

Figure 1: Joint distribution $p(x_1, x_2)$ over two parameters

- 1. Rejection sampling in high dimensions: Consider a true distribution $p(\mathbf{x}) = \mathcal{N}(\mathbf{0}, \sigma_p^2 \mathbf{I})$ and a proposal distribution $q(\mathbf{x}) = \mathcal{N}(\mathbf{0}, \sigma_q^2 \mathbf{I})$ where $\sigma_q = 1.01\sigma_p$.
 - What is the value of *k* such that $kq \ge p$, if $x \in \mathbb{R}^{1000}$?
 - What is the acceptance rate of rejection sampling with this proposal distribution?
- 2. Why may Gibbs sampling have a hard time exploring the parameter distribution in Fig. 1?
- 3. In Metropolis-Hasting, we accept samples with probability α , where

$$\alpha(x', x^{(t)}) = \min\left(1, \frac{\tilde{p}(x')q(x^{(t)}|x')}{\tilde{p}(x^{(t)})q(x'|x^{(t)})}\right).$$

Show that p(x) is an invariant distribution of the Markov chain by showing detailed balance.

4. Consider the following model:

$$p(y) = \mathcal{N}(\mu, \tau^{-1})$$
$$p(\mu) = \mathcal{N}(\mu_0, s_0)$$
$$p(\tau) = \text{Gamma}(a, b)$$

Here, τ is the precision (inverse variance) and y is an observed data point.

- Draw the corresponding directed graphical model, including all deterministic parameters
- Write down the expressions for the joint distribution $p(y, \mu, \tau)$
- Write down expressions for the conditional distributions $p(\mu|y,\tau)$ and $p(\tau|\mu,y)$, which are required to apply Gibbs sampling to the posterior distribution $p(\mu,\tau|y)$