

# Problem Sheet 5

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```
• begin
•   using Pkg; Pkg.activate("."); Pkg.add(["Distributions", "Plots", "PyPlot",
"PlutoUI"])
•   using Distributions, LinearAlgebra
•   using PlutoUI
•   using Plots, StatsPlots; pyplot();
•   default(lw = 3.0, legendfontsize= 15.0)
• end
```

UndefVarError: TableOfContents not defined

1. top-level scope @ ( Local: 1

```
• TableOfContents()
```

## 1. Variational inference

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Assume we have  $n$  observations  $D = (x_1, \dots, x_n)$  generated independently from a Gaussian density  $\mathcal{N}(x|\mu, 1/\tau)$ , i.e.

$$p(D|\mu, \tau) = \left(\frac{\tau}{2\pi}\right)^{n/2} \exp \left[ -\frac{\tau}{2} \sum_{i=1}^n (x_i - \mu)^2 \right]$$

We also assume prior densities  $p(\mu|\tau) = \mathcal{N}(\mu|\mu_0, (\lambda_0\tau)^{-1})$  and  $p(\tau) = \text{Gamma}(\tau|a_0, b_0)$ .  $\lambda_0$  and  $\mu_0$  as well as  $a_0, b_0$  are given hyper parameters.

Our goal is to approximate the posterior density  $p(\mu, \tau|D)$  by a **factorising density**  $q(\mu, \tau) = q_1(\mu)q_2(\tau)$  which minimises the variational free energy

$$F[q] = \int q(\mu, \tau) \ln \frac{q(\mu, \tau)}{p(\mu, \tau, D)} d\mu d\tau$$

**(a) [MATH] Show that the optimal  $q_1(\mu)$  is a *Gaussian density* and give expressions for the mean and variance in terms of expectations with respect to  $q_2$ .**

(b) [MATH] Show that the optimal  $q_2(\tau)$  is a *Gamma density* and give expressions for the parameters in terms of expectations with respect to  $q_1$ .

**Tip**

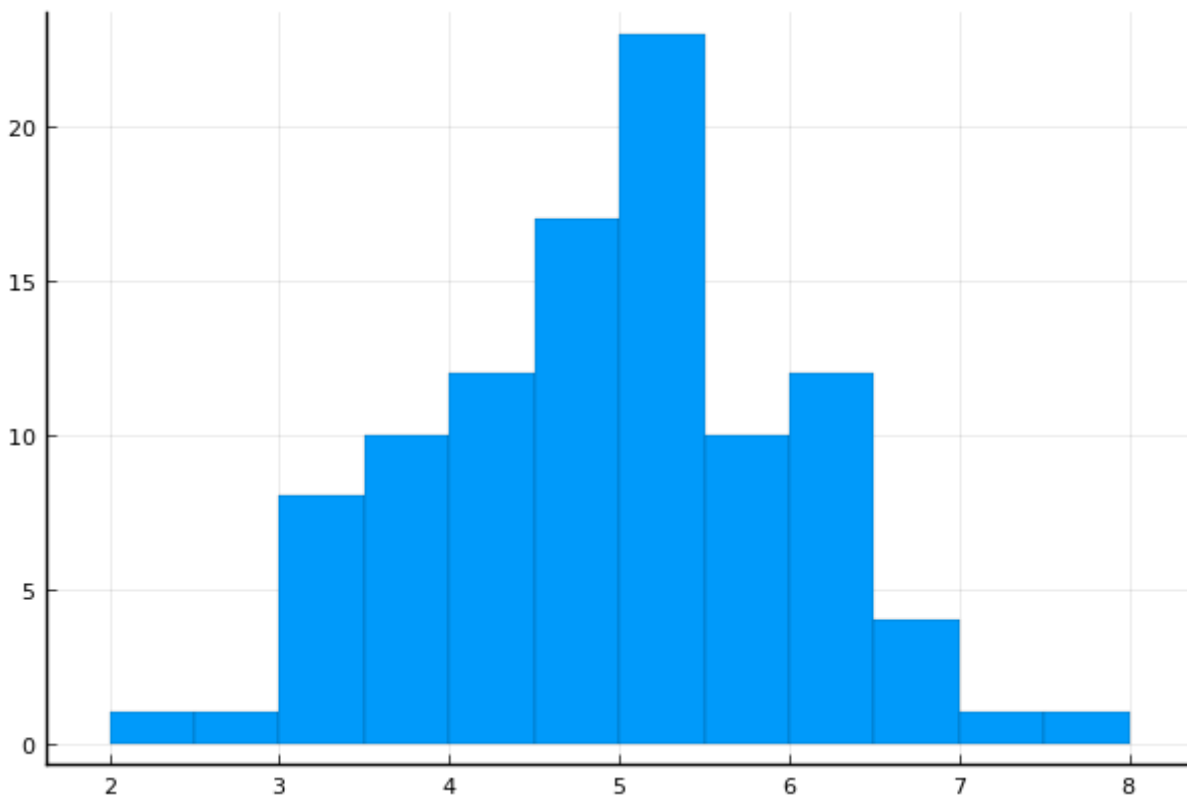
You can use the following results which follow from the derivations given in the lecture

$$q_1(\mu) \propto \exp [E_\tau [\ln p(\mu, \tau, D)]]$$
$$q_2(\tau) \propto \exp [E_\mu [\ln p(\mu, \tau, D)]]$$

*Fill in your answer here or on paper*

c) [CODE] From the generated dataset implement a coordinate ascent scheme, updating variational parameters of  $\tau$  and  $\mu$  in an alternated way.

```
• begin
•   # We create some data
•   N = 100
•   μ₀ = 5.0
•   λ₀ = 2.0
•   a₀ = 1.0
•   b₀ = 2.0
•   τ = rand(Gamma(a₀, b₀))
•   μ = rand(Normal(μ₀, inv(sqrt(τ * λ₀))))
•   x = rand(Normal(μ, inv(sqrt(τ))), N)
• end;
```



syntax: invalid syntax (incomplete #<julia: "incomplete: premature end of input">)

1. top-level scope @ none:1

• `expμ(x, λ0, μ0, n) = ## FILL IN ## Return Eq[μ]`

syntax: invalid syntax (incomplete #<julia: "incomplete: premature end of input">)

1. top-level scope @ none:1

• `varμ(eτ, λ0, n) = ## FILL IN ## Return Varq[μ]`

exp<sub>τ</sub> (generic function with 1 method)

• `expτ(a, b) = a / b`

syntax: invalid syntax (incomplete #<julia: "incomplete: premature end of input">)

1. top-level scope @ none:1

• `an(a0, n) = ## FILL IN ## Return first parameter of q(τ)`

syntax: invalid syntax (incomplete #<julia: "incomplete: premature end of input">)

1. top-level scope @ none:1

• `bn(b0, x, eμ, varμ, λ0, μ0) = ## FILL IN ## Return second parameter of q(τ)`

coordinate\_ascent (generic function with 5 methods)

```
• function coordinate_ascent(T, a = 1.0, b = 1.0, μ = 1.0, σ² = 1.0)
•   as = vcat(a, zeros(T))
•   bs = vcat(b, zeros(T))
•   μs = vcat(μ, zeros(T))
•   σs = vcat(σ², zeros(T))
•   for i in 2:T+1
•     a = a_n(a_0, N); as[i] = a;
•     b = b_n(b_0, x, μ, σ², λ_0, μ_0); bs[i] = b
•     e_τ = expec_τ(a, b)
•     μ = expec_μ(x, λ_0, μ_0, N); μs[i] = μ
•     σ² = var_μ(e_τ, λ_0, N); σs[i] = σ²
•   end
•   return as, bs, μs, σs
• end
```

UndefVarError: a\_n not defined

1. coordinate\_ascent(::Int64, ::Float64, ::Float64, ::Float64, ::Float64) @ **Other: 7**
2. coordinate\_ascent(::Int64) @ **Other: 2**
3. top-level scope @ **Local: 3**

```
• begin
•   T = 20
•   as, bs, mus, σs = coordinate_ascent(T)
•   plt1 = plot([as bs], label = ["a" "b"])
•   plt2 = plot([mus σs], label = ["μ" "σ²"])
•   plot(plt1, plt2)
• end
```

UndefVarError: a\_n not defined

1. coordinate\_ascent(::Int64, ::Int64, ::Int64, ::Int64, ::Float64) @ **Other: 7**
2. top-level scope @ **Local: 3**

```
• begin
•   T_hard = 20
•   as_hard, bs_hard, mus_hard, σs_hard = coordinate_ascent(T_hard, 1000, 2000, 300,
1e3)
•   p1 = plot([as_hard bs_hard], label = ["a" "b"], yaxis = :log)
•   p2 = plot([mus_hard σs_hard], label = ["μ" "σ²"], yaxis = :log)
•   plot(p1, p2)
• end
```