

Examiner: Jonas Wallin, tel 772 53 23.

Allowed aids: None.

Correct, and **well motivated solutions** gives points indicated within the parentheses at each exercise. Total number of points is 49. The grade limits are 3:19, 4:28 and 5:38 points, respectively and G:19, and VG:35.

1. Show that for the bivariate distribution  $\pi(x_1, x_2)$ , the two stage Gibbs sampler has  $\pi$  as stationary distribution. (5p)
2. Show that Multinomial resampling strategy adds no bias to self normalizing importance sampling. (5p)
3. (a) Assume that  $y \sim \pi(y|\theta) = Po(\theta)$ , show that  $\Gamma(\alpha, \beta)$  is a conjugate prior for  $\theta$ . (3p)  
 (b) Assume that  $y \sim \pi(y|\theta) = NB(r, \theta)$ , where  $r$  is known. Show that  $Beta(\alpha, \beta)$  is a conjugate prior for  $\theta$ . (3p)  
 (c) Let  $y \sim \mathcal{N}(\theta, \sigma^2)$ , show that  $\pi(\theta) \sim \mathcal{N}(\mu_\theta, \sigma_\theta^2)$  is a conjugate prior to  $\theta$ . Derive the parameters for the posterior distribution given that one observes  $y_1$ . (4p)
4. In this exercise we will be studying a Markov chain  $\{X_i\}_{i=0}^N$  with transition kernel  $q(x|y)$  and initial distribution  $\pi_0$ 
  - (a) We are interested in estimating  $\mathbb{P}(X_{0:n} \in A)$ , using a SISR method. What function,  $h(X_{0:n})$  should we use to estimate the probability in SISR algorithm? (3p)
  - (b) Write down the pseudo algorithm of the SISR sampler. (Either general form or special case when the density of interest is a Markov chain) (6p)
5. We are interested in sampling from the posterior distribution of the following model:

$$\begin{aligned}\pi(\beta_k) &= \Gamma(\beta_k, 1, 0.1), k = 0, 1, \\ \pi(p) &\propto 1, \\ \pi(\mathbf{y}|\beta_0, \beta_1, p, \alpha) &= \prod_{i=1}^n ((1-p)\Gamma(y_i; \alpha, \beta_0) + p\Gamma(y_i; \alpha, \beta_1)),\end{aligned}$$

where  $\alpha$  is assumed to be known constant. The goal of this exercise is to generate the posterior distribution using MCMC.

- (a) Suggest auxiliary variables  $\mathbf{z}$  which makes it possible to Gibbs sample the posterior. Derive the joint distribution  $\pi(\mathbf{z}, \mathbf{y}, \beta_0, \beta_1, p)$ . (4p)
- (b) Derive the conditional distribution for each variable in its parametric form. (4p)
- (c) Instead of Gibbs sampling one can use Metropolis Hasting random walk. Derive the acceptance probability as a function  $\beta_0, \beta_1, p$ . (4p)
6. Assume that one want to estimate  $\mathbb{E}_f[x]$ , where  $f(x) \propto e^{-\frac{x^2}{2}} \mathbb{I}(x > c)$ ,  $c > 0$ , using IS. For proposal distribution it suggested using  $g(x) \propto e^{-\frac{x^2}{2\sigma^2}}$ . Can one use this proposal for any  $\sigma$ ? (8p)

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- Normal distribution.

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Notation	$\mathcal{N}(x; \mu, \sigma^2)$
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Suport	$x \in \mathbb{R}$
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parameter	$\sigma^2 > 0, \mu \in \mathbb{R}$
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pdf	$\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$
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- Beta distribution.

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Notation	$\mathcal{B}(x; \alpha, \beta)$
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Suport	$x \in [0, 1]$
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parameter	$\alpha > 0, \beta > 0$
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pdf	$\frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1}$
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- Gamma distribution.

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Notation	$\Gamma(x; \alpha, \beta)$
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Suport	$x \in [0, \infty)$
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parameter	$\alpha > 0, \beta > 0$
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pdf	$\frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$
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- inverse Gamma distribution.

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Notation       $IG(x; \alpha, \beta)$

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Suport       $x \in [0, \infty)$

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parameter       $\alpha > 0, \beta > 0$

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pdf       $\frac{\beta^\alpha}{\Gamma(\alpha)} x^{-\alpha-1} e^{-\beta x^{-1}}$

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- Exponential distribution.

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Notation       $exp(x; \lambda)$

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Suport       $x \in [0, \infty)$

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parameter       $\lambda > 0$

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pdf       $\lambda e^{-\lambda x}$

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cdf       $1 - e^{-\lambda x}$

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- Raised cosine

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Notation       $Rcos(x; \mu)$

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Suport       $x \in [\mu - 1, \mu + 1]$

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parameter       $\mu \in \mathbb{R}$

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pdf       $\frac{1}{2}[1 + \cos(\pi(x - \mu))]$

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- Poisson

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Notation  $Pois(x; \lambda)$

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Suport  $x \in \mathbb{N}$

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parameter  $\lambda > 0$

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pdf  $\frac{\lambda^x}{x!} e^{-\lambda}$

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- Binomial

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Notation  $Bin(x; n, p)$

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Suport  $x \in \{0, 1, \dots, n\}$

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parameter  $p \in [0, 1], n \in \mathbb{N}$

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pdf  $\binom{n}{x} (1-p)^{n-x} p^x$

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- Negative binomial

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Notation  $NB(x; r, p)$

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Suport  $x \in \{0, 1, \dots\}$

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parameter  $p \in [0, 1], r \in \mathbb{N}$

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pdf  $\binom{x+r-1}{x} (1-p)^r p^x$

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