
Supplement to Bayesian Markov Blanket Estimation

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1 Model

Let $\mathbf{S} \sim \mathcal{W}_{p+q}(n, \mathbf{W})$ be Wishart distributed with n degrees of freedom and inverse covariance $\mathbf{W} = \mathbf{\Sigma}^{-1}$, and define the following partitioning:

$$\mathbf{S} = \begin{matrix} & p & q \\ p & \begin{pmatrix} \mathbf{S}_{11} & \mathbf{S}_{12} \\ \mathbf{S}_{21} & \mathbf{S}_{22} \end{pmatrix} \end{matrix}, \quad \mathbf{W} = \begin{matrix} & p & q \\ q & \begin{pmatrix} \mathbf{W}_{11} & \mathbf{W}_{12} \\ \mathbf{W}_{21} & \mathbf{W}_{22} \end{pmatrix} \end{matrix}. \quad (1)$$

Likelihood

$$\begin{aligned} p(\mathbf{S}|\mathbf{W}) &\propto \det(\mathbf{W})^{\frac{n}{2}} \exp \operatorname{tr} \left(-\frac{1}{2} \mathbf{W} \mathbf{S} \right) \\ &= \det(\mathbf{W}_{11})^{\frac{n}{2}} \det(\mathbf{W}_{22 \cdot 1})^{\frac{n}{2}} \\ &\quad \times \exp \operatorname{tr} \left(-\frac{1}{2} (\mathbf{W}_{11} \mathbf{S}_{11} + \mathbf{W}_{12} \mathbf{S}_{21} + \mathbf{W}_{21} \mathbf{S}_{12} + \mathbf{W}_{11}^{-1} \mathbf{W}_{12} \mathbf{S}_{22} \mathbf{W}_{21}) \right) \\ &\quad \times \exp \operatorname{tr} \left(-\frac{1}{2} \mathbf{W}_{22 \cdot 1} \mathbf{S}_{22} \right) \end{aligned} \quad (2)$$

Prior

$$\begin{aligned} p(\mathbf{W}) &\propto \exp \operatorname{tr} \left(-\frac{1}{2} \mathbf{I} \mathbf{W} \right) \prod_{w_{ij} \in \mathbf{W}_{12}} \frac{1}{\sqrt{2\pi t_{ij}}} \exp \left(-\frac{w_{ij}^2}{2t_{ij}} \right) \frac{\gamma^2}{2} \exp \left(-\frac{\gamma^2}{2} t_{ij} \right) \\ &= \exp \operatorname{tr} \left(-\frac{1}{2} \mathbf{I} \mathbf{W} \right) \prod_{i=1}^p \exp \left(-\frac{1}{2} \beta_i^\top \mathbf{D}_i \beta_i \right) \prod_{j=p+1}^{p+q} \frac{1}{\sqrt{2\pi t_{ij}}} \frac{\gamma^2}{2} \exp \left(-\frac{\gamma^2}{2} t_{ij} \right) \end{aligned} \quad (3)$$

Here, $\beta_i = (\mathbf{W}_{12})_i$ denotes the i th row of \mathbf{W}_{12} , $\mathbf{D}_i = \operatorname{diag}((T_i \cdot)^{-1})$, and $T_i = \operatorname{diag}(t_{i1}, \dots, t_{ip})$.

Posterior

$$\begin{aligned} p(\mathbf{W}_{11}, \mathbf{W}_{12}|\mathbf{S}) &\propto \det(\mathbf{W}_{11})^{\frac{n}{2}} \\ &\quad \times \exp \operatorname{tr} \left(-\frac{1}{2} (\mathbf{W}_{11} (\mathbf{S}_{11} + \mathbf{I}) + \mathbf{W}_{12} \mathbf{S}_{21} + \mathbf{W}_{21} \mathbf{S}_{12} + \mathbf{W}_{11}^{-1} \mathbf{W}_{12} (\mathbf{S}_{22} + \mathbf{I}) \mathbf{W}_{21}) \right) \\ &\quad \times \prod_{k=1}^p \exp \left(-\frac{1}{2} \beta_k^\top \mathbf{D}_k \beta_k \right) \prod_{j=p+1}^{p+q} \frac{1}{\sqrt{2\pi t_{ij}}} \frac{\gamma^2}{2} \exp \left(-\frac{\gamma^2}{2} t_{ij} \right) \end{aligned} \quad (4)$$

2 Wishart Distribution

If $\mathbf{S} \sim \mathcal{W}_{p+q}(n, \mathbf{\Sigma})$, $n > p + q - 1$, then

$$\begin{aligned} p(\mathbf{S}) &\propto \det(\mathbf{\Sigma})^{-\frac{n}{2}} \det(\mathbf{S})^{\frac{n-(p+q)-1}{2}} \exp \operatorname{tr} \left(-\frac{1}{2} \mathbf{\Sigma}^{-1} \mathbf{S} \right) \\ &= \det(\mathbf{W})^{\frac{n}{2}} \det(\mathbf{S})^{\frac{n-(p+q)-1}{2}} \exp \operatorname{tr} \left(-\frac{1}{2} \mathbf{W} \mathbf{S} \right) \end{aligned} \quad (5)$$

where $\mathbf{W} = \mathbf{\Sigma}^{-1}$.

We will prove Lemma 1 from the paper.

Lemma 1 The probability density function of a Wishart distribution factorises in \mathbf{W} as follows:

$$\mathcal{L}_{\mathbf{S}}(\mathbf{W}) \propto \mathcal{L}_1(\mathbf{W}_{11}, \mathbf{W}_{12}) \mathcal{L}_2(\mathbf{W}_{22 \cdot 1}) \quad (6)$$

Proof The proof is similar to [2]. Factorising the Wishart density according to

$$\begin{aligned} \mathbf{W} \mathbf{S} &= \mathbf{W}_{11} \mathbf{S}_{11} + \mathbf{W}_{12} \mathbf{S}_{21} + \mathbf{W}_{21} \mathbf{S}_{12} + \mathbf{W}_{22} \mathbf{S}_{22} \\ &= \mathbf{W}_{11} \mathbf{S}_{11} + \mathbf{W}_{12} \mathbf{S}_{21} + \mathbf{W}_{21} \mathbf{S}_{12} + (\mathbf{W}_{22} - \mathbf{W}_{21} \mathbf{W}_{11}^{-1} \mathbf{W}_{12} + \mathbf{W}_{21} \mathbf{W}_{11}^{-1} \mathbf{W}_{12}) \mathbf{S}_{22} \quad (7) \\ &= \mathbf{W}_{11} \mathbf{S}_{11} + \mathbf{W}_{12} \mathbf{S}_{21} + \mathbf{W}_{21} \mathbf{S}_{12} + \mathbf{W}_{21} \mathbf{W}_{11}^{-1} \mathbf{W}_{12} \mathbf{S}_{22} + \mathbf{W}_{22 \cdot 1} \mathbf{S}_{22} \end{aligned}$$

and

$$\det(\mathbf{W})^{\frac{n}{2}} = \det(\mathbf{W}_{11})^{\frac{n}{2}} \det(\mathbf{W}_{22 \cdot 1})^{\frac{n}{2}} \quad (8)$$

the factorisation follows immediately with

$$\begin{aligned} \mathcal{L}_1(\mathbf{W}_{11}, \mathbf{W}_{12}) &= \det(\mathbf{W}_{11})^{\frac{n}{2}} \exp \operatorname{tr} \left(-\frac{1}{2} (\mathbf{W}_{11} \mathbf{S}_{11 \cdot 2} + \mathbf{W}_{12} \mathbf{S}_{21} + \mathbf{W}_{21} \mathbf{S}_{12} + \mathbf{W}_{21} \mathbf{W}_{11}^{-1} \mathbf{W}_{12} \mathbf{S}_{22}) \right) \quad (9) \end{aligned}$$

and

$$\mathcal{L}_2(\mathbf{W}_{22 \cdot 1}) = \det(\mathbf{W}_{22 \cdot 1})^{\frac{n}{2}} \exp \operatorname{tr} \left(-\frac{1}{2} (\mathbf{W}_{22 \cdot 1} \mathbf{S}_{22}) \right). \quad (10)$$

3 Sampling from $vec(\mathbf{W}_{12}^T) | \mathbf{W}_{11}, \mathbf{S}, \{t_{ij}\}$

We will prove Theorem 2, part 1 from the paper.

Theorem 2, Part 1 Vectorised rows of \mathbf{W}_{12} follow a joint Normal distribution

$$vec(\mathbf{W}_{12}^T) | \mathbf{W}_{11}, \mathbf{S} \sim \mathcal{N}_{pq} \left(vec(-(\mathbf{S}_{22} + \mathbf{I})^{-T} \mathbf{S}_{12}^T \mathbf{W}_{11}^T), \mathbf{C}^{-1} \right) \quad (11)$$

where the matrix is $\mathbf{C} = \mathbf{W}_{11}^{-1} \otimes (\mathbf{S}_{22} + \mathbf{I}) + \text{diag}(\mathbf{D}_1, \dots, \mathbf{D}_p)$, and $\mathbf{D}_i = \text{diag}((T_{i\cdot})^{-1})$ is a diagonal matrix containing $T_{i\cdot} = (t_{i1}, \dots, t_{iq})$.

Proof The first step of the proof is similar to [2]. We factorise the Wishart likelihood according to

$$\begin{aligned} \mathbf{W}\mathbf{S} &= \mathbf{W}_{11}\mathbf{S}_{11} + \mathbf{W}_{12}\mathbf{S}_{21} + \mathbf{W}_{21}\mathbf{S}_{12} + \mathbf{W}_{22}\mathbf{S}_{22} \\ &= \mathbf{W}_{11}(\mathbf{S}_{11} - \mathbf{S}_{12}\mathbf{S}_{22}^{-1}\mathbf{S}_{21} + \mathbf{S}_{12}\mathbf{S}_{22}^{-1}\mathbf{S}_{21}) + \mathbf{W}_{12}\mathbf{S}_{21} + \mathbf{W}_{21}\mathbf{S}_{12} \\ &\quad + (\mathbf{W}_{22} - \mathbf{W}_{21}\mathbf{W}_{11}^{-1}\mathbf{W}_{12} + \mathbf{W}_{21}\mathbf{W}_{11}^{-1}\mathbf{W}_{12})\mathbf{S}_{22} \\ &= \mathbf{W}_{11}\mathbf{S}_{11.2} + \mathbf{W}_{11}\mathbf{S}_{12}\mathbf{S}_{22}^{-1}\mathbf{S}_{21} + \mathbf{W}_{12}\mathbf{S}_{21} + \mathbf{W}_{21}\mathbf{S}_{12} + \mathbf{W}_{21}\mathbf{W}_{11}^{-1}\mathbf{W}_{12}\mathbf{S}_{22} + \mathbf{W}_{22.1}\mathbf{S}_{22} \\ &= \underbrace{\mathbf{W}_{11}\mathbf{S}_{11.2}}_{\mathcal{W}} + \underbrace{\mathbf{S}_{22}(\mathbf{W}_{12} + \mathbf{W}_{11}\mathbf{S}_{12}\mathbf{S}_{22}^{-1})^T \mathbf{W}_{11}^{-1}(\mathbf{W}_{12} + \mathbf{W}_{11}\mathbf{S}_{12}\mathbf{S}_{22}^{-1})}_{\mathcal{MN}} + \underbrace{\mathbf{W}_{22.1}\mathbf{S}_{22}}_{\mathcal{W}} \end{aligned} \quad (12)$$

where we changed variables $\mathbf{W}_{22.1} = \mathbf{W}_{22} - \mathbf{W}_{21}\mathbf{W}_{11}^{-1}\mathbf{W}_{12}$ with Jacobian 1 and integrated over $\mathbf{W}_{22.1}$. Factorise also the determinants according to

$$\begin{aligned} \det(\mathbf{W})^{\frac{n}{2}} &= \det(\mathbf{W}_{11})^{\frac{n}{2}} \det(\mathbf{W}_{22.1})^{\frac{n}{2}} \\ &= \det(\mathbf{W}_{11})^{\frac{n+p}{2}} \underbrace{\det(\mathbf{W}_{11})^{-\frac{p}{2}}}_{\mathcal{MN}} \det(\mathbf{W}_{22.1})^{\frac{n}{2}} \end{aligned} \quad (13)$$

and

$$\begin{aligned} \det(\mathbf{S})^{\frac{n-(p+q)-1}{2}} &= \det(\mathbf{S}_{22})^{\frac{n-p-q-1}{2}} \det(\mathbf{S}_{11.2})^{\frac{n-p-q-1}{2}} \\ &= \det(\mathbf{S}_{22})^{\frac{n-p-2q-1}{2}} \underbrace{\det(\mathbf{S}_{22}^{-1})^{-\frac{q}{2}}}_{\mathcal{MN}} \det(\mathbf{S}_{11.2})^{\frac{n-p-q-1}{2}}. \end{aligned} \quad (14)$$

To construct the posterior, we include the Wishart part of the prior. This results in

$$\mathbf{W}_{12} | \mathbf{W}_{11}, \mathbf{S} \sim \mathcal{MN}_{p \times q} \left(-\mathbf{W}_{11}\mathbf{S}_{12}(\mathbf{S}_{22} + \mathbf{I})^{-1}, \mathbf{W}_{11}, (\mathbf{S}_{22} + \mathbf{I})^{-1} \right) \quad (15)$$

which is equivalent to

$$vec(\mathbf{W}_{12}^T) | \mathbf{W}_{11}, \mathbf{S} \sim \mathcal{N}_{pq} \left(vec(-(\mathbf{S}_{22} + \mathbf{I})^{-T} \mathbf{S}_{12}^T \mathbf{W}_{11}^T), \mathbf{W}_{11}^{-1} \otimes (\mathbf{S}_{22} + \mathbf{I}) \right) \quad (16)$$

where $vec(\mathbf{W}_{12}^T)$ are the vectorised rows of matrix \mathbf{W}_{12} and \otimes denotes the Kronecker product. For inclusion of the double exponential prior, it has to be rewritten as

$$\begin{aligned} \prod_{w_{ij} \in \mathbf{W}_{12}} \frac{1}{\sqrt{2\pi t_{ij}}} \exp \left(-\frac{w_{ij}^2}{2t_{ij}} \right) &= \prod_{i=1}^p \exp \left(-\frac{1}{2} (\mathbf{W}_{12})_{i\cdot}^T \mathbf{D}_i (\mathbf{W}_{12})_{i\cdot} \right) \\ &= \exp \left(-\frac{1}{2} vec(\mathbf{W}_{12}^T)^T \mathbf{D} vec(\mathbf{W}_{12}^T) \right) \end{aligned} \quad (17)$$

where $(\mathbf{W}_{12})_{i\cdot}$ denotes the i th row of \mathbf{W}_{12} , $\mathbf{D}_i = \text{diag}((T_{i\cdot})^{-1})$, and $\mathbf{D} = \text{diag}(\mathbf{D}_1, \dots, \mathbf{D}_p)$. The result follows from multiplying the prior by the expanded density in Eq. (16).

4 Matrix Generalised Inverse Gaussian (\mathcal{MGIG}) Distribution

$\mathbf{X} \in \mathbb{R}^{p \times p}$ follows a Matrix Generalised Inverse Gaussian (MGIG) distribution, if $\lambda > \frac{p-1}{2}$ and

$$\mathbf{X} \sim \mathcal{MGIG}_{p \times p}(\lambda, \mathbf{A}, \mathbf{B}) \propto \det(\mathbf{X})^{-\lambda-1} \exp \operatorname{tr} \left(-\frac{1}{2}(\mathbf{A}\mathbf{X} + \mathbf{B}\mathbf{X}^{-1}) \right). \quad (18)$$

Lemma 2 Let $p(\mathbf{X}) \propto \det(\mathbf{X})^{-\lambda-1} \exp \operatorname{tr} \left(-\frac{1}{2}(\mathbf{A}\mathbf{X} + \mathbf{B}\mathbf{X}^{-1}) \right)$ be MGIG distributed, then $\mathbf{W} = \mathbf{X}^{-1}$ is also MGIG distributed:

$$p(\mathbf{W}) \propto \det(\mathbf{W})^{\lambda-p} \exp \operatorname{tr} \left(-\frac{1}{2}(\mathbf{A}\mathbf{W}^{-1} + \mathbf{B}\mathbf{W}) \right) \quad (19)$$

Proof Transforming $\mathbf{W} = \mathbf{X}^{-1}$ with $d\mathbf{X} = -\mathbf{W}^{-1}d\mathbf{W}\mathbf{W}$

$$\begin{aligned} p(\mathbf{W}) &\propto \det(\mathbf{W})^{-(p+1)} \det(\mathbf{W}^{-1})^{-\lambda-1} \exp \operatorname{tr} \left(-\frac{1}{2}(\mathbf{A}\mathbf{W}^{-1} + \mathbf{B}\mathbf{W}) \right) \\ &= \det(\mathbf{W})^{\lambda-p} \exp \operatorname{tr} \left(-\frac{1}{2}(\mathbf{A}\mathbf{W}^{-1} + \mathbf{B}\mathbf{W}) \right). \end{aligned} \quad (20)$$

Theorem 2, Part 2 If $\mathbf{S} \sim \mathcal{W}_{p+q}(n, \mathbf{\Sigma})$ is Wishart distributed, then

$$\mathbf{W}_{11}|\mathbf{W}_{12}, \mathbf{S} \sim \mathcal{MGIG}_{p \times p} \left(\frac{n}{2} + p, \mathbf{W}_{12}(\mathbf{S}_{22} + \mathbf{I})\mathbf{W}_{21}, \mathbf{S}_{11} + \mathbf{I} \right) \quad (21)$$

is MGIG distributed.

Proof The proof is similar to [1]. Factorising the Wishart density according to

$$\begin{aligned} \exp \operatorname{tr} \left(-\frac{1}{2}\mathbf{W}\mathbf{S} \right) &= \exp \operatorname{tr} \left(-\frac{1}{2}(\mathbf{W}_{11}\mathbf{S}_{11} + \mathbf{W}_{12}\mathbf{S}_{21} + \mathbf{W}_{21}\mathbf{S}_{12} + \mathbf{W}_{22}\mathbf{S}_{22}) \right) \\ &= \exp \operatorname{tr} \left(-\frac{1}{2}(\mathbf{W}_{11}\mathbf{S}_{11} + \mathbf{W}_{12}\mathbf{S}_{21} + \mathbf{W}_{21}\mathbf{S}_{12} + (\mathbf{W}_{22} - \mathbf{W}_{21}\mathbf{W}_{11}^{-1}\mathbf{W}_{12} + \mathbf{W}_{21}\mathbf{W}_{11}^{-1}\mathbf{W}_{12})\mathbf{S}_{22}) \right) \\ &= \exp \operatorname{tr} \left(-\frac{1}{2}(\mathbf{W}_{11}\mathbf{S}_{11} + \mathbf{W}_{12}\mathbf{S}_{21} + \mathbf{W}_{21}\mathbf{S}_{12} + (\mathbf{W}_{22 \cdot 1} + \mathbf{W}_{21}\mathbf{W}_{11}^{-1}\mathbf{W}_{12})\mathbf{S}_{22}) \right) \\ &= \exp \operatorname{tr} \left(-\frac{1}{2} \left(\mathbf{W}_{11} \underbrace{\mathbf{S}_{11}}_{=\mathbf{B}} + \mathbf{W}_{12}\mathbf{S}_{21} + \mathbf{W}_{21}\mathbf{S}_{12} + \underbrace{\mathbf{W}_{12}\mathbf{S}_{22}\mathbf{W}_{21}}_{=\mathbf{A}} \mathbf{W}_{11}^{-1} + \mathbf{W}_{22 \cdot 1}\mathbf{S}_{22} \right) \right) \end{aligned} \quad (22)$$

and

$$\det(\mathbf{W})^{\frac{n}{2}} = \det(\mathbf{W}_{11})^{\frac{n}{2}} \det(\mathbf{W}_{22 \cdot 1})^{\frac{n}{2}} \quad (23)$$

where we changed variables $\mathbf{W}_{22 \cdot 1} = \mathbf{W}_{22} - \mathbf{W}_{21}\mathbf{W}_{11}^{-1}\mathbf{W}_{12}$ with Jacobian 1 and integrated over $\mathbf{W}_{22 \cdot 1}$. Comparing Eq. (19) with Eq. (22), we can identify \mathbf{A} , \mathbf{B} , and λ :

$$\frac{n}{2} \stackrel{!}{=} \lambda - p \quad \Rightarrow \quad \lambda = \frac{n}{2} + p. \quad (24)$$

such that

$$\mathbf{W}_{11}|\mathbf{W}_{12}, \mathbf{S} \sim \mathcal{MGIG}_{p \times p} \left(\frac{n}{2} + p, \mathbf{W}_{12}\mathbf{S}_{22}\mathbf{W}_{21}, \mathbf{S}_{11} \right) \quad (25)$$

Since the double exponential prior does not affect the distribution of \mathbf{W}_{11} , we only have to include the Wishart prior. The result follows immediately.

References

- [1] Ronald W Butler. Generalized inverse gaussian distributions and their wishart connections. *Scandinavian journal of statistics*, 25(1):69–75, 1998.
- [2] Arjun K Gupta and Daya K Nagar. *Matrix variate distributions*, volume 104. CRC Press, 1999.