



[Course](#) > [Unit 5 Bayesian statistics](#) > [Homework 9: Bayesian Statistics](#) > 3. Jeffreys prior

### 3. Jeffreys prior

*Note: The concepts discussed in the recitation on Jeffreys prior may be helpful to you for these homework exercises.*

#### Instructions:

For each of the following statistical models, compute Jeffreys prior distribution and determine whether it is proper or not.

(a)

2/2 points (graded)

For a family of distribution  $\{\text{Ber}(p)\}_{p \in (0,1)}$ , Jeffreys prior is proportional to:

$\pi_j(p) \propto$

$1/\sqrt{p(1-p)}$

✓ Answer:  $p^{-0.5}(1-p)^{-0.5}$

$\frac{1}{\sqrt{p(1-p)}}$

Therefore, the Jeffreys prior is:

☒ Proper

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☐ Improper



**Solution:**

Recall that  $\pi_j \propto \sqrt{\det(I(\theta))}$ .

$$I(p) = \frac{1}{p(1-p)}$$

$$\pi_j \propto \frac{1}{\sqrt{p(1-p)}}$$

Therefore, the prior is proper; **Beta (0.5, 0.5)**.

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You have used 2 of 3 attempts

**i** Answers are displayed within the problem

(b)

2/2 points (graded)

For a family of distribution  $\{\text{Exp}(\lambda)\}_{\lambda>0}$ , Jeffreys prior is proportional to:

$\pi_j(\lambda) \propto$

1/lambda

✓ Answer: 1/lambda

$\frac{1}{\lambda}$

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Therefore, the Jeffreys prior is:

☐ Proper

☒ Improper



**Solution:**

Recall that  $\pi_j \propto \sqrt{\det(I(\lambda))}$ .

$$I(\lambda) = \frac{1}{\lambda^2}$$

$$\pi_j \propto \frac{1}{\lambda}$$

Since  $\frac{1}{\lambda}$  integrates to infinity, the prior is improper.

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You have used 1 of 3 attempts

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**i** Answers are displayed within the problem

---

(c)

2/2 points (graded)

For a family of distribution  $\{\text{Poiss}(\lambda)\}_{\lambda>0}$ , Jeffreys prior is proportional to:

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$$\pi_j(\lambda) \propto$$

$$1/\sqrt{\lambda}$$

✓ Answer:  $\lambda^{-1/2}$

$$\frac{1}{\sqrt{\lambda}}$$

Therefore, the Jeffreys prior is:

☐ Proper

☒ Improper



**Solution:**

Recall that  $\pi_j \propto \sqrt{\det(I(\lambda))}$ .

$$I(\lambda) = \frac{1}{\lambda}$$

$$\pi_j \propto \frac{1}{\sqrt{\lambda}}$$

Since  $\frac{1}{\sqrt{\lambda}}$  integrates to infinity, the prior is improper.

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You have used 1 of 3 attempts

❗ Answers are displayed within the problem

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## (d) Properties of Jeffreys prior

0/1 point (graded)

For each of the statements below about Jeffreys prior, determine whether it is true or false. Select all the true statements.

☒ It allows us to reflect our prior belief about the possible hypotheses. In other words, Jeffreys prior is not obtained from the statistical model alone.

☐ Jeffreys prior is always proper.

☒ For a Bernoulli statistical model, the Jeffreys prior  $\pi(\theta_1)$ , computed from using  $\theta_1 = p^2$  as the parameter (i.e. the model is  $\text{Ber}(\theta_1) = \text{Ber}(p^2)$ ), and the Jeffreys prior  $\tilde{\pi}(\theta_2)$ , computed from using  $\theta_2 = p^3$  as the parameter (i.e. the model is  $\text{Ber}(\theta_2) = \text{Ber}(p^3)$ ), satisfy  $\mathbf{P}_{\pi(\theta_1)}(a^2 < \theta_1 < b^2) = \mathbf{P}_{\tilde{\pi}(\theta_2)}(a^3 < \theta_2 < b^3)$  for any  $0 < a < b < 1$ . That is the probability of  $\theta_1$  being between  $a^2$  and  $b^2$  under the distribution  $\pi(\theta_1)$  is equal to the probability of  $\theta_2$  being between  $a^3$  and  $b^3$  under the distribution  $\tilde{\pi}(\theta_2)$ , for any pair  $a < b$  within  $(0, 1)$ . ✓

✗

### Solution:

- **The first choice is false.** Recall that Jeffreys prior is obtained from the model, there is nothing we reflect about our prior belief. Hence, the first choice is false.
- **The second choice is false.** We have seen examples where it is not necessarily a proper prior as it does not have a finite integral.
- **The third choice is true.** The last choice is true because Jeffreys prior is invariant under reparametrization.

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You have used 2 of 2 attempts

📘 Answers are displayed within the problem

## (e) Review: Reparametrization in the frequentist view

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0/1 point (graded)

In the previous units, three of the frequentist methods of estimation we've covered are the maximum likelihood estimation (MLE), the method of moments, and M-estimation. Let our original parameter is  $\theta$ , and suppose that our original estimator produces a unique estimate  $\theta^*$ . We then apply a bijective transformation  $f(\theta) = \eta$ . For which of the three frequentist methods would the estimator applied to the transformed values  $\eta^*$  be equal to  $f(\theta^*)$ ?

☒ MLE ✓

☐ method of moments ✓

☒ M-estimation ✓



### Solution:

The answer is that the estimator applied to the transformed values  $\eta^*$  will always be equal to  $f(\theta^*)$ . This is because in the frequentist approach, a true parameter is assumed, and thus all our estimator functions (of the observation data) will correspond to a particular parameter value. This value can be converted through different parametrizations, and it will correspond to the exact same value as long as the original estimator produces a unique estimate.

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**i** Answers are displayed within the problem

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1

Generating Speech Output using SymPy @vars x λ k L1 = λ \* exp(- λ \* x) IL1 = log(L1) dL1 λ = diff(IL1,λ) dL2 λ = simplify(diff(dL1 λ,λ)) L1 = λ^x \* exp(-λ)/k IL1 = log(L1) dL1 λ = simplify(diff(IL1,λ))...

? <u>Not sure about the answer of question (b)</u>	2
(e) "estimator applied to the transformed values $\eta$ "? I don't understand how an estimator could be applied to the new parameter $\eta$ . Isn't an estimator for a parameter a function of the data (only the data) to calculate or estim...	13
? [STAFF] Please clarify the question in part (e).	2
[Staff] - Minor typo/gramatical error - Part (e). The sentence reads: ..."Let our original parameter $s^*$ $\theta$ ..." should be "Let our original parameter $\eta$ $\theta$ ...".	2
e) Unsure where I'm going wrong. I reasoned that one of them seems to be preserved only under linearity and another seems to include this as a special case (per the examples we saw in lecture / in the slides)...	1

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