

Computer Vision

Simon J.D. Prince (2012)

p. 40

3.6 Normal-scaled inverse gamma distribution

The *normal-scaled inverse gamma distribution* (figure 3.6) is defined over a pair of continuous values μ, σ^2 , the first of which can take any value and the second of which is constrained to be positive. As such it can define a distribution over the mean and variance parameters of the normal distribution.

The normal-scaled inverse gamma has four parameters $\alpha, \beta, \gamma, \delta$ where α, β , and γ are positive real numbers but δ can take any value. It has pdf:

$$Pr(\mu, \sigma^2) = \frac{\sqrt{\gamma}}{\sigma\sqrt{2\pi}} \frac{\beta^\alpha}{\Gamma[\alpha]} \left(\frac{1}{\sigma^2}\right)^{\alpha+1} \exp\left[-\frac{2\beta + \gamma(\delta - \mu)^2}{2\sigma^2}\right], \quad (3.13)$$

or for short

$$Pr(\mu, \sigma^2) = \text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta]. \quad (3.14)$$

正規分布 (normal distribution) に現れるパラメーター、平均 (mean) μ と分散 (variance) σ^2 に関する事前分布 (prior) として、よく用いられる。

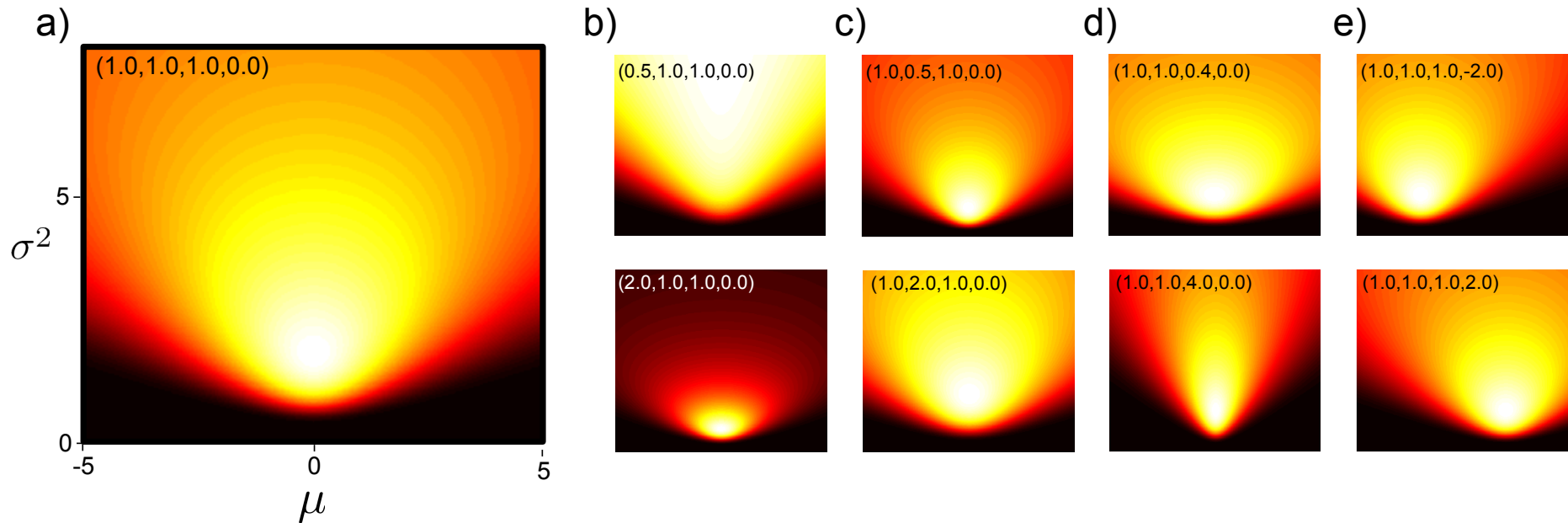


Figure 3.6 The normal-scaled inverse gamma distribution defines a probability distribution over bivariate continuous values μ, σ^2 where $\mu \in [-\infty, \infty]$ and $\sigma^2 \in [0, \infty]$. a) Distribution with parameters $[\alpha, \beta, \gamma, \delta] = [1, 1, 1, 0]$. b) Varying α . c) Varying β . d) Varying γ . e) Varying δ .

Problem 3.8 Calculate an expression for the mode (position of the peak in μ, σ^2 space) of the normal scaled inverse gamma distribution in terms of the parameters $\alpha, \beta, \gamma, \delta$.

$$\begin{aligned}
& \text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta] \\
&= \frac{\sqrt{\gamma}}{\sigma\sqrt{2\pi}} \frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{1}{\sigma^2}\right)^{\alpha+1} \exp\left[-\frac{2\beta + \gamma(\delta - \mu)^2}{2\sigma^2}\right], \\
&= \underbrace{\sqrt{\frac{\gamma}{2\pi\sigma^2}} \exp\left[-\frac{\gamma(\delta - \mu)^2}{2\sigma^2}\right]}_{\text{Norm}} \underbrace{\frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{1}{\sigma^2}\right)^{\alpha+1} \exp\left[-\frac{\beta}{\sigma^2}\right]}_{\text{InvGam}}.
\end{aligned}$$

$$\begin{aligned}
& \int_{-\infty}^{\infty} d\mu \text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta] \\
&= \frac{1}{\sqrt{\pi}} \int_{-\infty}^{\infty} e^{-x^2} dx \left[\frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{1}{\sigma^2}\right)^{\alpha+1} e^{-\beta/\sigma^2} \right], \quad \text{where } x \equiv \sqrt{\frac{\gamma}{2\sigma^2}}(\mu - \delta) \\
&= \frac{1}{\Gamma(\alpha)} \left(\frac{\beta}{\sigma^2}\right)^{\alpha+1} e^{-\beta/\sigma^2} \frac{1}{\beta}, \\
&\equiv \text{InvGam}_{\sigma^2}[\alpha, \beta].
\end{aligned}$$

$$\begin{aligned}
& \int_0^\infty d\sigma^2 \int_{-\infty}^\infty d\mu \text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta] \\
&= \int_0^\infty \text{InvGam}_{\sigma^2}[\alpha, \beta] d\sigma^2, \\
&= \int \text{InvGam}_{\sigma^2}[\alpha, \beta] \left| \frac{d\sigma^2}{d\lambda} \right| d\lambda.
\end{aligned}$$

$\lambda = 1/\sigma^2$ (precision) に選ぶと、

$$\begin{aligned}
\text{InvGam}_{\sigma^2}[\alpha, \beta] \left| \frac{d\sigma^2}{d\lambda} \right| &= \frac{1}{\Gamma(\alpha)} (\beta\lambda)^{\alpha-1} e^{-\beta\lambda} \beta, \\
&\equiv \text{Gam}(\lambda|\alpha, \beta) \quad (\text{after Bishop}).
\end{aligned}$$

$$\begin{aligned}
\int_0^\infty d\sigma^2 \int_{-\infty}^\infty d\mu \text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta] &= \int_0^\infty \text{Gam}(\lambda|\alpha, \beta) d\lambda, \\
&= \frac{1}{\Gamma(\alpha)} \int_0^\infty X^{\alpha-1} e^{-X} dX, \quad (\text{where } X \equiv \beta\lambda) \\
&= \frac{1}{\Gamma(\alpha)} \Gamma(\alpha), \\
&= 1.
\end{aligned}$$

Summary:

$$\begin{aligned}
& \text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta] \\
&= \frac{\sqrt{\gamma}}{\sigma\sqrt{2\pi}} \frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{1}{\sigma^2}\right)^{\alpha+1} \exp\left[-\frac{2\beta + \gamma(\delta - \mu)^2}{2\sigma^2}\right], \\
&= \underbrace{\sqrt{\frac{\gamma}{2\pi\sigma^2}} \exp\left[-\frac{\gamma(\delta - \mu)^2}{2\sigma^2}\right]}_{\text{Norm}} \underbrace{\frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{1}{\sigma^2}\right)^{\alpha+1} \exp\left[-\frac{\beta}{\sigma^2}\right]}_{\text{InvGam}}.
\end{aligned}$$

Transformation: $(\mu, \sigma^2) \rightarrow (x, X),$

$$\begin{aligned}
\text{where } x &= \sqrt{\frac{\gamma}{2\sigma^2}}(\mu - \delta), \\
X &= \frac{\beta}{\sigma^2}.
\end{aligned}$$

Then
$$\left| \frac{\partial(x, X)}{\partial(\mu, \sigma^2)} \right| = \frac{\partial x}{\partial \mu} \left| \frac{dX}{d\sigma^2} \right| = \sqrt{\frac{\gamma}{2\sigma^2}} \frac{X^2}{\beta}.$$

$$\text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta] = \frac{1}{\sqrt{\pi}} e^{-x^2} \frac{1}{\Gamma(\alpha)} X^{\alpha-1} e^{-X} \left| \frac{\partial(x X)}{\partial(\mu \sigma^2)} \right|.$$

$$\begin{aligned} \int_0^\infty d\sigma^2 \int_{-\infty}^\infty d\mu \text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta] &= \int \int \frac{1}{\sqrt{\pi}} e^{-x^2} \frac{1}{\Gamma(\alpha)} X^{\alpha-1} e^{-X} \left| \frac{\partial(x X)}{\partial(\mu \sigma^2)} \right| d\mu d\sigma^2, \\ &= \int \int \frac{1}{\sqrt{\pi}} e^{-x^2} \frac{1}{\Gamma(\alpha)} X^{\alpha-1} e^{-X} dx dX, \\ &= \frac{1}{\sqrt{\pi}} \int_{-\infty}^\infty e^{-x^2} dx \frac{1}{\Gamma(\alpha)} \int_0^\infty X^{\alpha-1} e^{-X} dX, \\ &= \frac{1}{\sqrt{\pi}} \sqrt{\pi} \frac{1}{\Gamma(\alpha)} \Gamma(\alpha), \\ &= 1. \end{aligned}$$

Return to

Problem 3.8 Calculate an expression for the mode (position of the peak in μ, σ^2 space) of the normal scaled inverse gamma distribution in terms of the parameters $\alpha, \beta, \gamma, \delta$.

$$\text{NormInvGam}_{\mu, \sigma^2}[\alpha, \beta, \gamma, \delta] \propto \left(\frac{1}{\sigma^2}\right)^{\alpha+3/2} \exp\left[-\frac{2\beta + \gamma(\delta - \mu)^2}{2\sigma^2}\right].$$

$$\ln(\text{NormInvGam}) = -\left(\alpha + \frac{3}{2}\right) \ln \sigma^2 - \frac{2\beta + \gamma(\delta - \mu)^2}{2\sigma^2} + \text{constant.}$$

$$\begin{aligned} \frac{\partial \text{NormInvGam}}{\partial \mu} &= \frac{\gamma(\delta - \mu)}{\sigma^2} \text{NormInvGam}, \\ &= 0, \end{aligned}$$

$$\therefore \mu = \delta. \tag{a}$$

$$\frac{\partial \text{NormInvGam}}{\partial \sigma^2} = \left[-\left(\alpha + \frac{3}{2}\right) \frac{1}{\sigma^2} + \frac{2\beta + \gamma(\delta - \mu)^2}{2\sigma^4} \right] \text{NIG},$$

$$\begin{aligned}
&= \frac{1}{2\sigma^4} [2\beta + \gamma(\delta - \mu)^2 - (2\alpha + 3)\sigma^2] \text{ NIG}, \\
&= 0, \\
\therefore \sigma^2 &= \frac{2\beta + \gamma(\delta - \mu)^2}{2\alpha + 3}.
\end{aligned} \tag{b}$$

From Eqs. (a) and (b),

$$(\mu, \sigma^2) = \left(\delta, \frac{2\beta}{2\alpha + 3} \right).$$

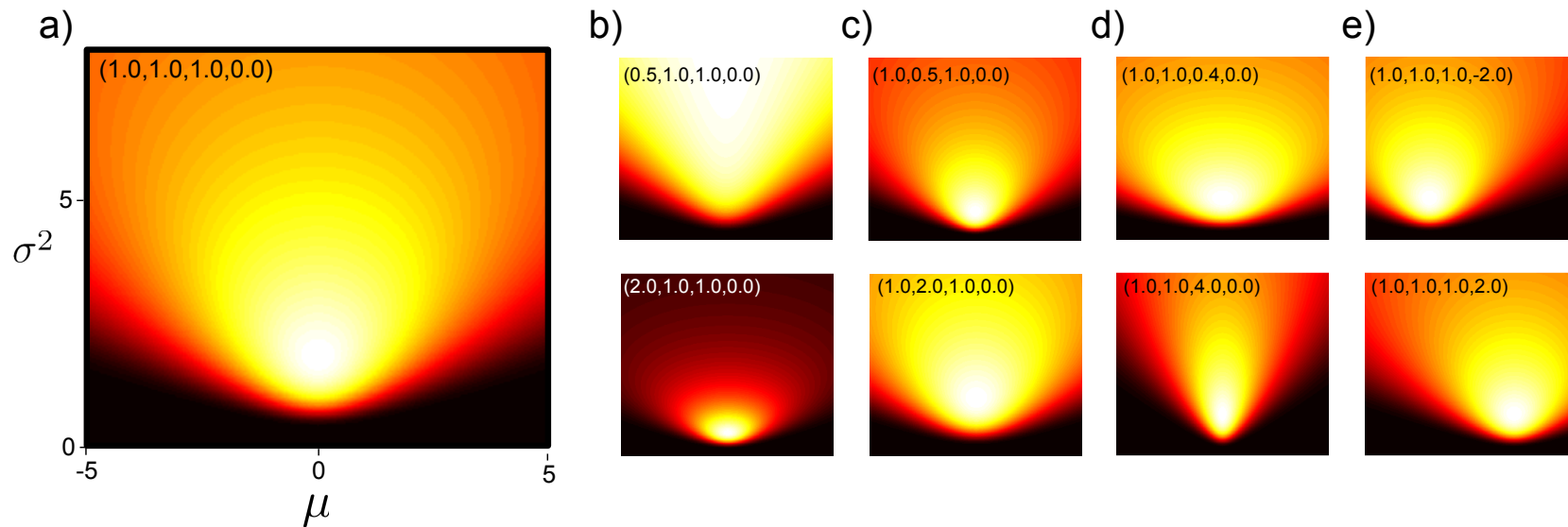


Figure 3.6 The normal-scaled inverse gamma distribution defines a probability distribution over bivariate continuous values μ, σ^2 where $\mu \in [-\infty, \infty]$ and $\sigma^2 \in [0, \infty]$. a) Distribution with parameters $[\alpha, \beta, \gamma, \delta] = [1, 1, 1, 0]$. b) Varying α . c) Varying β . d) Varying γ . e) Varying δ .