

Differential PP

Thursday, October 20, 2022 3:36 PM

Purpose: compare two groups of experiments from flow cytometry, compare the groups by finding the regions where they are the most different.

Data: X is $[n_1 \times p]$ -dim results for treatment, Y is $[n_2 \times p]$ -dim results for control group. Both have millions of rows for data.

Method 1:

1. Combine X and Y -> a combined large dataset, calculate the population proportion of observations, $\pi = n_1/(n_1+n_2)$
2. Construct m data nuggets for the combined dataset
3. For each nugget, there are observations from two groups. Calculate the proportion of observations in each nugget, $p_i = n_{1i}/(n_{1i}+n_{2i})$ for $i=1,2,\dots,m$.
4. For each nugget, do the test for $H_0: p_i = \pi$, $H_1: p_i \neq \pi$, then we have p values for each nugget.
5. Make a plot: x-axis is weights of nuggets/sqrt(weights)/log(weights), y-axis is $-\log(p\text{-values})$, check the p-values' change as weights change -> **purpose? Check:** $-\log(p\text{-values})$ is measurement of difference. What happens when if weights are big, two groups are very big, it could easily be significantly different. Eg, $\pi = 44\%$, $p_i = 43\%$. If the number of observations is very big, it would be significantly different. Decide **a linear boundary to decide** whether to refuse the test.
6. By p-values, we divide nuggets into two groups: one with no difference, one with significant difference. By compare values, we divide the second group into two. Finally we have three groups: one with no difference, one with p_i bigger than π , one with p_i smaller than π .
7. Reassignment: for those nuggets with $|p_i - \pi| < 0.2$, we put those into the no difference group. -> **how to decide 0.2, maybe according to dataset? Consider 3 or 2 percentages. According to step 5. Some formula?**
8. Consider Another group: different between p_i and $1-p_i$ -> $2p_i-1$ or $1-2p_i$. If $1-2p_i > 0.8$, it means in that nugget, most of observations are from one group. The nugget is almost pure.
9. Find the regions with difference, we may focus on the group with p_i bigger than π (treatment works better/activates more B-cells), conduct:
 - (1). Weighted PCA -> loadings for each protein
 - (2). PPFor this sub-group?
And another way: conduct weighted-Kmeans to get clusters for the combined data -> cluster information for all data, and then compare the group information and cluster information, eg, in each cluster, the proportions of each group -> find clusters with higher proportions -> target areas
(consider Classification and regression tree: CART for data nuggets)

#imbalance problem between treatment and control?

#multiple treatments compare for this method?

Method 2: By differential PP

1. **For one treatment and one control: Find a d-dim projection that has a very large difference between two distributions**

Step 1: For each projection, (apply same projection matrix on both data), $f_1(y)$: projected data density for treatment group, $f_2(y)$: projected data density for control group, f : combined density, **how to combine**, simple way: $f(y) = 1/2 \cdot (f_1(y) + f_2(y))$

Step 2: Calculate differential PP index:

$$I = \int_{\mathbb{R}^d} (f_1(y) - f_2(y))^2 f(y) dy = c \cdot \left[\int_{\mathbb{R}^d} (f_1(y) - f_2(y))^2 f(y) dy + \int_{\mathbb{R}^d} (f_2(y) - f_1(y))^2 f(y) dy \right]$$
$$= \int_{\mathbb{R}^d} (f_1(y)^2 - 2f_1(y)f_2(y) + f_2(y)^2) f(y) dy = \int_{\mathbb{R}^d} (f_1(y)^2 + f_2(y)^2) f(y) dy - 2 \int_{\mathbb{R}^d} (f_1(y) + f_2(y)) f(y)^2 dy + 2 \int_{\mathbb{R}^d} f(y)^3 dy$$

Step 3: maximize differential PP index to find optimal projection

Step 4: Conduct optimal projection on combined data, check two groups distribution and any area.

Step 5: apply varimax rotation to get protein information (loadings)/ apply clustering methods to projected data to check area and protein expressions in each clusters, other way?

2. **For multiple treatments and one control: in total k groups**

Similar things but different index: $f_i(y)$: projected data density for each group, $i=1,2,\dots,k$, f : combined density, **how to combine**, simple way: $f(y) = 1/k \cdot (f_1(y) + \dots + f_k(y))$

$$\int_{\mathbb{R}^d} (f_1 - f_2)^2 f dy + \int_{\mathbb{R}^d} (f_1 - f_3)^2 f dy + \int_{\mathbb{R}^d} (f_2 - f_3)^2 f dy = c \cdot \left[\int_{\mathbb{R}^d} (f_1 - f)^2 f dy + \int_{\mathbb{R}^d} (f_2 - f)^2 f dy + \int_{\mathbb{R}^d} (f_3 - f)^2 f dy \right]$$

$$\int_{\mathbb{R}^d} (f_1 - f_2)^2 f \, dy + \int_{\mathbb{R}^d} (f_1 - f_3)^2 f \, dy + \int_{\mathbb{R}^d} (f_2 - f_3)^2 f \, dy = c \cdot \left[\int_{\mathbb{R}^d} (f_1 - f) f \, dy + \int_{\mathbb{R}^d} (f_2 - f) f \, dy + \int_{\mathbb{R}^d} (f_3 - f) f \, dy \right]$$

2. For k groups.

$$\sum_{i,j=1,\dots,k} \int_{\mathbb{R}^d} |f_i(y) - f_j(y)|^2 f(y) \, dy = c \cdot \sum_{i=1}^k \int_{\mathbb{R}^d} |f_i(y) - f(y)|^2 f(y) \, dy$$

$\downarrow C_k^2 = \frac{k(k-1)}{2} \text{ integrals}$
 $\downarrow k \text{ integrals}$

Verification of equations and it could help: we do not need make pairwise comparisons between k groups ($C_k^2 = k(k-1)/2$ pairs), just compare each group with the combined density.

Task: find dataset to simulate and test (Davitt's data or other flow cytometry data on that website), and think about things with ?