

## Distance Exam Bayesian Learning (732A73/732A91/TDDE07), 6 hp

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Time:	8-12
Allowable material:	All aids are permitted during the exam with the following exceptions: <ul style="list-style-type: none"><li>• You may not communicate with anyone else except the responsible teacher</li><li>• You may not look at the solutions of any other students.</li></ul>
Teacher:	Bertil Wegmann. Contact via email at <a href="mailto:bertil.wegmann@liu.se">bertil.wegmann@liu.se</a>
Exam scores:	Maximum number of credits on the exam: 40.
Grades (732A73/732A91):	A: 36 points B: 32 points C: 24 points D: 20 points E: 16 points F: <16 points
Grades (TDDE07):	5: 34 points 4: 26 points 3: 18 points U: <18 points

### INSTRUCTIONS:

For full information, see the document *Info distance exam.pdf*.

You should submit your solutions via the Submission in LISAM.

Your submitted solutions must contain the following two things:

- A PDF named *ComputerSol.pdf* containing your computer-based solutions
- A PDF named *PaperSol* containing your hand-written solutions (when asked to answer on **Paper**)

Full score requires clear and well motivated answers.

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## 1. DERIVATIONS AND COMPARING POSTERIOR DISTRIBUTIONS (*Total credits: 21p*)

Problems (a), (b) and (c) should only be solved on [Paper](#). Let  $x|\theta$ , where  $x = (x_1, \dots, x_n)$ , be an independent sample from a Poisson distribution and assume that the prior for  $\theta$  has the following density:

$$p(\theta) = \frac{1}{16}\theta^2 e^{-0.5\theta}, \theta > 0.$$

- Credits: 3p.* Derive the posterior distribution for  $\theta$ .
- Credits: 2p.* Compute the Bayes point estimate of  $\theta$  for  $n = 15$ ,  $\sum_{i=1}^n x_i = 75$ , by assuming the quadratic loss function.
- Credits: 5p.* Derive the marginal likelihood for an independent sample  $x|\theta$  from the model above. An expression is enough, you don't need to recognize any distributional family from the end result.
- Credits: 4p.* Write a function in R that computes the log posterior distribution of  $\theta$  for an independent sample  $x|\theta$ . Then, use this function to plot the posterior distribution of  $\theta$  based on the data in 1(b).
- Credits: 3p.* Use numerical optimization to obtain a normal approximation to the posterior distribution of  $\theta$  based on the data in 1(b). Use the `lines` command in R to plot this approximate posterior in the same graph as the posterior obtained in 1(d). [Hints: use the argument `lower=3` in `optim`, and `method=c("L-BFGS-B")`]. Is the posterior approximation accurate?
- Credits: 4p.* Assume that a person named Gunnar told you that the maximum value of an independent Poisson distributed sample  $x|\theta$  of size  $n = 15$  is  $T(x) = \max_{x_i} = 14$ . Calculate the posterior predictive p-value,  $\Pr(T(x^{rep}) \geq T(x))$ , by simulating from the posterior predictive density  $p(T(x^{rep})|x)$  using the model and data in this problem 1. Is it reasonable to think that the maximum value of 14 from Gunnar originates from the Poisson distribution in this problem?

## 2. LOGIT REGRESSION PROBLEMS (*Total credits: 9p*)

The file `Disease.RData`, which is loaded by the code in `ExamData.R` contains data on 32 patients in a country. For each patient  $i$  we have observed  $y_i = 1$  if patient  $i$  has a certain disease and 0 otherwise,  $x_{1i} = \text{measured value on a cell property for patient } i$  and  $x_{2i} = 1$  if patient  $i$  has a certain property and 0 otherwise. The variable  $x_1$  has been standardized to mean 0 and variance 1. The data matrix also contains a column `const` with ones to include an intercept in the model. Now, use `BayesLogitReg.R` in `ExamData.R` to sample from the joint posterior distribution of the model parameters in the following logistic regression model:

$$\Pr(y_i = 1 \mid x_1, x_2) = p_i = \frac{\exp(\beta_0 + \beta_1 \cdot x_{1i} + \beta_2 \cdot x_{2i})}{1 + \exp(\beta_0 + \beta_1 \cdot x_{1i} + \beta_2 \cdot x_{2i})}$$

Simulate 10000 draws from the joint posterior. Use the following prior distribution for  $\beta = (\beta_0, \beta_1, \beta_2)$ :

$$\beta \sim N(0, 4^2 I_3)$$

- Credits: 1p.* Compute the 90% equal tail credible interval for  $\beta_1$  and interpret it.
- Credits: 1p.* Compute the posterior probability that  $\beta_2 > 0$  and interpret it in terms of the effect from variable  $x_2$  on  $p$ .
- Credits: 1p.* Compute the joint posterior probability that both  $\beta_1 > 0$  and  $\beta_2 > 0$ .
- Credits: 3p.* Consider a patient  $j$  with  $x_{1j} = x_{2j} = 0$ . Plot the posterior distribution of  $p_j$  and compute the posterior probability that  $p_j > 0.5$  for this patient.
- Credits: 3p.* Consider a patient  $k$  with  $x_{2k} = 1$ . Compute 95 % equal tail posterior probability intervals for  $p_k$  on a grid of values of  $x_1$ . The grid of values of  $x_1$  shall span between the lowest and highest values of  $x_1$  with the distance 0.01 between any pair of values on the grid. Plot the lower and upper limits of the posterior probability intervals as a function of  $x_1$ .

3. NORMAL MODELING (*Total credits: 10p*)

Problem (a) should only be solved on [Paper](#). Let  $x_1, \dots, x_n | \mu$  be an independent sample from a normal distribution with unknown mean  $\mu$  and variance 50. Assume that a Bayesian data analysis of  $\mu$  with  $n = 10$  observations and  $\bar{x} = 90$  resulted in the following posterior:  $\mu | x_1, \dots, x_n \sim N(92, 2^2)$ .

- (a) *Credits: 3p.* Derive the normal prior distribution that have been used for this Bayesian data analysis.
- (b) *Credits: 2p.* Simulate draws from the posterior predictive density of a new observation  $x_{n+1}$  and plot the posterior predictive distribution.
- (c) *Credits: 5p.* Assume that  $x_i$  is the revenue for company  $i$  under a given time period in a certain industry. The utility function for a company in the industry spending  $c$  MSEK on advertisements is given by:

$$U(\mu, c) = 60 + \sqrt{c} \log(\mu) - c.$$

How much money  $c$  should a company spend on advertisements from a Bayesian perspective? Use at least 10000 draws in your simulations. Motivate your answer with a figure.

GOOD LUCK!  
BEST, BERTIL