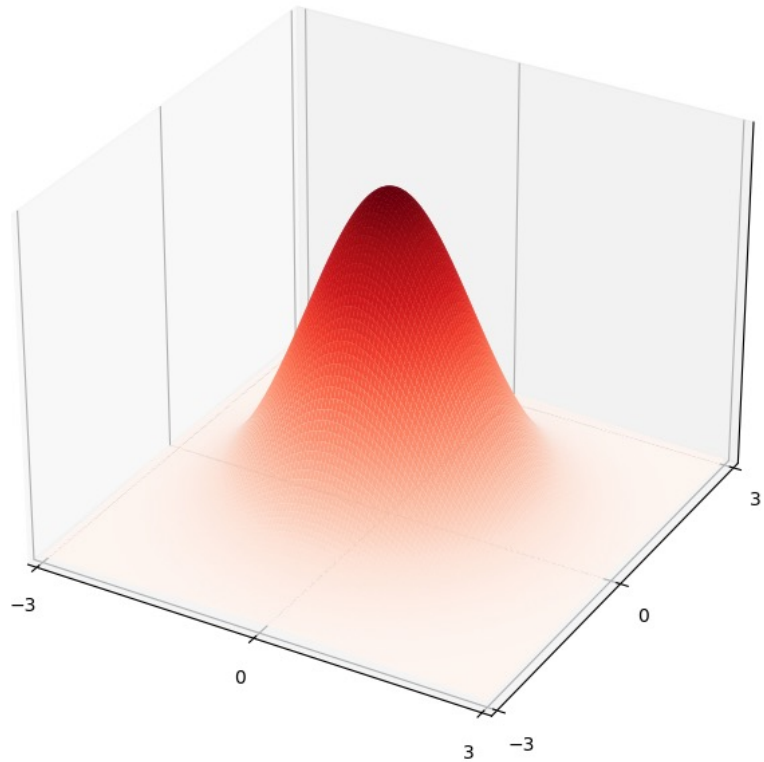
Convergence of HMC



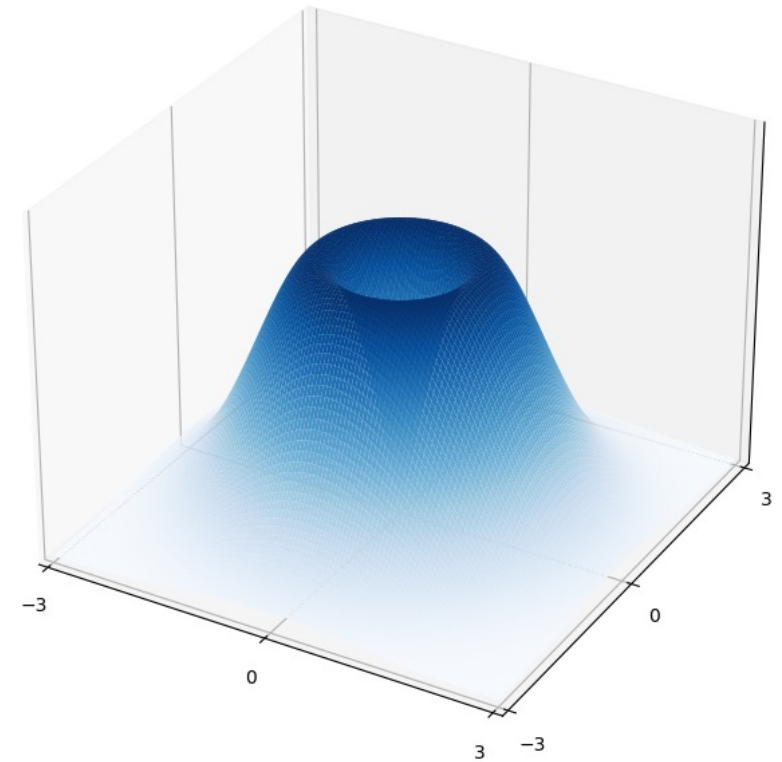
Bias imposed by
low-magnitude momentums
towards high probability regions



We prove...



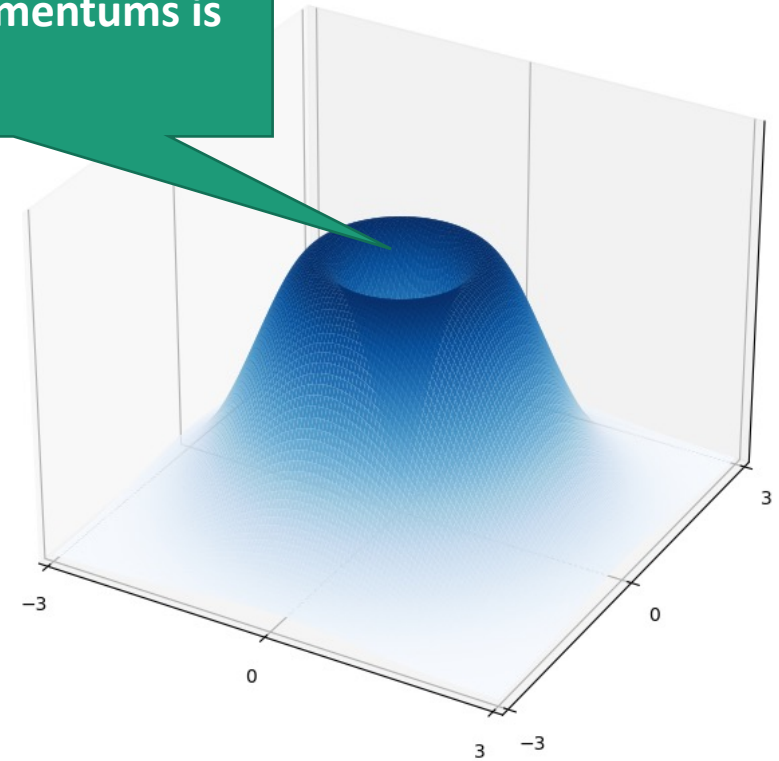
HMC's momentum distribution



FDHMC's momentum distribution

FDHMC's momentum distribution

Probability assigned to low-magnitude momentums is low

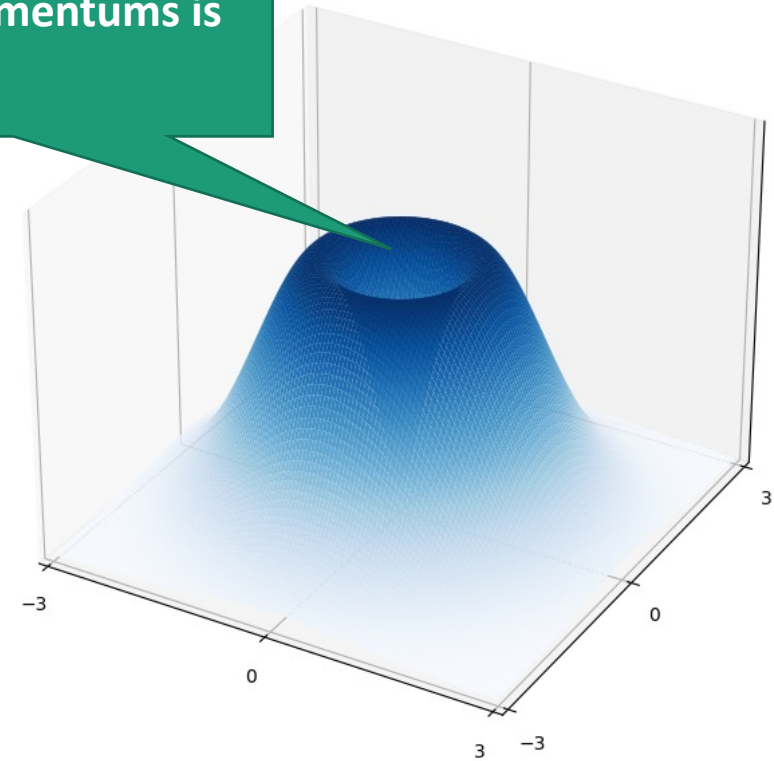


FDHMC's momentum distribution

FDHMC's momentum distribution

Probability assigned to low-magnitude momentums is low

The expected magnitude of FDHMC's momentum vector is higher



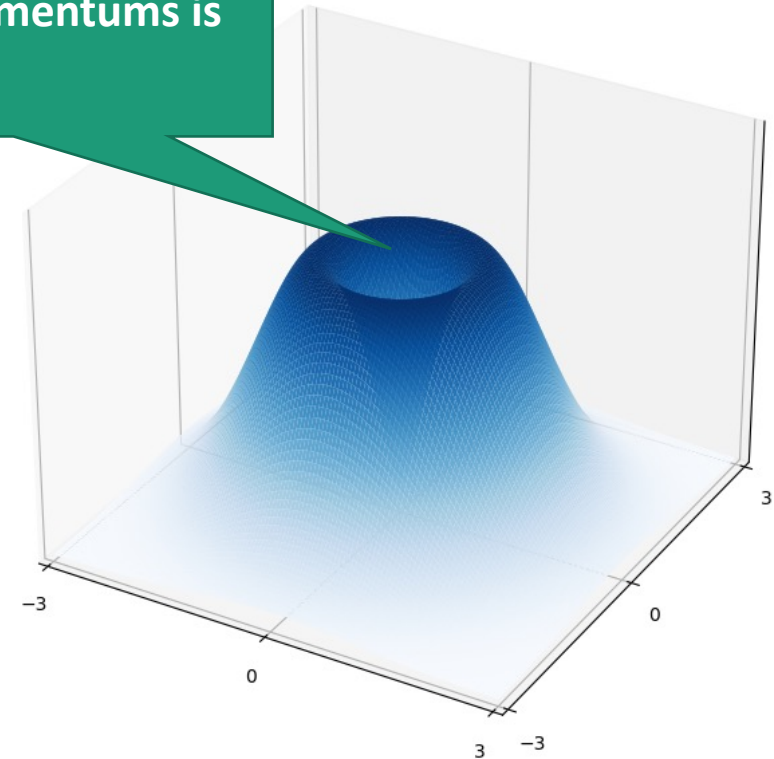
FDHMC's momentum distribution

FDHMC's momentum distribution

Probability assigned to low-magnitude momentums is low

The expected magnitude of FDHMC's momentum vector is higher.

This translates to higher ESS and better exploration of target probability modes.



FDHMC's momentum distribution

Conclusion

- With negligible computational overhead, we resolved two counter-balancing biases that exist in the core of HMC algorithms.

Conclusion

- With negligible computational overhead, we resolved two counter-balancing biases that exist in the core of HMC algorithms.
- Our experiments show that the resulting FDHMC has a higher ESS/grad.