A watercolor-style illustration of several pink flowers with delicate petals and visible stamens, set against a white background. The flowers are positioned in the upper half of the slide, partially overlapping a grey rectangular area that contains the text.

# A knowledge-based $T^2$ -statistic to perform pathway analysis for quantitative proteomic data

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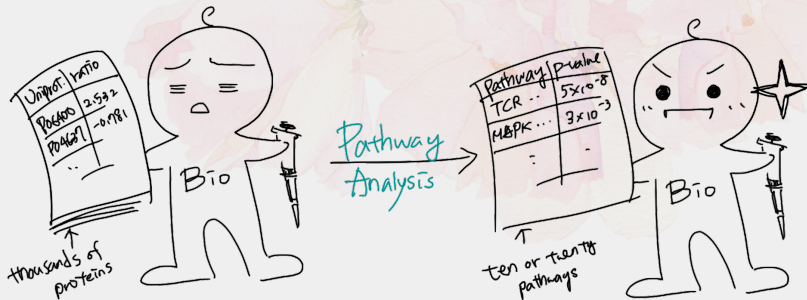
August 18, 2017

# Functional Analysis for Quantitative Proteomic Data



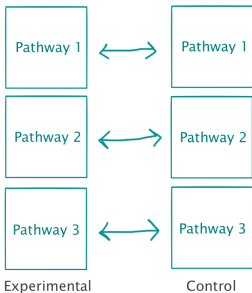
- ▶ Proteomic data v.s. gene expression data
  - ▶ **Smaller sample size** (number of experiments).
  - ▶ Fewer identified entities.
  - ▶ The results are sensitive to experimental conditions and instruments.
- ▶ Functional analysis
  - ▶ **Pathway analysis** (PLoS Computational Biology, 2017).
  - ▶ Responsive subpathway locating (working manuscript).

# Pathway analysis — Data interpretation

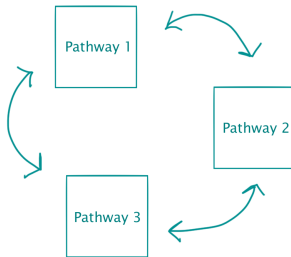


# Null hypothesis: **Self-contained** v.s. Competitive

Self-contained Null

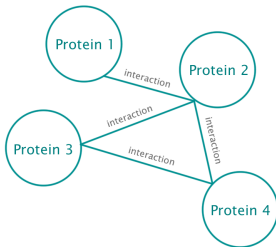


Competitive Null

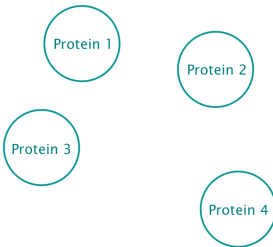


# Test statistic: **Multivariate** v.s. Univariate

Multivariate Statistic



Univariate Statistic



# The $T^2$ -statistic for pathway analysis

The proposed  $T^2$ -statistic for a specific pathway  $\mathcal{P}$  is then defined as,

$$T^2 = \mathbf{x}^T \mathbf{S}^{-1} \mathbf{x} \sim \chi_q^2$$

where

$\mathbf{x}$  is the vector of expression ratios,

$\mathbf{x}^T$  is the transpose of  $\mathbf{x}$ ,

$\mathbf{S}^{-1}$  is the inverse of **the covariance matrix  $\mathbf{S}$** , and

$q$  is the number of mapped proteins in  $\mathcal{P}$ .

# The $T^2$ -statistic for pathway analysis

- We use the **confidence score** provided by protein-protein interaction databases to represent the **strength of the covariance**, and the **expression direction** provided by the testing dataset to indicate the **sign of the covariance**.
- On the basis of the confidence scores in  $\mathfrak{J}$  and the protein expression ratios, each element  $s_{ij}$  of  $\mathbf{S}$  is determined by the following four rules:

$$s_{ij} = \begin{cases} 0.4 & \text{if } i = j. \\ c_{p_i p_j} & \text{if } i \neq j, c_{p_i p_j} \in \mathfrak{J}, \text{ and } x_i \cdot x_j \geq 0. \\ -c_{p_i p_j} & \text{if } i \neq j, c_{p_i p_j} \in \mathfrak{J}, \text{ and } x_i \cdot x_j < 0. \\ 0.0 & \text{if } i \neq j \text{ and } c_{p_i p_j} \notin \mathfrak{J}. \end{cases}$$

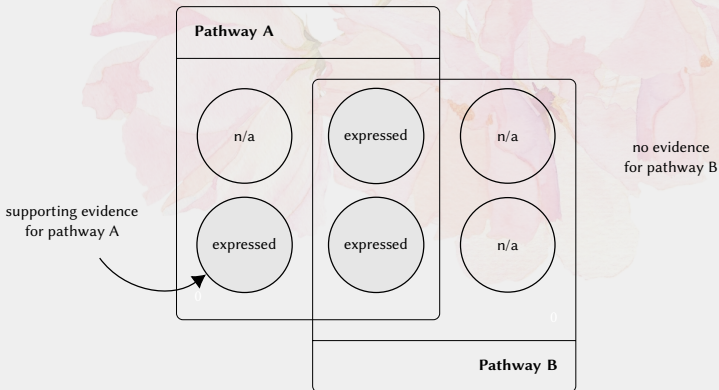
# The $T^2$ -statistic for pathway analysis

$$T^2 = \mathbf{x}^T \mathbf{S}^{-1} \mathbf{x} \sim \chi_q^2$$

- ▶ If  $\mathbf{S}$  is degenerate, we construct a Moore-Penrose pseudoinverse of  $\mathbf{S}$  as a substitute, and  $q$  becomes the rank of  $\mathbf{S}$ .
- ▶ The  $p$ -value of the pathway  $\mathcal{P}$  is derived from the  $\chi_q^2$  distribution.



# Pathway integration



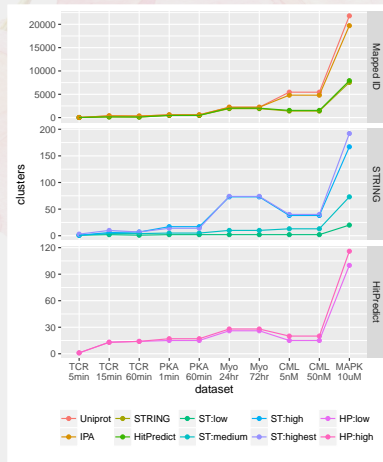
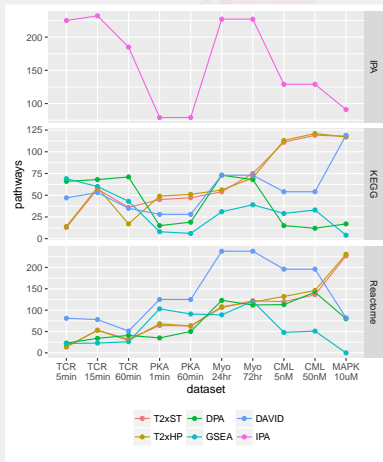
# Performance evaluation — Really HARD!!!

- ▶ No accepted gold-standards.
- ▶ We tried to match the results reported by these methods to the biological ideas provided by the original publication (true positives).
- ▶ A good statistic should be able to test if a pathway is significant:
  - ▶ if a statistic reports very little significant pathways, it might have the problem of false negatives;
  - ▶ if a statistic reports a large number of significant pathways, it might have the problem of false positive.
- ▶ We supposed that the number of significant pathways is one of the attribute to evaluate these methods.

# General comparison

Tools	Significance requirement	Ranking statistic
$T^2$	raw $p$ -value $\leq 0.05$	number of mapped proteins
DPA	raw $p$ -value $\leq 0.05$	$p$ -value
GSEA	FDR adjusted $p$ -value (i.e. $q$ -value) $\leq 0.25$	NES
DAVID	EASE $\leq 0.1$	$p$ -value
IPA	Benjamini corrected $p$ -value $\leq 0.05$	$p$ -value

# General comparison: Consistency



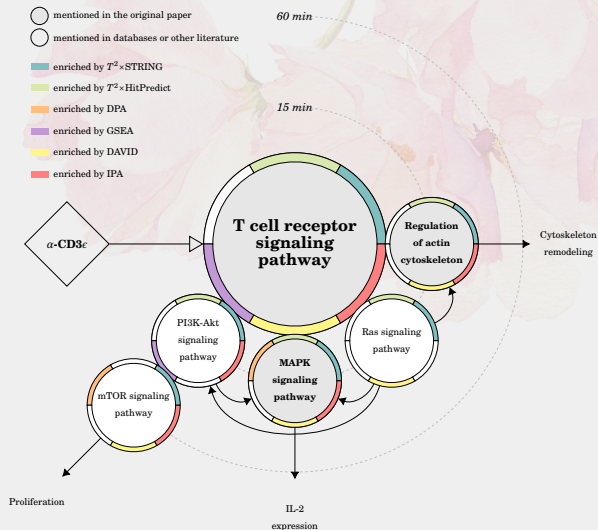
# General comparison: Consistency

Database	KEGG					Reactome					IPA
Method	$T^2 \times ST$	$T^2 \times HP$	DPA	GSEA	DAVID	$T^2 \times ST$	$T^2 \times HP$	DPA	GSEA	DAVID	IPA
Uniprot	+	+	-	-	+	+	+	+	-	-	-
IPA	+	+	-	-	+	+	+	+	-	-	-
STRING	+	+	-	-	+	+	+	+	-	+	-
HitPredict	+	+	-	-	+	+	+	+	-	+	-
ST: low	+	+	-	-	+	+	+	+	-	-	-
ST: medium	+	+	-	-	+	+	+	+	-	-	-
ST: high	+	+	-	-	+	+	+	+	-	+	-
ST: highest	+	+	-	-	+	+	+	+	-	+	-
HP: low	+	+	-	-	+	+	+	+	-	-	-
HP: high	+	+	-	-	+	+	+	+	-	-	-

# General comparison: Target pathways

Dataset	TCR			PKA		Myogenesis		CML		MAPK
Treatment	$\alpha$ -CD3 $\epsilon$			PGE2		Serum-free		Dasatinib		U0126
	5 min	15 min	60 min	1 min	60 min	24 hr	72 hr	5 nM	50 nM	10 $\mu$ M
KEGG pathway	T cell receptor signaling pathway			cAMP signaling pathway		ECM-receptor interaction		Chronic myeloid leukemia		MAPK signaling pathway
$T^2 \times ST$ <i>p</i> -value	1/13 < 0.0001	1/57 < 0.0001	2/36 0.0019	16/45 < 0.0001	17/47 < 0.0001	15/54 < 0.0001	16/75 < 0.0001	20/111 < 0.0001	23/119 < 0.0001	3/118 < 0.0001
$T^2 \times HP$ <i>p</i> -value	1/14 < 0.0001	1/59 0.0002	2/17 0.0011	17/49 < 0.0001	18/51 < 0.0001	16/56 < 0.0001	15/70 < 0.0001	20/113 < 0.0001	25/121 < 0.0001	3/117 < 0.0001
DPA <i>p</i> -value	-	8/68 0.0007	20/71 0.0024	11/15 0.0414	7/19 0.0178	18/73 0.0004	-	-	-	2/17 0.0007
GSEA <i>q</i> -value	68/69 0.1786	7/60 0.1039	-	-	-	22/31 0.1416	-	-	-	-
DAVID <i>p</i> -value	1/47 < 0.0001	1/53 < 0.0001	1/35 < 0.0001	-	-	40/73 0.0041	40/73 0.0041	8/54 0.0004	8/54 0.0004	8/119 0.0002
IPA <i>p</i> -value	1/225 < 0.0001	2/232 < 0.0001	89/185 0.0022	60/79 0.0186	60/79 0.0191	-	-	52/129 0.0013	52/129 0.0013	-

# Case Study: TCR downstream phosphoproteome

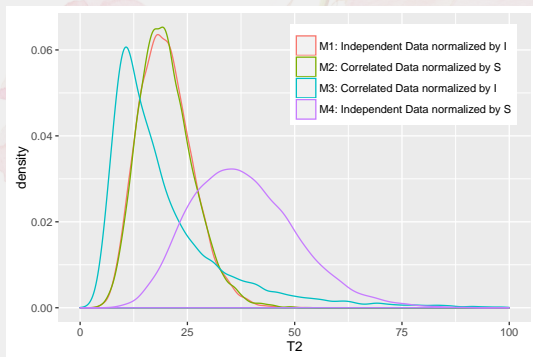


# Case Study: TCR downstream phosphoproteome

Dataset	Pathway Title	$T^2 \times ST$	$T^2 \times HT$	DPA	GSEA	DAVID	IPA
5 min	T cell receptor signaling pathway	1/13	1/14	-	68/69	1/47	1/225
15 min	Ras signaling pathway	2/57	2/59	-	-	10/53	-
	Regulation of actin cytoskeleton	3/57	3/59	-	-	9/53	15/232
	MAPK signaling pathway	4/57	4/59	65/68	-	41/53	4/232
	PI3K-Akt signaling pathway	24/57	25/59	-	6/60	-	169/232
60 min	mTOR signaling pathway	25/36	-	67/71	-	24/35	68/185



# Accurate and inaccurate estimation: Toy example



	Correlated Data (generated by $S$ )	Independent Data (generated by $I$ )
Correlated Null (normalized by $S$ )	M2: accurate estimation	M4: inaccurate due to false positive PPI scores
Independent Null (normalized by $I$ )	M3: inaccurate due to incomplete knowledge	M1: accurate estimation

# Accurate and inaccurate estimation: Real cases

Dataset	Experiment	Pathway Title	original <i>p</i> -value	30% permuted	60% permuted	30% purged	60% purged
TCR	5 min	T cell receptor signaling pathway	< 0.0001	100%	100%	100%	100%
	15 min	Ras signaling pathway	0.0017	100%	100%	100%	100%
		Regulation of actin cytoskeleton	< 0.0001	100%	100%	100%	100%
		MAPK signaling pathway	< 0.0001	100%	100%	100%	100%
		PI3K-Akt signaling pathway	< 0.0001	100%	100%	100%	100%
	60 min	mTOR signaling pathway	0.0002	97%	98%	100%	100%
PKA	1 min	Regulation of actin cytoskeleton	< 0.0001	100%	100%	100%	100%
		PI3K-Akt signaling pathway	< 0.0001	92%	81%	86%	84%
		MAPK signaling pathway	< 0.0001	100%	100%	100%	100%
		Rap1 signaling pathway	< 0.0001	100%	100%	100%	100%
		cAMP signaling pathway	< 0.0001	100%	100%	100%	100%
		Glycolysis / Gluconeogenesis	1	0%	0%	0%	0%
	60 min	Cell cycle	< 0.0001	100%	100%	100%	100%
		mTOR signaling pathway	0.0024	100%	100%	100%	100%
		Base excision repair	< 0.0001	100%	100%	100%	100%



Thank  
You ♥