## Chapter 9. Models

Fitting models to data is a popular activity. For data taking values in a group or homogeneous space, the associated representation theory gives neat families of models. Briefly, if P(x) is a probability on X, write  $P(x) = e^{h(x)}$  with  $h = \log P$ . Then expand h in a natural basis  $b_i$  of L(X):  $h(x) = \Sigma \theta_i b_i(x)$ . Any positive probability can be expanded in this way. Truncating the expansion to a fixed number of terms leads to families of simple measures or models; because  $b_i$  are orthogonal,  $\theta_i$  are identifiable.

It is an interesting fact that many models introduced by applied workers fall into this class. The general story is presented first, then a specialization to data on spheres, then a specialization to partially ranked data. A brief review of other approaches to ranked data is followed by a section supplying the relevant exponential family theory.

## A. EXPONENTIAL FAMILIES FROM REPRESENTATIONS

Let G be a group acting transitively on a compact set X. Let L(X) denote the real valued continuous functions on X. Suppose X has an invariant distribution dx. The following abstracts an idea introduced by Lo (1977) and Beran (1979). *Definition*. Let  $\Theta$  be an invariant subspace of L(X) containing the constants. Define a family of measures, one for each  $\theta \in \Theta$ , by specifying the densities to be

$$P_{\theta}(dx) = a(\theta)e^{\theta(x)}dx,$$

where  $a(\theta)$  is a normalizing constant forcing  $P_{\theta}(dx)$  to integrate to 1.

Suppose  $\Theta$  is finite dimensional. Let  $b_0 = \text{constant}, b_1, b_2, \ldots, b_p$  be a basis for  $\Theta$ . Then the family can be parameterized as

$$(*) P_{\theta}(dx) = a(\theta)e^{\theta' b}dx, \ \theta \in \mathbb{R}^p, b = (b_1(x), \dots, b_p(x)).$$

LEMMA 1. The family \* is well parameterized in the sense that  $P_{\theta} = P_{\theta'}$  if and only if  $\theta = \theta'$ .

*Proof.* Only the forward direction requires proof. If  $P_{\theta} = P_{\theta'}$ , then

$$(\theta - \theta') \cdot b(x) = \log(a(\theta')/a(\theta))$$
 for all x.

The left side is a linear combination of  $b_1, b_2, \ldots, b_p$  which is constant. But  $1, b_1, b_2, \ldots, b_p$  is a basis, so  $\theta = \theta'$ .

In applications there is a decomposition into invariant subspaces  $L(X) = V_0 \oplus V_1 \oplus V_2 \oplus \ldots$  and  $\Theta$ 's are chosen as a finite direct sum of subspaces. Usually