



Statistical Methods for Quantitative MS-Based Proteomics:

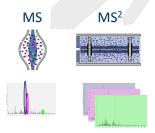
1. Identification

Lieven Clement

Statistics and Genomics Seminar, UCBerkeley, California

Challenges in Label Free MS-based Quatitative proteomics

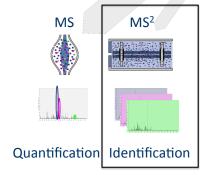




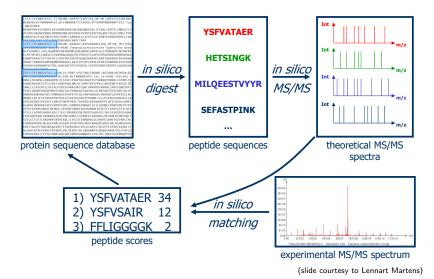
Quantification Identification

Challenges in Label Free MS-based Quatitative proteomics





Identification



	Called Bad	Called Correct	
Bad hit	TN	FP	m_0
Correct hit	FN	TP	m_1
 Total	NR	R	т

- TN: number of true negatives
- FP: number of false positives
- FN: number of false negatives
- TP: number of true positives
- NR: number of non-rejections, R: number of rejections

		Called Bad	Called Correct	
	Bad hit	TN	FP	m_0
	Correct hit	FN	TP	m_1
_	Total	NR	R	т

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Random Variables

		Called Bad	Called Correct	
	Bad hit	TN	FP	m_0
Unobservable	Correct hit	FN	TP	m_1
Observable	Total	NR	R	т



		Called Bad	Called Correct	
Unobservable	Bad hit	TN	FP	m_0
	Correct hit	FN	TP	m_1
Observable	Total	NR	R	m

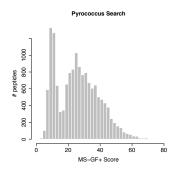
 $FDP = \frac{FP}{FP+TP}$. But is unknown! (FDP: false discovery proportion)



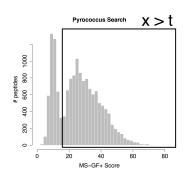
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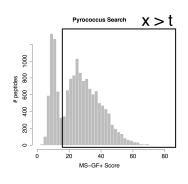
$$FDR = E\left[\frac{FP}{FP+TP}\right]$$
. (FDR: false discovery rate) What does it mean?





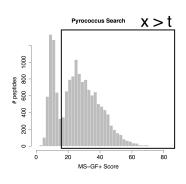






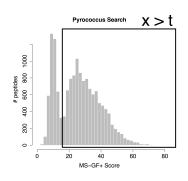
$$f(x) = \pi_0 f_0(x) + (1 - \pi_0) f_1(x)$$



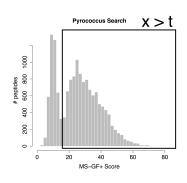


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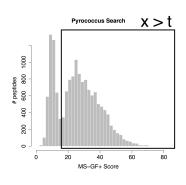
$$FDR(t) = E\left[\frac{FP}{FP+TP}\right]$$



Score threshold t? $f(x) = \pi_0 f_0(x) + (1 - \pi_0) f_1(x)$ $\text{FDR}(t) = E\left[\frac{FP}{FP + TP}\right]$ $\text{FDR}(t) = \frac{mPr[FP]Pr[x > t|FP]}{mPr[x > t]}$



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$$f(x) = \pi_0 f_0(x) + (1 - \pi_0) f_1(x)$$

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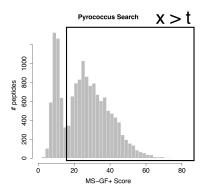
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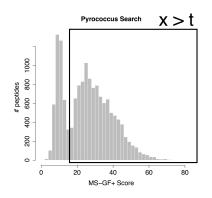
$$FDR(t) = \frac{\pi_0[1 - F_0(t)]}{1 - F(t)}$$
 with $F.(t) = \int_{-\infty}^{t} f.(x) dx$





$$FDR(t) = \frac{\pi_0 [1 - F_0(t)]}{1 - F(t)} = \frac{\pi_0 Pr[x > t | FP]}{Pr[x > t]}$$



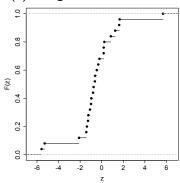


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$$FDR(t) = \frac{\pi_0 [1 - F_0(t)]}{1 - \frac{\#x \le t}{m}} = \frac{\pi_0 Pr[x > t | FP]}{\frac{\#x > t}{m}}$$

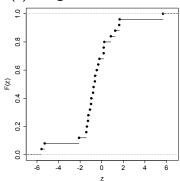


• F(t) using the ECDF



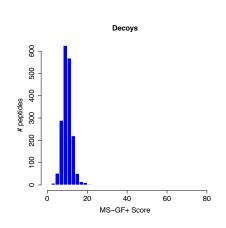


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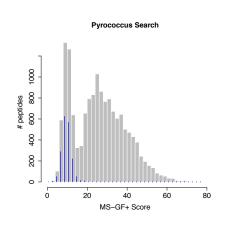


• How to characterize $F_0(t)$ and π_0 in proteomics?

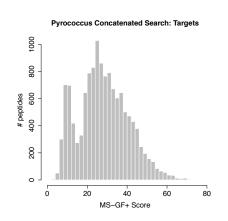




- Searching against decoy databases to generate representative bad hits
- Reversed databases are a popular choice



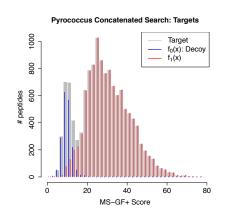
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- Assumption that bad hits have an equal probability to map on forward (target) and reverse database (decoy)

$$\hat{\pi}_0 = \frac{\# decoys}{\# targets}$$

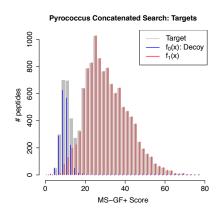




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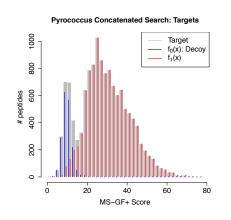




Score cuttoff?

$$FDR(x) = E\left[\frac{FP}{FP + TP}\right]$$

$$\widehat{\mathsf{FDR}}(x) = \frac{\#\mathsf{decoys}|X \ge x}{\#\mathsf{targets}|X \ge x}$$



Score cuttoff?

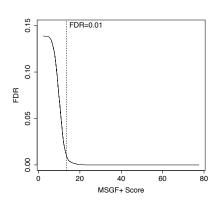
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$$\widehat{\mathsf{FDR}}(x) = \hat{\pi}_0 \frac{1 - \bar{F}_0(x)}{1 - \bar{F}(x)}$$





Score cuttoff?

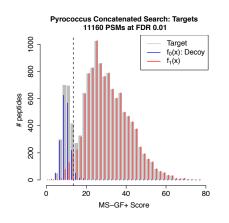
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Score cuttoff?

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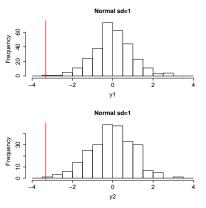
$$\widehat{\mathsf{FDR}}(x) = \hat{\pi}_0 \frac{1 - \bar{F}_0(x)}{1 - \bar{F}(x)}$$

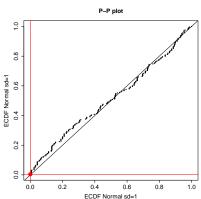


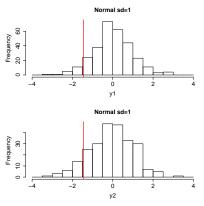
We have to evaluate that

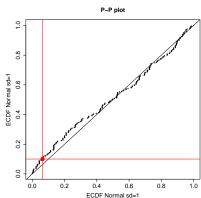
- The decoys are good simulations of the targets: compare $\bar{F}_0(x)$ with $\bar{F}(x)$
- $\hat{\pi}_0 = \frac{\#decoys}{\#targets}$ is a good estimator for π_0 .
- We will use Probability-Probability-plots for this purpose.
- They plot the ECDFs from two samples in function of each other.

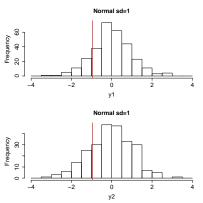


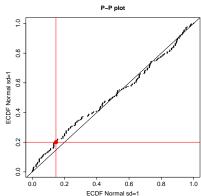


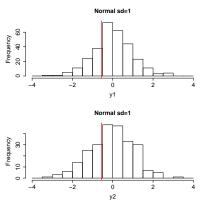


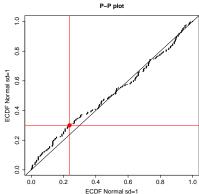


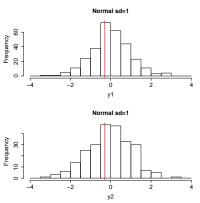


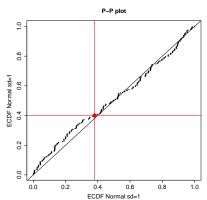


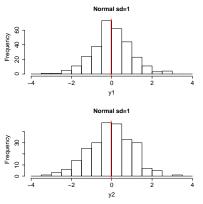


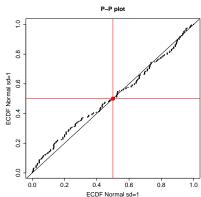


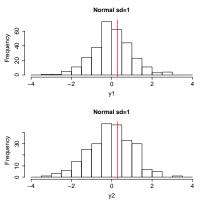


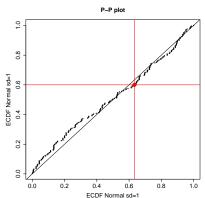


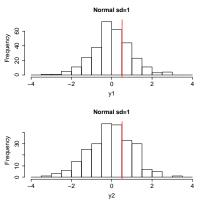


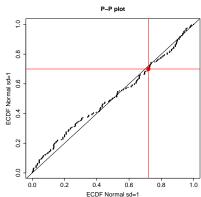


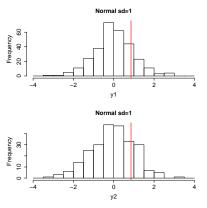


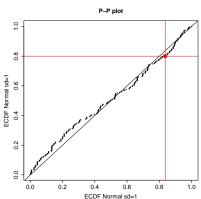


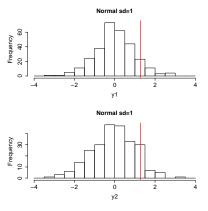


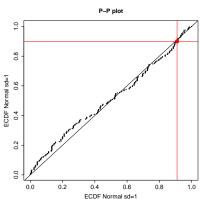


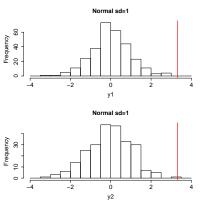


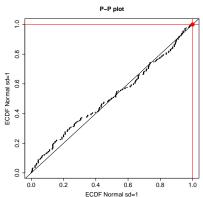




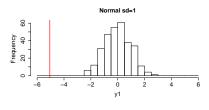


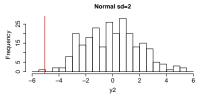


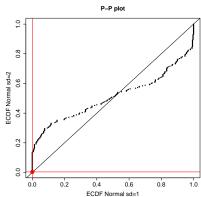




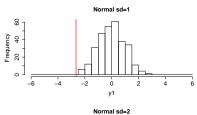


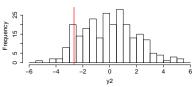


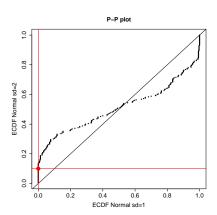




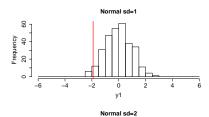


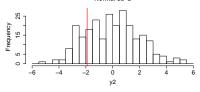


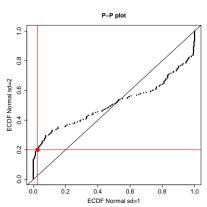




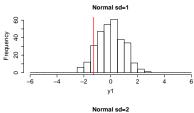


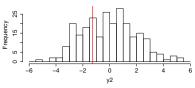


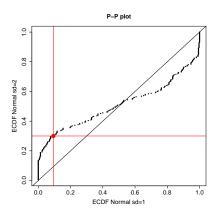




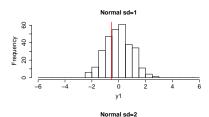


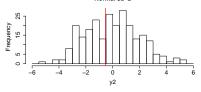


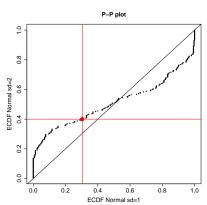




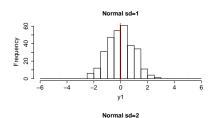


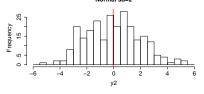


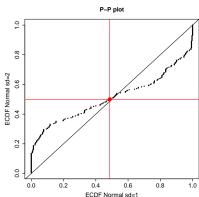




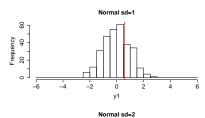


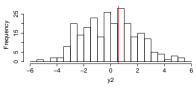


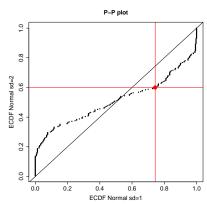




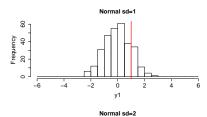


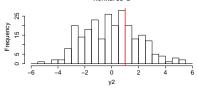


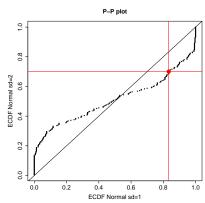




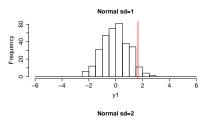


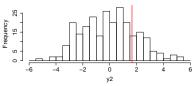


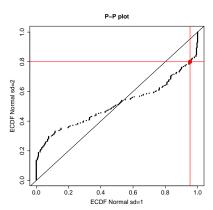




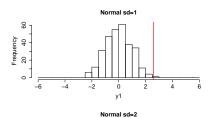


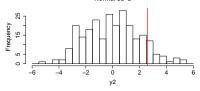


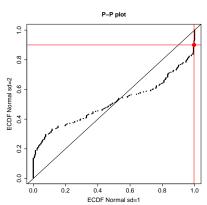




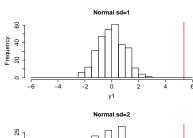


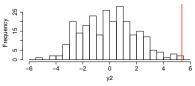


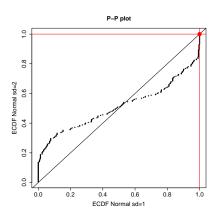




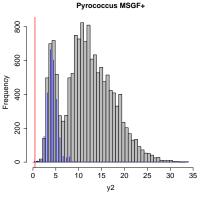


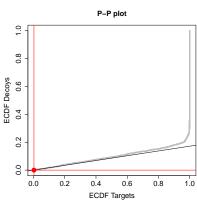




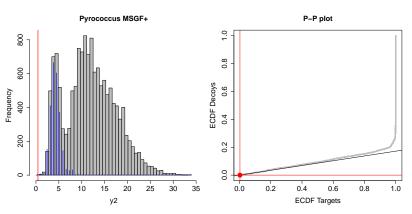












What about $\hat{\pi}_0$?



