## **Supplementary Materials**

## **1.** Derivation of solution for common *W* matrix:

Suppose  $X^{i}$ , i=1, 2, ..., m be matrices representing *m* types of genomic assays profiled on same set of *n* samples with  $p_i$  features, *W* be the common basis matrix,  $H^{i}$ , i=1,...,m be the individual data specific coefficient matrices and  $\theta^{i} \ge 0$ , i=1,...,m be the weights for several data types. Then we can write

$$\boldsymbol{X}_{n \times p_i}^i \approx \boldsymbol{W}_{n \times k} \boldsymbol{H}_{k \times p_i}^i \quad i = 1, 2, \dots m$$

Where  $W_{n \times k} \ge 0$ ,  $H_{k \times p_i}^i \ge 0$ , i = 1, 2, ..., m

The intNMF problem minimizes the following objective function

$$Q = \sum_{i=1}^{m} \theta^{i} \left( \mathbf{X}_{n \times p_{i}}^{i} - \mathbf{W}_{n \times k} \mathbf{H}_{k \times p_{i}}^{i} \right)^{2}$$
$$Q = \sum_{i=1}^{m} \left[ \sqrt{\theta^{i}} \left( \mathbf{X}_{n \times p_{i}}^{i} - \mathbf{W}_{n \times k} \mathbf{H}_{k \times p_{i}}^{i} \right) \right]^{2}$$

There is no global minimum solution for NMF problem. Therefore, it is unrealistic to design an algorithm to find the global minimum of Q. We derive approximate solution for W. Since our objective is to approximately factorize  $X_{n \times p_i}^i$  in to  $W_{n \times k}$  and  $H_{k \times p_i}^i$  matrices for i=1,2,...,m, and that each of the *m* terms in the above equation is strictly positive we can write,

$$\sqrt{\theta^{i}} \left( \boldsymbol{X}_{n \times p_{i}}^{i} - \boldsymbol{W}_{n \times k} \boldsymbol{H}_{k \times p_{i}}^{i} \right) \approx \boldsymbol{0}_{n \times p_{i}}$$
$$\sqrt{\theta^{i}} \boldsymbol{X}_{n \times p_{i}}^{i} \approx \sqrt{\theta^{i}} \boldsymbol{W}_{n \times k} \boldsymbol{H}_{k \times p_{i}}^{i}$$

For simplicity, writing the expression in the form of equation

$$\sqrt{\theta^{i}} \boldsymbol{X}_{n \times p_{i}}^{i} = \sqrt{\theta^{i}} \boldsymbol{W}_{n \times k} \boldsymbol{H}_{k \times p_{i}}^{i} + \boldsymbol{C}_{n \times p_{i}}^{i}$$

Where  $C_{n \times p_i}^i$  is matrix of constants having small numbers.

Multiplying both sides by transpose of  $H_{k \times p_i}^i$  and summing over *i* 

$$\sum_{i=1}^{m} \sqrt{\theta^{i}} \mathbf{X}_{n \times p_{i}}^{i} (\mathbf{H}^{i})_{p_{i \times k}}^{T} = \sum_{i=1}^{m} \sqrt{\theta^{i}} \mathbf{W}_{n \times k} \mathbf{H}_{k \times p_{i}}^{i} (\mathbf{H}^{i})_{p_{i \times k}}^{T} + \sum_{i=1}^{m} \mathbf{C}_{n \times p_{i}}^{i} (\mathbf{H}^{i})_{p_{i \times k}}^{T}$$
$$\sum_{i=1}^{m} \sqrt{\theta^{i}} \mathbf{X}_{n \times p_{i}}^{i} (\mathbf{H}^{i})_{p_{i \times k}}^{T} = \sum_{i=1}^{m} \sqrt{\theta^{i}} \mathbf{W}_{n \times k} \mathbf{H}_{k \times p_{i}}^{i} (\mathbf{H}^{i})_{p_{i \times k}}^{T} + \mathbf{C}_{n \times k}^{\prime}$$

Where,  $C'_{n \times k} = \sum_{i=1}^{m} C^{i}_{n \times p_{i}} (H^{i})^{T}_{p_{i \times k}}$  is a constant. Note that  $H^{i}$ , i=1,...,m are computed separately.

$$\sum_{i=1}^{m} \sqrt{\theta^{i}} X_{n \times p_{i}}^{i} \left(H^{i}\right)_{p_{i \times k}}^{T} = W_{n \times k} \sum_{i=1}^{m} \sqrt{\theta^{i}} \left(H^{i} \left(H^{i}\right)^{T}\right)_{k \times k} + C_{n \times k}'$$
$$W_{n \times k} = \left(\sum_{i=1}^{m} \sqrt{\theta^{i}} X_{n \times p_{i}}^{i} \left(H^{i}\right)_{p_{i \times k}}^{T}\right) \left(\sum_{i=1}^{m} \sqrt{\theta^{i}} \left(H^{i} \left(H^{i}\right)^{T}\right)_{k \times k}\right)^{-1} - C_{n \times k}' \left(\sum_{i=1}^{m} \sqrt{\theta^{i}} \left(H^{i} \left(H^{i}\right)^{T}\right)_{k \times k}\right)^{-1}$$
$$W_{k \times n}^{T} = \left[\left(\sum_{i=1}^{m} \sqrt{\theta^{i}} X_{n \times p_{i}}^{i} \left(H^{i}\right)_{p_{i \times k}}^{T}\right) \left(\sum_{i=1}^{m} \sqrt{\theta^{i}} \left(H^{i} \left(H^{i}\right)^{T}\right)_{k \times k}\right)^{-1}\right]^{T} + C_{k \times n}'$$

Where  $C'_{k\times n} = \left[ -C'_{n\times k} \left( \sum_{i=1}^{m} \sqrt{\theta^{i}} (H^{i}(H^{i})^{T})_{k\times k} \right)^{-1} \right]^{T}$  constant matrix. Writing back in approximate form,

$$\boldsymbol{W}_{k\times n}^{T} \approx \left[ \left( \sum_{i=1}^{m} \sqrt{\theta^{i}} \left( \boldsymbol{H}^{i} \left( \boldsymbol{H}^{i} \right)^{T} \right)_{k\times k} \right)^{T} \right]^{-1} \left[ \left( \sum_{i=1}^{m} \sqrt{\theta^{i}} \boldsymbol{X}_{n\times p_{i}}^{i} \left( \boldsymbol{H}^{i} \right)_{p_{i\times k}}^{T} \right)^{T} \right]$$

This equation can be solved using non negative alternating least square method. Finally  $W_{n \times k}$  is obtained by the transpose of  $W_{k \times n}^T$ .

## 2. Solution for H matrices

 $H^{i}$ , i=1,...,m matrices are solved using non-negativity constraint least square method.

$$\boldsymbol{H}_{k \times p_{i}}^{i} \approx (\boldsymbol{W}^{T} \boldsymbol{W})_{k \times k}^{-1} \boldsymbol{W}_{k \times n}^{T} \boldsymbol{X}_{n \times p_{i}}^{i} \quad i = 1, 2, \dots, m$$
  
such that  $\boldsymbol{H}_{k \times p_{i}}^{i} \ge 0$ 

## 3. Convergence of intNMF to standard NMF when $m \rightarrow 1$

For, i=1,...,m the solution for W is given by

$$\boldsymbol{W}_{k\times n}^{T} \approx \left[ \left( \sum_{i=1}^{m} \sqrt{\theta^{i}} \left( \boldsymbol{H}^{i} \left( \boldsymbol{H}^{i} \right)^{T} \right)_{k\times k} \right)^{T} \right]^{-1} \left[ \left( \sum_{i=1}^{m} \sqrt{\theta^{i}} \boldsymbol{X}_{n\times p_{i}}^{i} \left( \boldsymbol{H}^{i} \right)_{p_{i\times k}}^{T} \right)^{T} \right]$$

If  $m \to 1$ ,

$$\boldsymbol{W}_{k\times n}^{T} \approx \left[ \left( \sum_{i=1}^{1} \sqrt{\theta^{i}} \left( \boldsymbol{H}^{i} \left( \boldsymbol{H}^{i} \right)^{T} \right)_{k\times k} \right)^{T} \right]^{-1} \left[ \left( \sum_{i=1}^{1} \sqrt{\theta^{i}} \boldsymbol{X}_{n\times p_{i}}^{i} \left( \boldsymbol{H}^{i} \right)_{p_{i\times k}}^{T} \right)^{T} \right]$$

There is no need of weight for a single data. i.e.  $\theta^i \rightarrow 1$ 

$$\boldsymbol{W}_{k\times n}^{T} \approx \left[ \left( \sum_{i=1}^{1} \sqrt{1} \left( \boldsymbol{H}^{i} \left( \boldsymbol{H}^{i} \right)^{T} \right)_{k\times k} \right)^{T} \right]^{-1} \left[ \left( \sum_{i=1}^{1} \sqrt{1} \boldsymbol{X}_{n\times p_{i}}^{i} \left( \boldsymbol{H}^{i} \right)_{p_{i\times k}}^{T} \right)^{T} \right]$$
$$= \left[ \left( \left( \boldsymbol{H}^{1} (\boldsymbol{H}^{1})^{T} \right)_{k\times k} \right)^{T} \right]^{-1} \left[ \left( \boldsymbol{X}_{n\times p_{1}}^{1} (\boldsymbol{H}^{1})_{p_{1\times k}}^{T} \right)^{T} \right]$$
$$= \left( \boldsymbol{H}^{1} (\boldsymbol{H}^{1})^{T} \right)_{k\times k}^{-1} \boldsymbol{H}_{k\times p_{1}}^{1} \left( \boldsymbol{X}^{1} \right)_{p_{1\times n}}^{T}$$

Which is solution of

$$(X^{1})_{p_{1}\times n}^{T} \approx (H^{1})_{p_{1}\times k}^{T} W_{k\times n}^{T}$$
$$\Rightarrow X_{n\times p_{1}}^{1} \approx W_{n\times k} H_{k\times p_{1}}^{1}$$

For simplicity, writing p for  $p_1$ , X for  $X^1$  and H for  $H^1$ .

$$\boldsymbol{X}_{n \times p} \approx \boldsymbol{W}_{n \times k} \boldsymbol{H}_{n \times p}$$

Which is standard NMF problem.

(Complete set of supplementary figures are provided separately with S2 File. Complete set of Supplementary Figures).