High throughput, multiplex analysis of gene expression*

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*With special thanks to Chiara Mazzanti, Ph.D.

Revolutions in the analysis of gene expression

Protein function Protein concentration Post-translational modification Transcriptional control Genetic variation Genomics

Multiplex gene analysis: Why?

Complex diseases involve expression modulation 40,000 genes, perhaps more We are not so clever The ultimate goal is to reconstruct the biochemical and functional networks of the cell. One can not understand a network by studying one gene.

Methods for multiplex expression analysis

Two dimensional protein electrophoresis [2DE] Differential display Representational difference analysis [RDA] Serial analysis of gene expression [SAGE] RNA array methods



cDNA

Oligonucleotide chips

Probes: amplified spotted cDNAs http://cmgm.standford.edu http://www.nhgri.nih.gov Probes: Oligos synthesized *in situ* or conventionally & followed by immobilization Oligos as long as 80bp <u>http://www.affymetrix.com</u> <u>http://www.rii.com</u> (Agilent)

Array preparation

Microarray

Genechip

cDNA microarray

Denaturing

High-density oligonucleotide microarrays



Available genes: 10k – NCI [\$50] 20k - Affymetrix

Array preparation

Sample preparation (direct labeling)



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Arrays Hs-UniGem2-v2p9-060701 Blocks: (18x18) x32 Features/Spots: 9984 235um spacing

10k



Post-hybridization microarray analysis

software

Typical software (GenPix Pro)

Image scanning

Cy3/Cy5 signal normalization

Image cleaning

GAL (gene array list) loading

Image analysis

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cDNA Microarray

Ratio Cy5/Cy3

Ratio >2, < 0.5

Array vs Chip

Case Cy5 Control Cy3

Oligo-microchip

Ratio Array1/Array2

Microsoft Excel - Project 06-06.xls

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85	Hs.10240	2 1.254720	δ	1.269755	-1.27643	0.129453	-0.18873	0.892415	-1.1738	0.366394	1.56632
88	Hs.10242				-1.38841	0.845519	0.910564	-1.93052		2.535235	-0.82697
90	Hs.10247			-0.85649	0.7785	0.216988	1.118692	2.321802	-0.88262	-1.04328	-0.30813
91	Hs.10247	1 -1.26308	5 1.652778	0.510338	2.936481	0.793835	0.22033	1.213758	1.029589	-0.49348	-0.4854
94	Hs.10248	4 1.555718	3 -1.59462	3.104822	0.605855	0.575335	-0.12753	-3.97658	-0.37731	3.861236	-0.76947
144	Hs.10359	-0.76171		-0.71029	-0.57138	-0.17907	1.15	1.047846	1.105924	-0.5134	-0.56282
150	Hs.10380	4 -0.08697	7 -1.63766	2.163743	2.87046	-0.14155	0.268424	2.827162	-4.05E-01	-1.8208	0.80428
164	Hs.10432	0.514454	4 -0.23651	-0.98676	-0.78038	-0.25038	-6.18E-01	-1.0595	0.075987	-0.25107	-1.37771
167	Hs.10438	2 -0.7771	3.80E-01	-0.86862	0.032264	0.817435	0.670793	0.535541	0.073581	-1.07135	-0.86672
183	Hs.10474	4 -0.39398	0.125976	-0.92687	-0.71241	-0.29256	-0.43788	-0.2574	0.6149	-8.43E-01	-1.15899
184	Hs.10476	7 -0.11989	-0.02745	-1.26101	-0.90296	0.266887	-0.54701	-0.04424	0.460116	-0.51732	-1.53592
185	Hs.10478	8	-0.1059	-1.18769	-0.60656	0.352824	-0.29736	-1.13804	0.518579	-0.25856	-1.4465

http://linus.nci.nih.gov/cgi-bin/brb/download.cgi

scheduled to begin at 9am and last until 3pm (6 hours). Posted 8:30am 6/11/2002

- <u>Gateway</u> Data Upload, Access and Analysis Tools (Note: Must be a registered mAdb user - Login/Password required.)
- <u>Array Ordering/Tracking WEBsite</u> (Note: Different WEBsite Administered by the ATC, a separate login needed)
- mAdb <u>Account Request</u> Request a new user account.
- mAdb Training via <u>CIT</u>
 o mAdb Basic Informatics Course Description and Sign up for 2 hour introductory class

List of up- and downregulated genes

Schulze et al, Nature Cell Biology, August 2001

Array Analysis Methods

• Gene Discovery

- Outlier detection simple and group logic retrieval tools; single and multiple array viewers
- Scatter plots
- Pattern Discovery
 - Clustering Hierarchical, K-means
 - Principal Components Analysis; Multidimensional Scaling
- Pattern Prediction

What is Clustering?

Clustering algorithms are discovery algorithms that find structure in data.

Discovery algorithms develop their own sorting parameters without human intervention.

Hierarchical Clustering

Division of Clinical Sciences

NCI

CIT

Center for Information Technology

mAdb Hierarchical Tool Form 1.00

Note the -- def marks items which lead to additional help when clicked

Hierarchica	al Clustering Options … 🐠
Genes	Non-centered Metric
Arrays	Non-centered Metric
Distance Me	tric Pearson Correlation 💌
Na	me (optional) DEMO
	Cluster

Home | Analysis Tools | Forums | Reference Info | Program Downloads | GeneCards

Hierarchical clustering

Mock phylogenetic dendrograms

Dendrogram Construction for Hierarchical Clustering

- Merge nearest neighbors
- Merge more distant datapoints

K-means clustering

Genes are divided into k user-defined, equallysized groups.

Gene group centroids and individual genes are then reassigned.

The process is reiterated until group compositions stabilize.

Clustering

Clusters are also developed from <u>random data:</u>

- Are clustered genes related in function?
- Do clusters group related samples/tissues/diseases/treatments?

Clustering of Melanoma Tumors Using Single Linkage

Clustering of Melanoma Tumors Using Complete Linkage

Clustering algorithms: Differing results

(Bittner et al., Nature, 2000)

mAdb BioInformatics Project: Current

- Links to external data sources
 - dbEST
 - GeneCards
 - LocusLink
 - MGD
 - GenBank
 - Stanford GO (Genome Ontology)
 - PubMed
 - UniGene

Automatic updates of external data sources

*	А	Α	A *			••	• •	•	••••
18	#19	#20	#21	Well ID	Feature ID	Map	UniGene	Gene	Descriptions/Gene Ontology Terms
206	1.0844		2.5936	164548	IncytePD:1804666	2q11.2	Hs.153884	APACD	ATP binding protein associated with cell different
402	0.0509	0.8579	0.5107	160680	IncytePD:3948420	3q26.2-qter	Hs.75736	APOD	apolipoprotein D lipid metabolism; extracellular space; high-density
240	0.8441		4.6527	164724	IncytePD:2345343	11p15	Hs.74515	ARNTL	aryl hydrocarbon receptor nuclear translocator-lil circadian rhythm
690	0.9009	3.5553	2.9125	160419	IncytePD:2056163	1p12	Hs.184270	CAPZA1	capping protein (actin filament) muscle Z-line, alp cell motility; actin binding; actin cytoskeleton; acti
152	0.2024	0.6577	5.6715	161765	IncytePD:4271973	7q31.1	Hs.74034	CAV1	caveolin 1, caveolae protein, 22kD caveolae; tumor suppressor; structural protein; inf
022	1.3037	2.0487	3.7294	168759	IncytePD:157510	1q31.1	Hs.75184	CHI3L1	chitinase 3-like 1 (cartilage glycoprotein-39) extracellular space; extracellular matrix; extracellu
701			6.7045	166687	IncytePD:2257362	3q28-q29	Hs.7327	CLDN1	claudin 1 cell adhesion; tight junction; integral membrane pr
556			9.5932	165783	IncytePD:782235	17q21.3-q22.1	Hs. 172928	COL1A1	collagen, type I, alpha 1 collagen; collagen type I; skeletal development; e
180	0.1037	0.4395	0.3940	169365	IncytePD:585432	15q24	Hs.346950	CRABP1	cellular retinoic acid binding protein 1 cytoplasm; retinoid binding; signal transduction; d
702	2.1474	3.3447	3.4755	169317	IncytePD:2806166	8p22	Hs.297939	CTSB	cathepsin B lysosome; cathepsin B; intracellular; proteolysis a
255	0.3610	0.2738	0.1496	165580	IncytePD:1514989	1p31-p22	Hs.8867	CYR61	cysteine-rich, angiogenic inducer, 61 cell proliferation; embryogenesis and morphogene
378	0.1601	0.2189	1.4210	164649	IncytePD:1969055	1q12-q23	Hs.80552	DPT	dermatopontin cell adhesion; protein binding

Hierarchical Clustering (Alizadeh *et al.*, Nature, Feb. 2000)

Issues for Study Design

Appropriateness of model Exposure, genotype, timing Tissue and cells Controls: Sieving the genes The issue of completeness # genes expressed in tissue % modulated at chosen level

Pitfalls in multiplex analysis of gene expression

Nonspecific changes in gene expression Quantitatively small changes can be critical mRNA modulation is often not the mechanism Allosteric regulation of enzymes Receptor desensitization

Conclusions

Multiplex analysis of gene expression has great potential for identifying intermediate processes in neural diseases Realization of the potential of these inherently industrial methods requires more incisive experimental designs The downstream biology is inherently nonindustrial Appropriate controls can narrow the field of candidate genes These methods should lead to the identification of new pathways in disease and an understanding of sharing of pathways between diseases