

# ImmunoTherapy scoring function

The statistics I propose for  $s(\bullet)$  will be the proportion vector  $\{a, b, c, d, e\}$  where a is the proportion of progenitor cells, b, the proportion for effector cells, c, for terminal exhausted, d, for the cycling cells and e, for other cells.

So  $s(P_i)$  will be  $\{a_i, b_i, c_i, d_i, e_i\}$  where  $i$  stands for the index of the knocked out gene yielding the corresponding gene expression distribution  $P_i$ .

Let  $P_0$  correspond to the proportion vector  $\{a_0, b_0, c_0, d_0, e_0\}$  and Q be the proportion vector  $\{a_q, b_q, c_q, d_q, e_q\}$ . The proposed score function will be given by:

$$\frac{\exp(\sum_{i=0}^n \frac{1}{var_i})}{\exp((\sum_{i=1}^n \frac{L_i}{var_i}) - \frac{M}{var_0})} \quad \text{where}$$

$L_i$  is the  $L_1$  loss between  $P_i$  and Q given by

$$|a_i - a_q| + |b_i - b_q| + |c_i - c_q| + |d_i - d_q| + |e_i - e_q|$$

M is the  $L_1$  loss between  $P_0$  and Q,

$$|a_0 - a_q| + |b_0 - b_q| + |c_0 - c_q| + |d_0 - d_q| + |e_0 - e_q|$$

$n$  is the number of genes

$var_0$  is the variance of M and

$var_i$  where  $i = 1, \dots, n$  is the variance of  $L_i$

Dividing M and  $L_i$  by respective variance will ensure that quantities with more uncertainties can be given less weight in the calculation of the score.

The variance of  $L_i$  is the sum of the variances,  
 $var(a_i) + var(b_i) + var(c_i) + var(d_i) + var(e_i)$

Similarly the variance of M is calculated by taking the sum  
 $var(a_0) + var(b_0) + var(c_0) + var(d_0) + var(e_0)$  from the data representing  $P_0$

The resulting score is the pseudo-inverse of a weighted average of the knockout losses minus the loss of unperturbed distribution,  $P_0$  with Q where each loss is weighted by the inverse of the corresponding variance. Less loss in the knockout will yield greater the score and thus indicative of better perturbations. Taking the exponent will help in avoiding negative values in the denominator thus being supportive to the purpose of the scoring function. Finally greater M with less  $L_i$  will suggest that the model is performing well irrespective of the unperturbed distribution not already close to Q.