### **Supplementary Methods**

#### Details of the three introduced parameters in bSDTNBI

To improve the method performance of the original SDTNBI (Wu *et al.*, 2016), bSDTNBI introduced three tunable parameters,  $\alpha \in [0,1)$ ,  $\beta \in [0,1)$ , and

 $\gamma \in (-\infty, +\infty)$ , to its resource diffusion processes, described as below.

Firstly, in the initial resource allocation, for each drug, a total amount = 1 of initial resource was allocated to all its neighbor nodes. Specially, a total amount =  $\alpha$  of resource was equally allocated to its substructure nodes, and a total amount = 1- $\alpha$  of resource was equally allocated to its target nodes. In other words, for each drug,  $\alpha$  is the ratio of the amount of the resource located in its substructures to its total amount of resource (i.e. 1). Hence, by varying the value of parameter  $\alpha$  from 0 to 1, we can search which type of nodes (target or substructure nodes) will occupy the dominant position in the initial resource allocation (**Figure 1B**).

Secondly, in the substructure-drug-target network, the weighted values of all drug-substructure associations were set to  $\beta$ , and the weighted values of all DTIs were set to 1- $\beta$ . Hence, by varying the value of parameter  $\beta$  from 0 to 1, we can search which type of edges (DTIs or drug-substructure associations) will occupy the dominant position in the resource diffusion processes (**Figure 1C**).

Thirdly, another parameter  $\gamma$  was introduced to adjust the influence of hub nodes, namely those nodes connected with a large number of neighbor nodes. In current bSDTNBI, a positive value of  $\gamma$  will strengthen the influence of hub nodes, while a negative value of  $\gamma$  will weaken the influence of hub nodes, similar to what described in our previously developed NWNBI method (Cheng *et al.*, 2012).

Moreover, considering that new chemical entities can be labeled as special drugs without known targets, for each new chemical entity, a total amount =  $\alpha$  of initial resource was equally allocated to its substructure nodes in the initial resource allocation, and the weighted values of all its new chemical entity-substructure

associations were set to  $\beta$ .

#### Details of the resource diffusion processes in bSDTNBI

Denoting the set of new chemical entities as  $C = \{C_1, C_2, ..., C_{N_c}\}$ , the set of known drugs as  $D = \{D_1, D_2, ..., D_{N_D}\}$ , the set of chemical substructures as  $S = \{S_1, S_2, ..., S_{N_s}\}$ , and the set of target proteins as  $T = \{T_1, T_2, ..., T_{N_T}\}$ , the substructure-drug (and new chemical entity)-target network can be represented as a graph G(V, E), where  $V = C \cup D \cup S \cup T$  is the vertex set which contains four types of nodes (i.e. new chemical entities, drugs, substructures, and targets), and E is the edge set which contains three types of edges (i.e. new chemical entity-substructure associations, drug-substructure associations, and DTIs).

Mathematically, let  $M_{CS}$  and  $A_{CS}$  be three  $N_C \times N_S$  matrices, let  $M_{DS}$  and  $A_{DS}$  be two  $N_D \times N_S$  matrices, and let  $M_{DT}$  and  $A_{DT}$  be two  $N_D \times N_T$  matrices, respectively defined as:

$$M_{CS}(i,j) = \begin{cases} 1 & \text{if } C_i \text{ is linked with } S_j \\ 0 & \text{otherwise} \end{cases}$$
(1)

$$M_{DS}(i,j) = \begin{cases} 1 & \text{if } D_i \text{ is linked with } S_j \\ 0 & \text{otherwise} \end{cases}$$
(2)

$$M_{DT}(i,j) = \begin{cases} 1 & \text{if } D_i \text{ is linked with } T_j \\ 0 & \text{otherwise} \end{cases}$$
(3)

$$A_{CS}(i,j) = \alpha \cdot \frac{M_{CS}(i,j)}{\sum_{l=1}^{N_S} M_{CS}(i,l)}$$
(4)

$$A_{DS}(i,j) = \alpha \cdot \frac{M_{DS}(i,j)}{\sum_{l=1}^{N_S} M_{DS}(i,l)}$$
(5)

$$A_{DT}(i,j) = (1-\alpha) \cdot \frac{M_{DT}(i,j)}{\sum_{l=1}^{N_T} M_{DT}(i,l)}$$
(6)

The initial resource matrix A of graph G can be represented as:

$$A = \begin{bmatrix} O & O & A_{CS} & O \\ O & O & A_{DS} & A_{DT} \\ A_{CS}^{T} & A_{DS}^{T} & O & O \\ O & A_{DT}^{T} & O & O \end{bmatrix}$$
(7)

Let B, C, and W be three  $(N_D+N_S+N_T)$  order square matrices, respectively defined as:

$$B = \begin{bmatrix} O & O & O & O \\ O & O & A_{DS} & A_{DT} \\ O & A_{DS}^{T} & O & O \\ O & A_{DT}^{T} & O & O \end{bmatrix}$$
(8)

$$C(i, j) = B(i, j) \cdot \left[\sum_{l=1}^{N_C + N_D + N_S + N_T} B(l, j)\right]^{\gamma}$$
(9)

$$W(i,j) = \begin{cases} \frac{C(i,j)}{\sum_{l=1}^{N_c + N_D + N_S + N_T} C(i,l)} & \text{if } \sum_{l=1}^{N_c + N_D + N_S + N_T} C(i,l) \neq 0\\ 0 & \text{otherwise} \end{cases}$$
(10)

The final resource matrix F can be calculated via following equation:

$$F = A \times W^k \tag{11}$$

Where  $W^k$  is the transfer matrix, k is the number of resource spreading processes. The value of F'(i, N<sub>C</sub>+N<sub>D</sub>+N<sub>S</sub>+j) (0 < i ≤ N<sub>C</sub>, 0 < j ≤ N<sub>T</sub>) is the score of the interaction between new chemical entity C<sub>i</sub> and target T<sub>j</sub>. The value of F(N<sub>C</sub>+i, N<sub>C</sub>+N<sub>D</sub>+N<sub>S</sub>+j) (N<sub>C</sub> < i ≤ N<sub>C</sub>+N<sub>D</sub>, 0 < j ≤ N<sub>T</sub>) is the score of the interaction between drug D<sub>i</sub> and target T<sub>j</sub>.

#### References

Cheng FX, Zhou YD, Li WH, Liu GX, Tang Y (2012). Prediction of chemical-protein interactions network with weighted network-based inference method. *PLoS One* **7**(7): e41064.

Wu ZR, Cheng FX, Li J, Li WH, Liu GX, Tang Y (2016). SDTNBI: an integrated network and chemoinformatics tool for systematic prediction of drug–target interactions and drug repositioning. *Brief. Bioinform.* 10.1093/bib/bbw012.

# **Supplementary Figures**



Figure S1. The relationship among two parameters  $\alpha$ ,  $\beta$ , and the average AUC value for the models of GPCRs in 10-fold cross validation.



Figure S2. The relationship among two parameters  $\alpha$ ,  $\beta$ , and the average AUC value for the models of Kinases in 10-fold cross validation.

## **Supplementary Tables**

| Target  | FP      | N <sub>D</sub> | N <sub>S</sub> | N <sub>DS</sub> | Sparisty (%) |  |
|---------|---------|----------------|----------------|-----------------|--------------|--|
| GPCRs   | FP4     | 4,741          | 131            | 74,753          | 12.0         |  |
|         | KR      | 4,741          | 1,834          | 299,751         | 3.4          |  |
|         | MACCS   | 4,741          | 153            | 250,631         | 34.6         |  |
|         | PubChem | 4,741          | 627            | 742,237         | 25.0         |  |
| Kinases | FP4     | 2,827          | 122            | 44,290          | 12.8         |  |
|         | KR      | 2,827          | 1,618          | 129,906         | 2.8          |  |
|         | MACCS   | 2,827          | 149            | 148,122         | 35.2         |  |
|         | PubChem | 2,827          | 614            | 523,019         | 30.1         |  |
| Global  | FP4     | 1,872          | 148            | 26,714          | 9.6          |  |
|         | KR      | 1,872          | 1,959          | 93,738          | 2.6          |  |
|         | MACCS   | 1,872          | 153            | 86,446          | 30.2         |  |
|         | PubChem | 1,872          | 632            | 257,332         | 21.8         |  |

**Table S1.** The statistics of drug-substructure networks.

FP: the fingerprint type used to generate drug-substructure associations,  $N_D$ : the number of drugs,  $N_S$ : the number of substructures,  $N_{DS}$ : the number of drug-substructure associations, Sparsity: the ratio of  $N_{DS}$  to the number of all possible drug-substructure associations.

| Compound  | Structure | Rank | EC <sub>50</sub><br>(µM) | IC <sub>50</sub><br>(μM) |
|-----------|-----------|------|--------------------------|--------------------------|
| Z18499127 | HO OH N   | 1    |                          |                          |
| Z92457891 | HO OH F F | 1    | 0.33                     | 1.07                     |
| Z25218907 | НО ОН     | 1    |                          | 8.51                     |
| Z25218929 | HO OH Br  | 1    |                          | 1.28                     |
| Z25219066 | HO OH Br  | 1    |                          | 0.37                     |
| Z54500757 | HO OH S   | 1    |                          |                          |
| Z54108926 | HO OH O   | 1    |                          |                          |

**Table S2.** The *in vitro* bioassay results of newly predicted ligands for estrogen receptor  $\alpha$ .















The compound identifiers are from the Enamine database, except two control compounds (i.e. Estradiol and Tamoxifen). Rank is the position of estrogen receptor  $\alpha$  in the predicted target list for the corresponding compound.