

A Constancy of Curvature of Statistical Manifolds with α -Connection

Min Chol Rim, Ro Kwang Won

In this paper, we study a constancy of curvature of a statistical manifold, which is importantly studied in the information geometry recently.

In [2], an α -connection was introduced into a statistical manifold and a condition for a statistical manifold to be α -flat was studied, in [3], a condition for a statistical manifold with an α -connection to be conjugate symmetric, and in [1], a condition for a statistical manifold with an α -connection to be conjugate Ricci-symmetric

It has been found that an equiaffine structure of a statistical manifold is concerned with a constancy of curvature of a statistical manifold and there has been consideration of properties of a Hessian manifold of a constant Hessian curvature, which is a special statistical manifold (see [3, 4]).

The studies in [4] imply a necessity for considering a constancy of curvature of a statistical manifold with an α -connection. Such a study is also a continuation of the previous study [1–3].

Hence, we study conditions that a statistical manifold with an α -connection is of constant curvature.

For any $\alpha \in \mathbf{R}$, a α -connection is defined in a statistical manifold (M, g, ∇) by

$$\nabla^{(\alpha)} = \frac{1+\alpha}{2}\nabla + \frac{1-\alpha}{2}\nabla^*,$$

where ∇ and ∇^* are dual connections on M . Then $(M, g, \nabla^{(\alpha)})$ is also a statistical manifold.

Let's considerate the condition that a statistical manifold $(M, g, \nabla^{(\alpha)})$ is of constant curvature for any $\alpha \in \mathbf{R}$.

Theorem 1 A statistical manifold $(M, g, \nabla^{(\alpha)})$ is of constant curvature for any $\alpha \in \mathbf{R}$ if there exist $\alpha_1, \alpha_2 \in \mathbf{R}$ ($|\alpha_1| \neq |\alpha_2|$) such that statistical manifolds $(M, g, \nabla^{(\alpha_1)})$ and $(M, g, \nabla^{(\alpha_2)})$ are of constant curvature.

Proof Without generality, we assume $\alpha_1 \neq 0$. Then since $\nabla^{(\alpha)} = \frac{\alpha_1 + \alpha}{2\alpha_1}\nabla^{(\alpha_1)} + \frac{\alpha_1 - \alpha}{2\alpha_1}\nabla^{(-\alpha_1)}$ holds for any $\alpha \in \mathbf{R}$, the following relation

$$\begin{aligned} R^{(\alpha)}(X, Y)Z &= \frac{\alpha_1 + \alpha}{2\alpha_1}R^{(\alpha_1)}(X, Y)Z + \frac{\alpha_1 - \alpha}{2\alpha_1}R^{(-\alpha_1)}(X, Y)Z + \\ &\quad + (\alpha_1^2 - \alpha^2)[K(Y, K(Z, X)) - K(X, K(Y, Z))] \end{aligned}$$

holds, where $K(X, Y) = \nabla_X Y - \nabla_X^g Y$ is the difference tensor field of a statistical manifold.

From the relations

$$R^{(\alpha_1)}(X, Y)Z = k_1\{g(Y, Z)X - g(X, Z)Y\}, \quad (1)$$

$$R^{(\alpha_2)}(X, Y)Z = k_2\{g(Y, Z)X - g(X, Z)Y\} \quad (2)$$

the relation

$$R^{(\alpha)}(X, Y)Z = \frac{k_2\alpha_1^2 - k_1\alpha_2^2 + (k_1 - k_2)\alpha^2}{\alpha_1^2 - \alpha_2^2} \{g(Y, Z)X - g(X, Z)Y\}$$

holds, that is, a statistical manifold $(M, \nabla^{(\alpha)}, g)$ is of constant curvature $\frac{k_2\alpha_1^2 - k_1\alpha_2^2 + (k_1 - k_2)\alpha^2}{\alpha_1^2 - \alpha_2^2}$.

Example 1 Let (M, g) be a family of normal distributions:

$$M = \left\{ p(x, \theta) \mid p(x, \theta) = \frac{1}{\sqrt{2\pi(\theta^2)^2}} \exp\left\{-\frac{1}{2(\theta^2)^2}(x - \theta^1)^2\right\} \right\},$$

$$g := 2(\theta^2)^{-2} \sum d\theta^i d\theta^i \quad x \in P, \quad \theta^1 \in P, \quad \theta^2 > 0$$

We define an α -connection by the following:

$$\nabla^{(\alpha)}_{\frac{\partial}{\partial \theta^1}} \frac{\partial}{\partial \theta^1} = (-1 + 2\alpha)(\theta^2)^{-1} \frac{\partial}{\partial \theta^2}, \quad \nabla^{(\alpha)}_{\frac{\partial}{\partial \theta^2}} \frac{\partial}{\partial \theta^2} = (1 + \alpha)(\theta^2)^{-1} \frac{\partial}{\partial \theta^2}$$

$$\nabla^{(\alpha)}_{\frac{\partial}{\partial \theta^1}} \frac{\partial}{\partial \theta^2} = \nabla^{(\alpha)}_{\frac{\partial}{\partial \theta^2}} \frac{\partial}{\partial \theta^1} = 0.$$

Then the statistical manifold $(M, g, \nabla^{(0)})$ is of constant curvature $\left(-\frac{1}{2}\right)$, and the statistical manifold $(M, g, \nabla^{(1)})$ is of constant curvature 0. Hence for any $\alpha \in \mathbf{R}$, the statistical manifold $(M, g, \nabla^{(\alpha)})$ is of constant curvature $((\alpha^2 - 1)/2) -$.

Example 2 Let (M, g) be a family of random walk distributions [1]:

$$M = \left\{ p(x; \theta^1, \theta^2) \mid p(x; \theta^1, \theta^2) = \sqrt{\frac{\theta^2}{2\pi x}} \exp\left\{-\frac{\theta^2 x}{2} + \frac{\theta^2}{\theta^1} - \frac{\theta^2}{2(\theta^1)^2 x}\right\}, \quad x, \mu, \lambda > 0 \right\},$$

$$g = \frac{\theta^2}{(\theta^1)^3} (d\theta^1)^2 + \frac{1}{2(\theta^2)^2} (d\theta^2)^2.$$

We define an α -connection by the following:

$$\nabla^{(\alpha)}_{\frac{\partial}{\partial \theta^1}} \frac{\partial}{\partial \theta^1} = \frac{-3(1 + \alpha)}{2} (\theta^1)^{-1} \frac{\partial}{\partial \theta^1} + (-1 + \alpha)(\theta^1)^{-3} (\theta^2)^2 \frac{\partial}{\partial \theta^2}$$

$$\nabla^{(\alpha)}_{\frac{\partial}{\partial \theta^1}} \frac{\partial}{\partial \theta^2} = \nabla^{(\alpha)}_{\frac{\partial}{\partial \theta^2}} \frac{\partial}{\partial \theta^1} = -\frac{1 + \alpha}{2} (\theta^2)^{-1} \frac{\partial}{\partial \theta^1}, \quad \nabla^{(\alpha)}_{\frac{\partial}{\partial \theta^2}} \frac{\partial}{\partial \theta^2} = (-1 + \alpha)(\theta^2)^{-1} \frac{\partial}{\partial \theta^2}.$$

Then the statistical manifold $(M, g, \nabla^{(0)})$ is of constant curvature $\left(-\frac{1}{2}\right)$, and the statistical manifold $(M, g, \nabla^{(1)})$ constant curvature 0. Hence for any $\alpha \in \mathbf{R}$, the statistical manifold $(M, g, \nabla^{(\alpha)})$ is of constant curvature $\frac{\alpha^2 - 1}{2}$.

Theorem 1 implies the following facts.

Corollary 1 If there exist $\alpha_1, \alpha_2 \in \mathbf{R}$ ($|\alpha_1| \neq |\alpha_2|$) such that the statistical manifold $(M, g, \nabla^{(\alpha_1)})$ is of constant curvature k_1 and the statistical manifold $(M, g, \nabla^{(\alpha_2)})$ is of constant curvature k_2 , and $k_1 = k_2 = k$ holds, then for $\alpha \in \mathbf{R}$, the statistical manifold $(M, g, \nabla^{(\alpha)})$ is of constant curvature k .

Corollary 2 If there exist $\alpha_1, \alpha_2 \in \mathbf{R}$ ($|\alpha_1| \neq |\alpha_2|$) such that the statistical manifold $(M, g, \nabla^{(\alpha_1)})$ is of constant curvature k_1 and the statistical manifold $(M, g, \nabla^{(\alpha_2)})$ constant curvature k_2 , and $k_1 \neq k_2$ holds, then for $\alpha \in \mathbf{R}$ satisfying that $\alpha^2 = \frac{k_2\alpha_1^2 - k_1\alpha_2^2}{k_2 - k_1}$, the statistical manifold $(M, g, \nabla^{(\alpha)})$ is flat.

Example 3 $k_1 = -1/2$, $k_2 = 0$, $\alpha_1 = 0$ and $\alpha_2 = 1$ hold in example 1 and example 2. Hence for $\alpha \in \mathbf{R}$ satisfying that $\alpha^2 = 1$, the statistical manifold $(M, g, \nabla^{(\alpha)})$ is flat.

Theorem 2 If the Hessian manifold (M, g, ∇) is of constant Hessian curvature, the statistical manifold $(M, g, \nabla^{(\alpha)})$ is of constant curvature for any $\alpha \in \mathbf{R}$.

Proof If the Hessian manifold (M, g, ∇) is of constant Hessian curvature,

$$(\nabla K)(Y, Z; X) = -\frac{c}{2}\{g(X, Y)Z + g(X, Z)Y\}, \quad c \in \mathbf{R}$$

holds. On the other hand, the curvature tensor R^{∇^g} of Levi-Civita connection ∇^g is written by

$$\begin{aligned} R^{\nabla^g}(X, Y)Z &= R^\nabla(X, Y)Z - (\nabla K)(Y, Z; X) + (\nabla K)(Z, X; Y) + \\ &\quad + K(X, K(Y, Z)) - K(Y, K(Z, X)) \end{aligned}$$

where R is the curvature tensor of ∇ and K is difference tensor $K(X, Y) := \nabla_X Y - \nabla_Y X$. Then

$$\begin{aligned} (\nabla K)(Y, Z; X) - (\nabla K)(Z, X; Y) &= 2\{K(X, K(Y, Z)) - K(Y, K(X, Z))\} + \\ &\quad + \frac{1}{2}\{R^\nabla(X, Y)Z - R^\nabla(Y, X)Z\} \end{aligned}$$

implies $R^{\nabla^g}(X, Y)Z = -\frac{c}{4}\{g(Y, Z)X - g(X, Z)Y\}$, where R^* is curvature tensor of dual connection ∇^* , that is, the statistical manifold (M, ∇^g, g) is of constant curvature. On the other hand, the statistical manifold (M, ∇, g) is flat, that is, constant curvature 0. Therefore we finish the proof of theorem from theorem 1.

3. Conclusion

We found a condition that the statistical manifold $(M, \nabla^{(\alpha)}, g)$ is of constant curvature for any $\alpha \in \mathbf{R}$. We also showed that if the Hessian manifold (M, ∇, g) is of constant Hessian curvature, the statistical manifold $(M, \nabla^{(\alpha)}, g)$ is of constant curvature for any $\alpha \in \mathbf{R}$,

References

- [1] 민철립 등; 조선민주주의인민공화국 과학원통보, 2, 11, 주체98(2009).
- [2] J. Zhang; AISM, 59, 161, 2007.
- [3] H. Matsuzoe et al.; Diff. Geom. Appl., 24, 567, 2006.
- [4] H. Furuhashi; Diff. Geom. Appl., 27, 420, 2009.