# Clustering of Distributions on 1-Dimensional Manifolds

### Setup

Suppose we have two manifolds  $\mathcal{M}_1$  and  $\mathcal{M}_2 \in \mathbb{R}^d$ , each of length 1, defined by  $f_1(t)$  and  $f_2(t)$  respectively  $(f_i : [0,1] \mapsto \mathbb{R}^d)$ . Define  $\delta$  as the minimum distance between the two manifolds, i.e.,  $\delta = \min_t \|f_1(t) - f_2(t)\|$ , and let  $\delta > 0$ . For now, restrict each  $f_i$  such that the distance along the manifold between  $f_i(t)$  and  $f_i(s)$  is equal to the difference between t and s (this also implies that each manifold is of length 1). We sample  $T_1, ..., T_n \stackrel{iid}{\sim} F$  for continuous F with support [0, 1] and use  $f_1$  to map the first  $n_1$  points to  $\mathcal{M}_1$  and  $f_2$  to map the remaining  $n_2 = n - n_1$  points to  $\mathcal{M}_2$ . Let  $X_i = f_1(T_i)$  and  $Y_j = f_2(T_j)$ . Without loss of generality, assume  $n_1 \leq n_2$ .

# Theory

Let  $D_i = X_{(i)} - X_{(i-1)}$ . Then if  $\max_i D_i < \delta$ , we have sufficient separation of points in  $\mathcal{M}_1$ .

#### Uniform case

It can be shown that if  $X_i \stackrel{iid}{\sim} Uniform(0,1)$ , then  $D_i \sim Beta(1,n)$ . Therefore,  $P(\max_i D_i < \delta) \ge (P(D_i < \delta))^n = (1 - (1 - \delta)^n)^n$ , which is a decreasing function for sufficiently large n. This gives us the result  $\max_i D_i \stackrel{a.s.}{\to} 0$ .

# Beta case

#### General case

If F is absolutely continuous, then it can be shown that

$$P(D_i \le \delta) = 1 - \int_0^{1-\delta} (i-1) \binom{n}{i-1} f(x) (F(x))^{i-1} (1 - F(x+\delta))^{n-i} dx$$

Under certain conditions (namely  $F(x + \delta) \ge F(x) + \delta$ ), we can make the substitution u = F(x) and have the result

$$P(D_i \le \delta) \le 1 - \int_0^{1-\delta} (i-1) \binom{n}{i-1} u^{i-1} (1-u-\delta)^{n-i} du$$

which gives the same result as in the uniform case.

#### Computational Results

TBD