

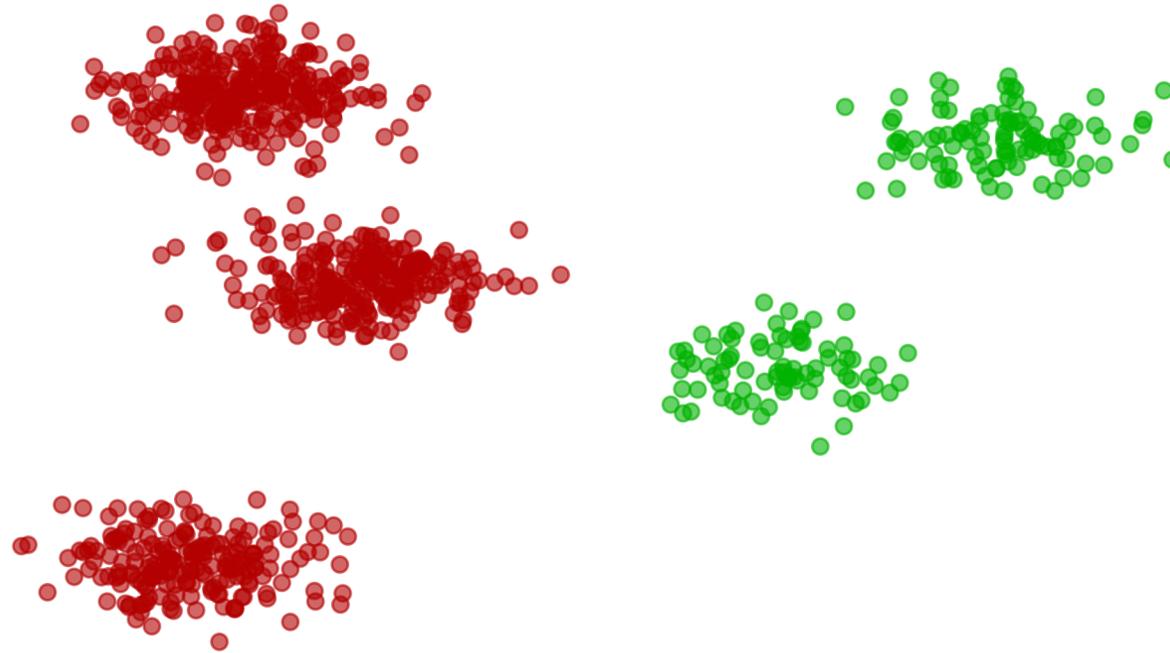


# Online Variance Reduction with Mixtures

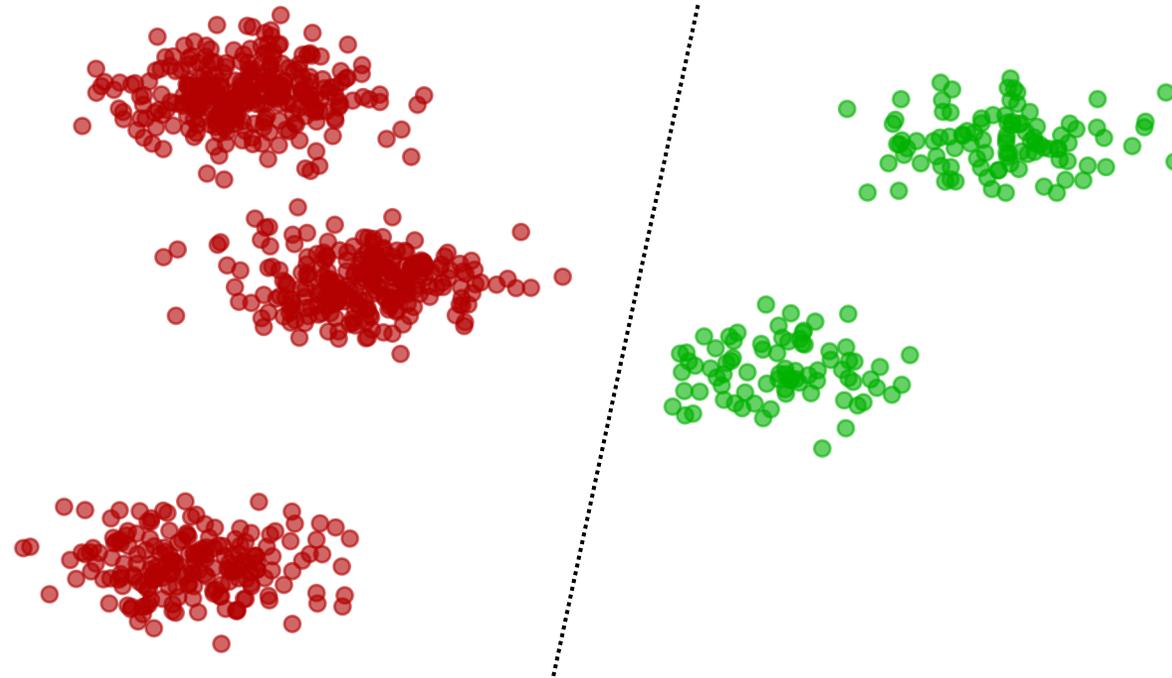
Zalán Borsos, Sebastian Curi, Kfir Y. Levy and Andreas Krause

*Department of Computer Science, ETH Zurich*

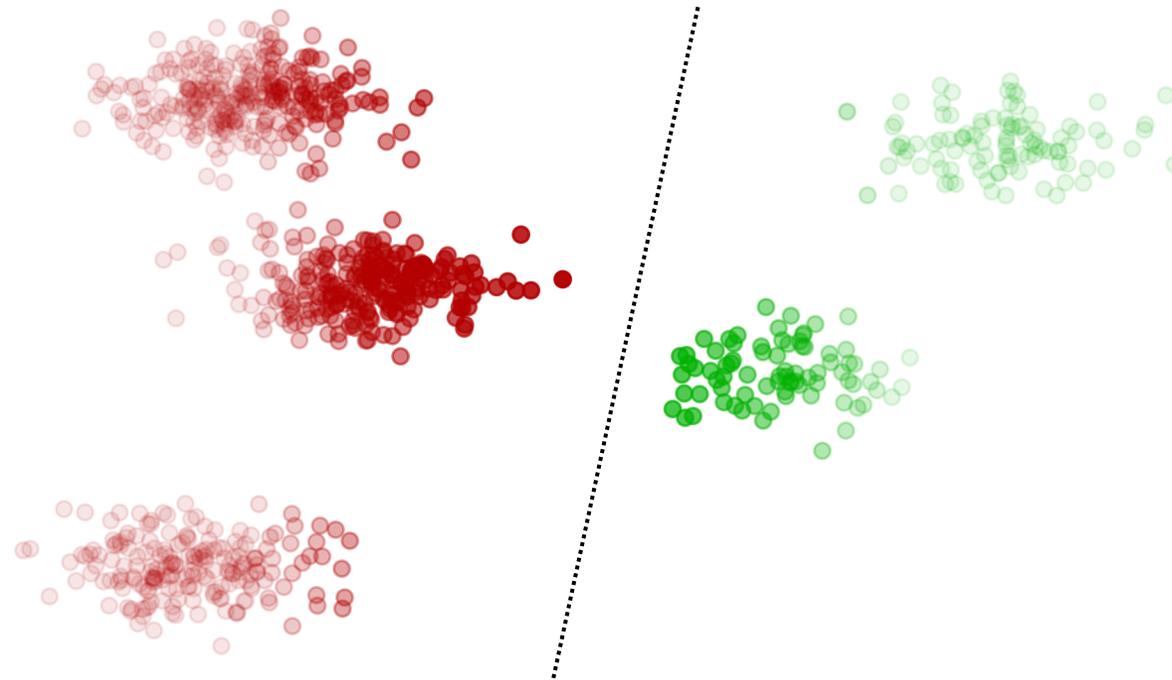
# Motivation



# Motivation

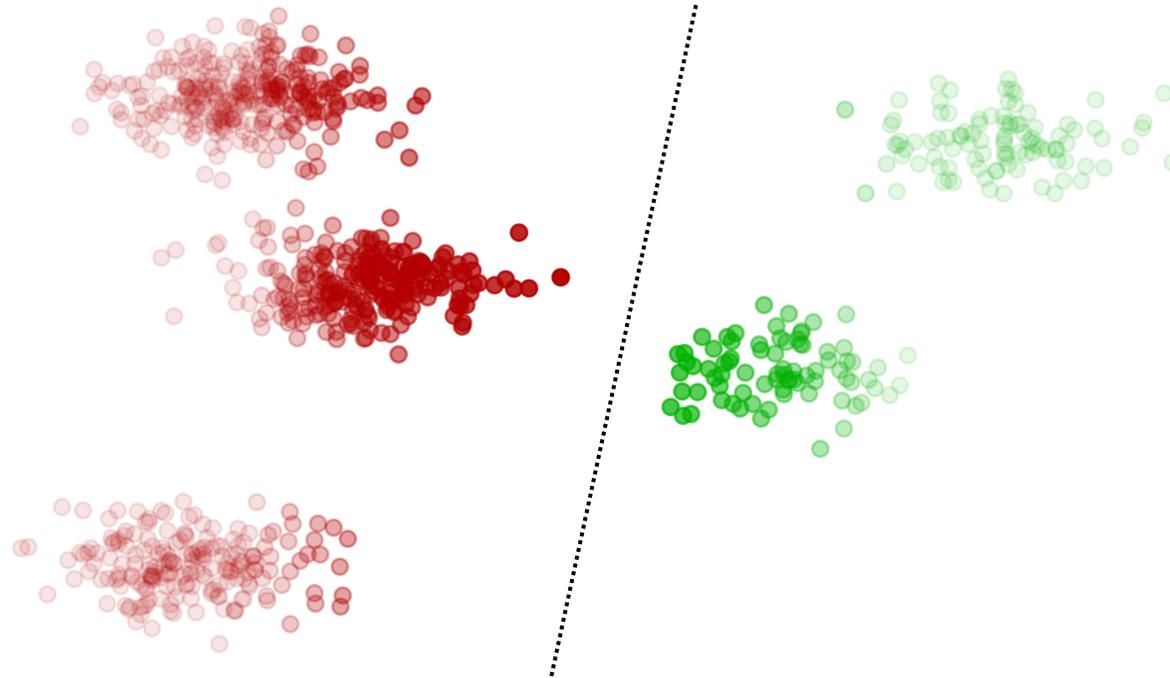


# Motivation



Adaptive importance sampling of data points

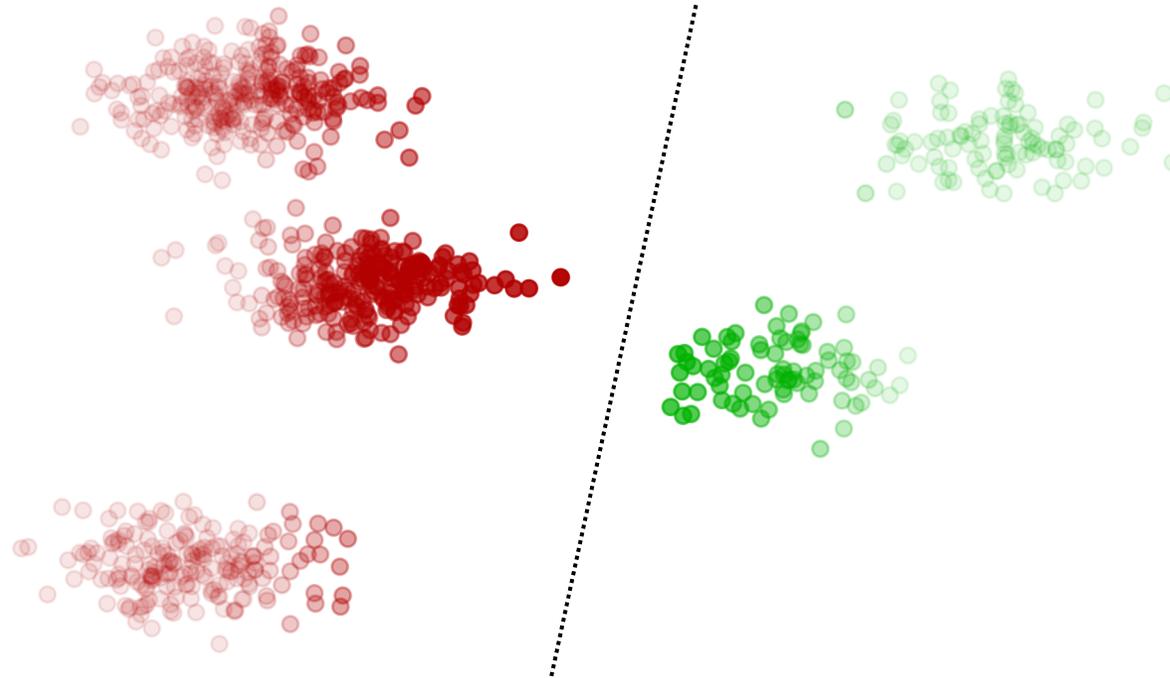
# Motivation



Adaptive importance sampling of data points

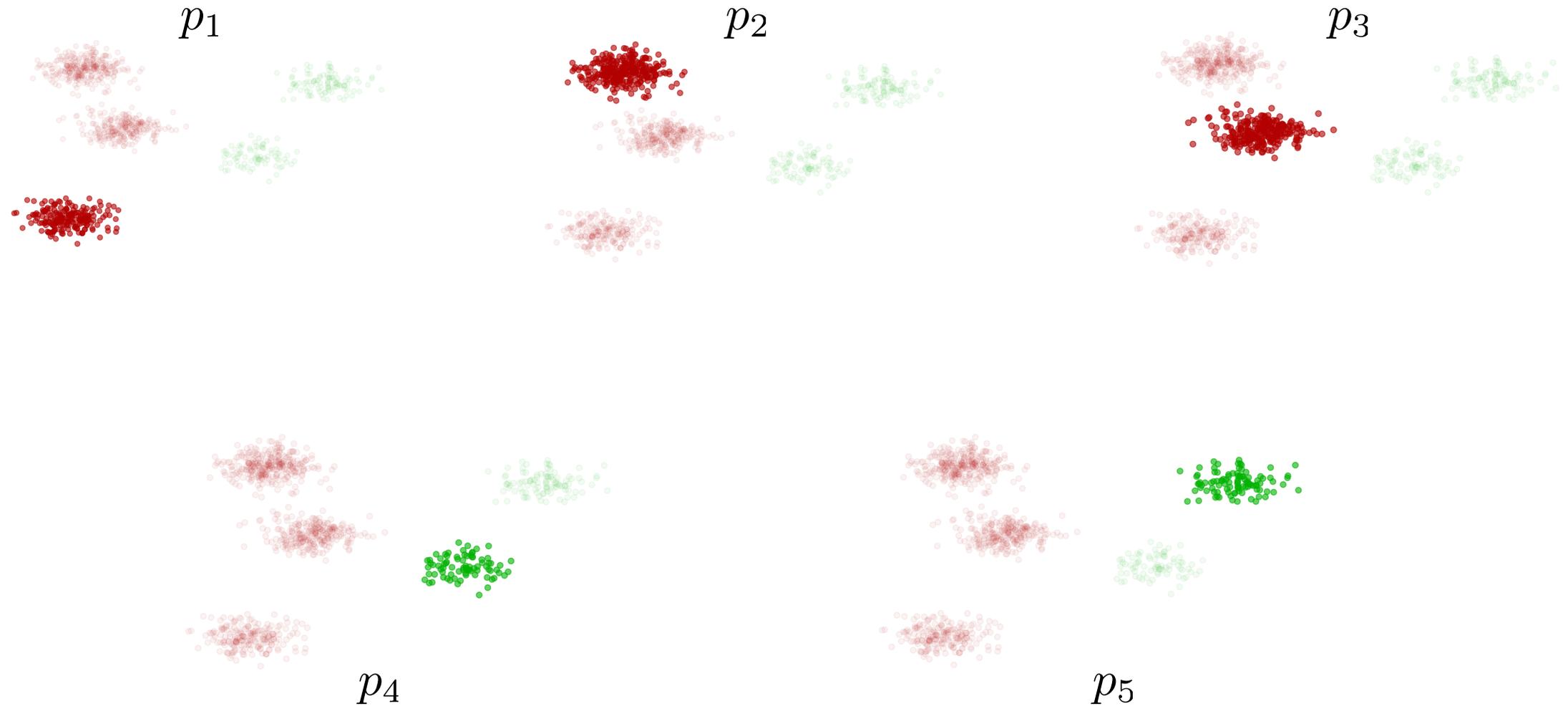
**Dependence on  $n$**

# Motivation

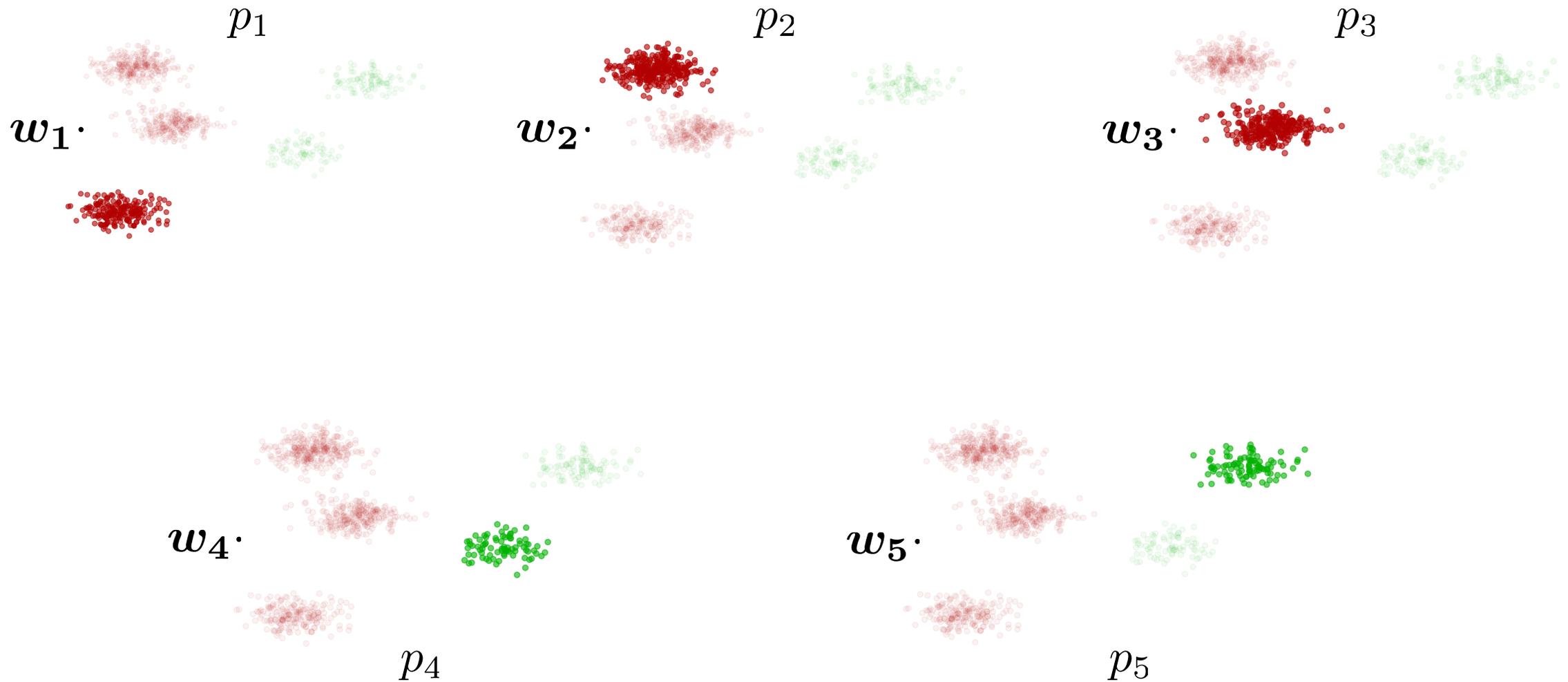


How can we exploit **prior knowledge** about the data?

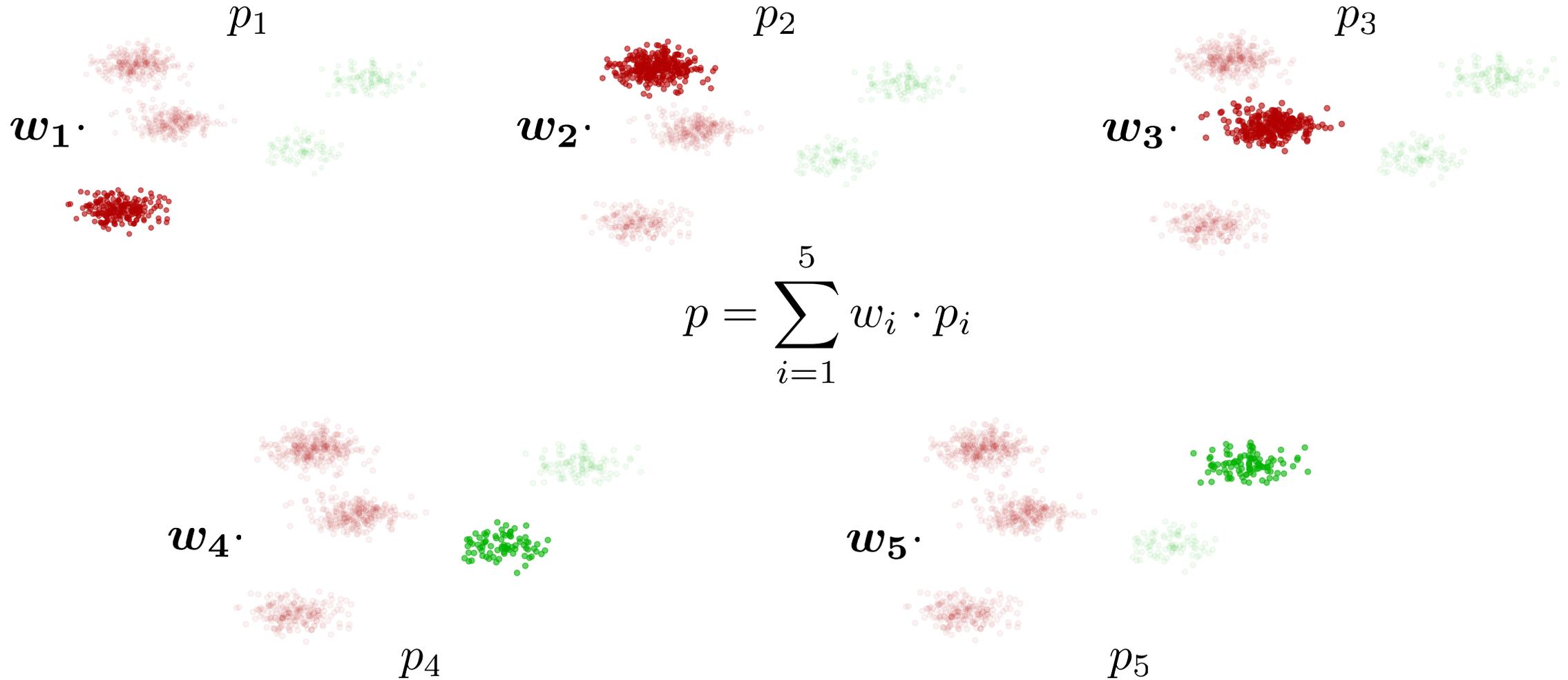
# Motivation



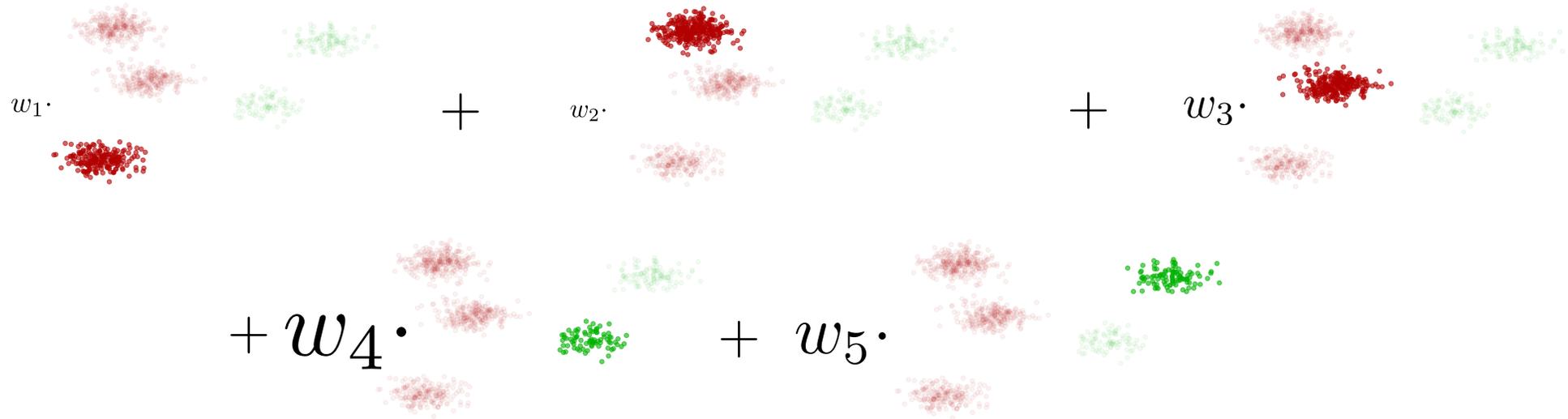
# Motivation



# Motivation

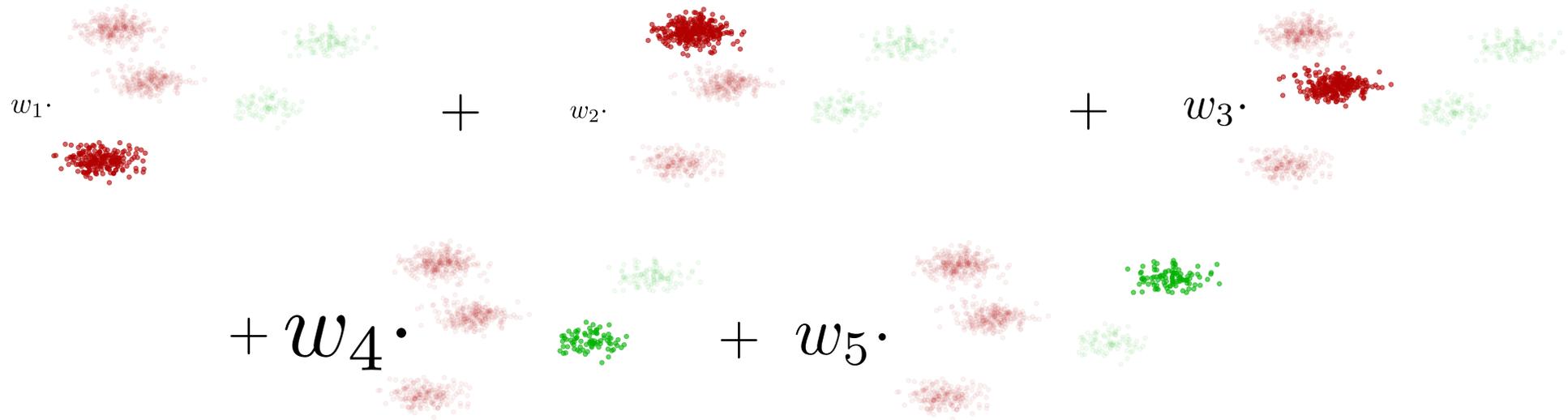


## Method



Adapt  $w_1, \dots, w_5$   $\longrightarrow$  Adaptive importance sampling with **mixtures**

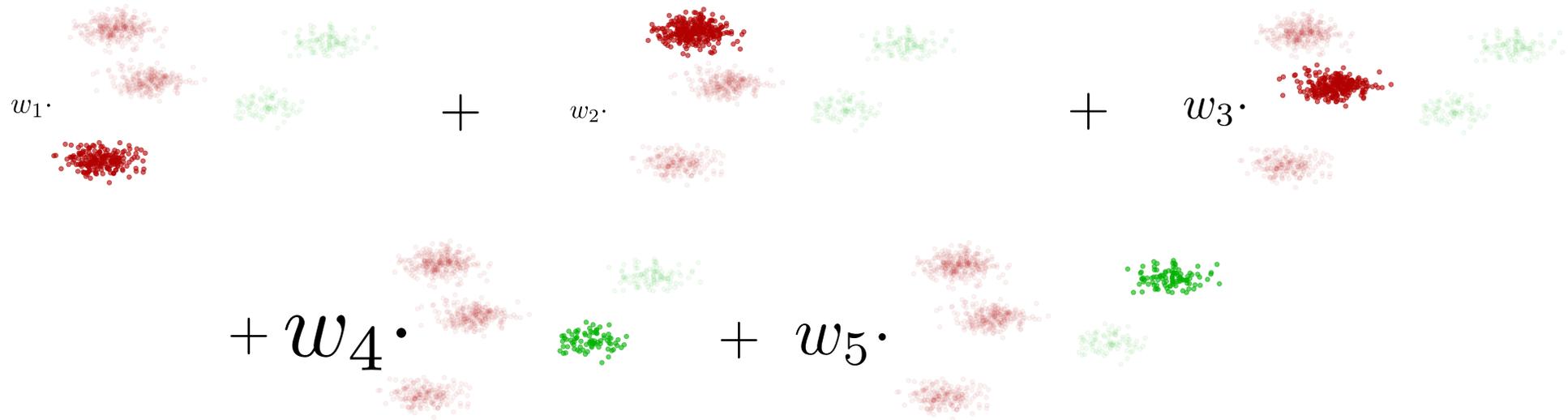
## Method



Adapt  $w_1, \dots, w_5 \longrightarrow$  Adaptive importance sampling with **mixtures**

Cost function  $f(w)$ : cumulative variance of loss estimates

## Method

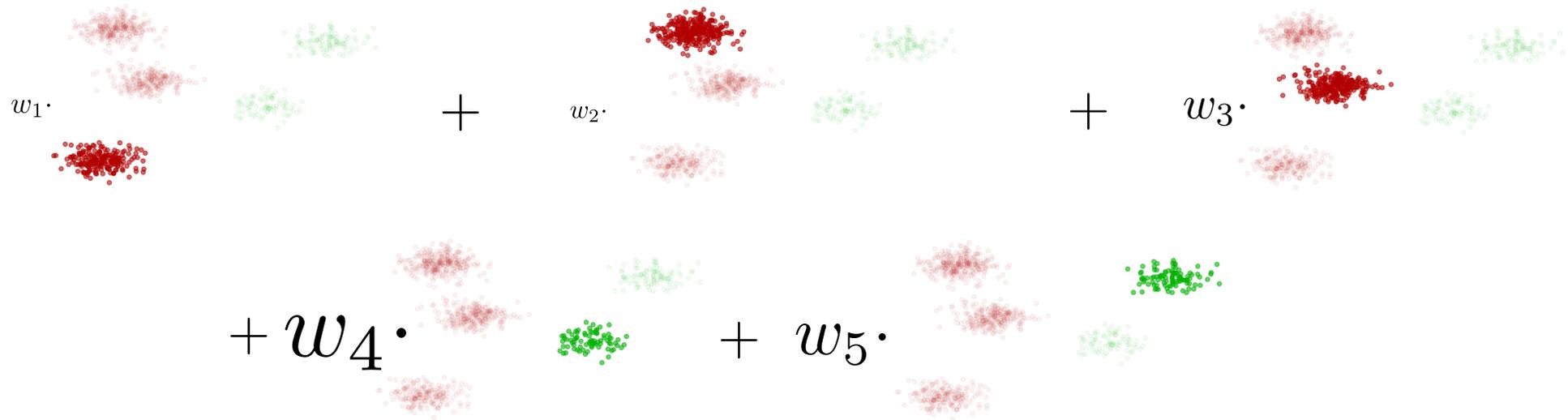


Adapt  $w_1, \dots, w_5$   $\longrightarrow$  Adaptive importance sampling with **mixtures**

Cost function  $f(w)$ : cumulative variance of loss estimates

No-regret algorithm for adapting the weights inspired by Online Newton Step

## Method



Adapt  $w_1, \dots, w_5$   $\longrightarrow$  Adaptive importance sampling with **mixtures**

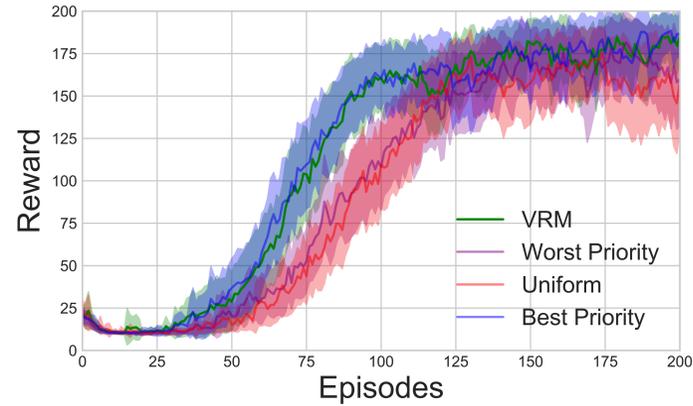
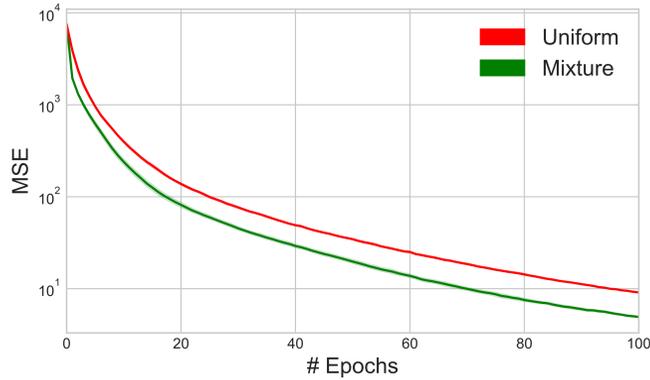
Cost function  $f(w)$ : cumulative variance of loss estimates

No-regret algorithm for adapting the weights inspired by Online Newton Step

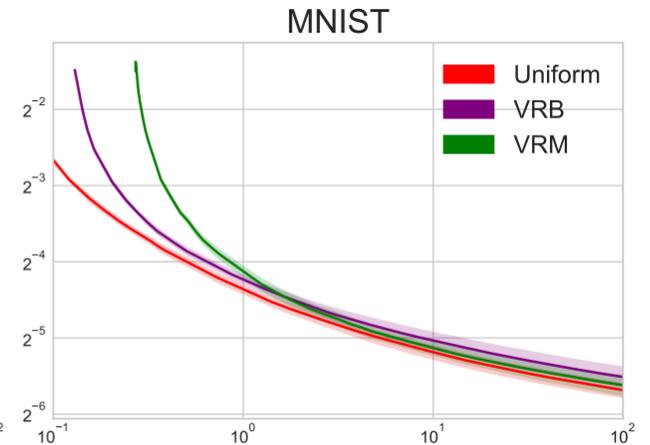
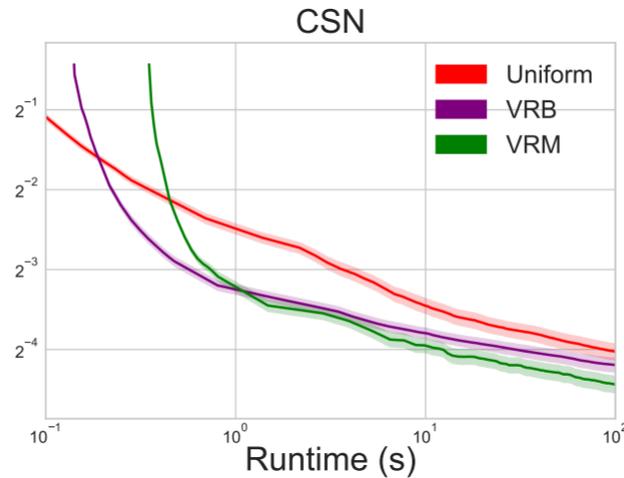
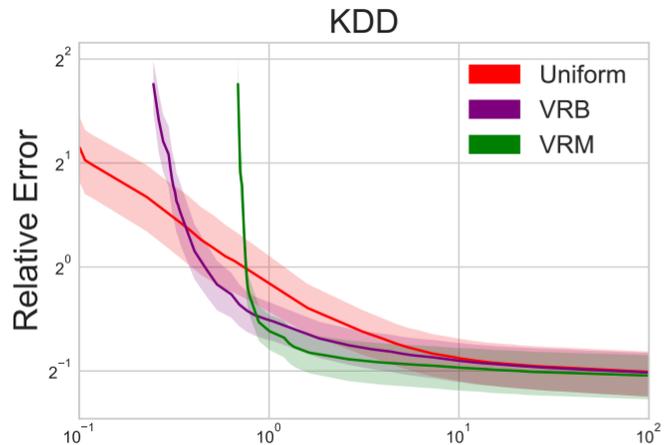
**Regret independent of  $n$**

# Results

k-DPP minibatch sampling in SGD for regression



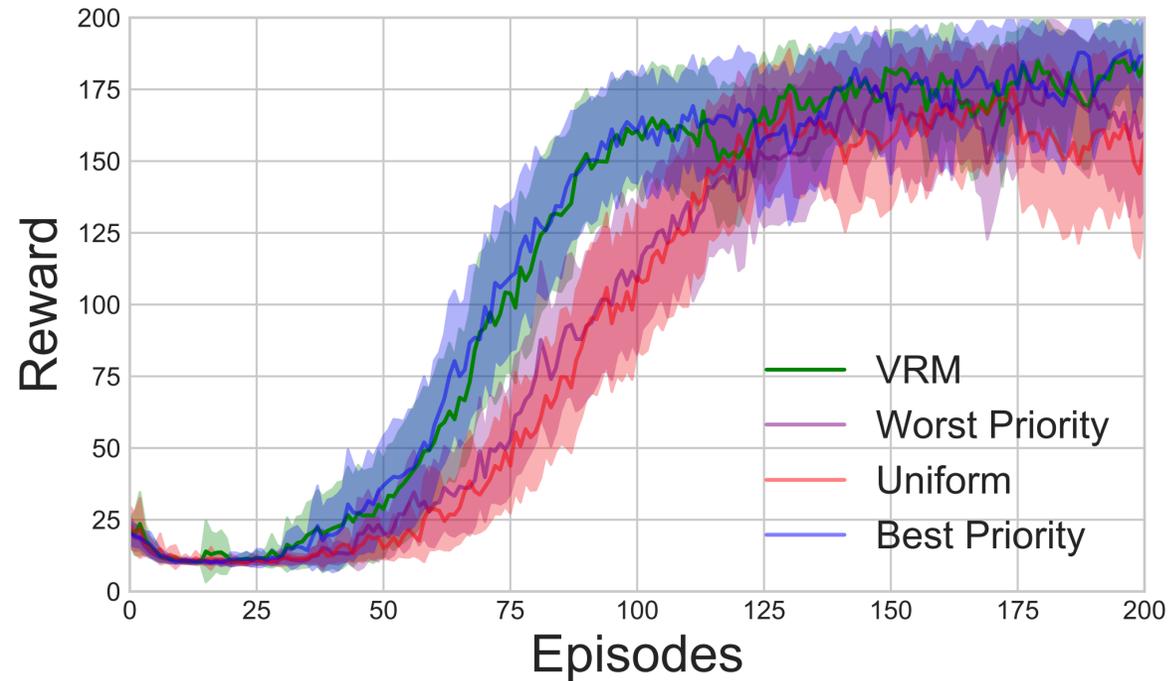
DQN with prioritized experience replay



Minibatch k-means

# Results

Prioritized experience replay:  $p_i \propto f(\delta_i, \epsilon, \alpha)$





# Online Variance Reduction with Mixtures

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