

DNA Chip Technology

Benedikt Brors

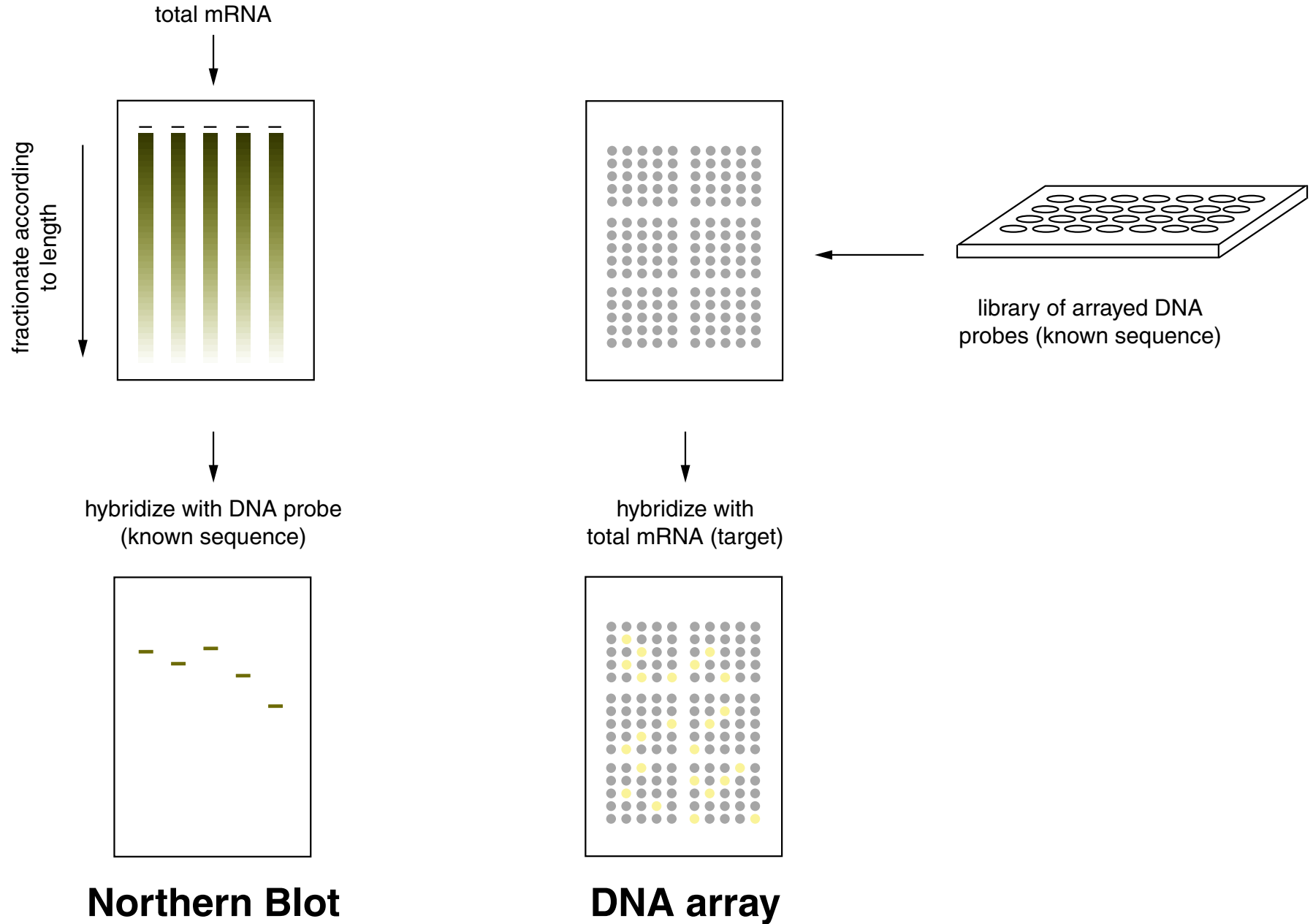
Dept. Intelligent Bioinformatics Systems

German Cancer Research Center

Why DNA Chips ?

- Functional genomics: get information about genes that is unavailable from sequence
- Understand how cells/organisms react to external stimuli
- Understand gene regulation networks
- Determine what makes the difference between healthy and diseased tissue
- Simply do 15,000 Northern Blots at a time

Comparison Northern blot ↔ DNA array

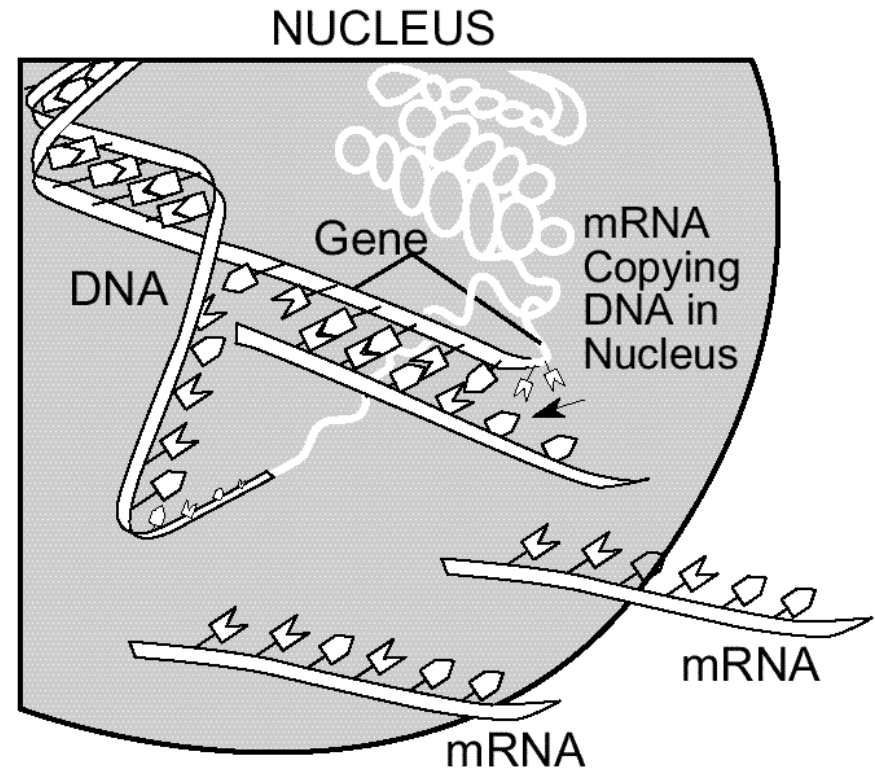


Functional Genomics

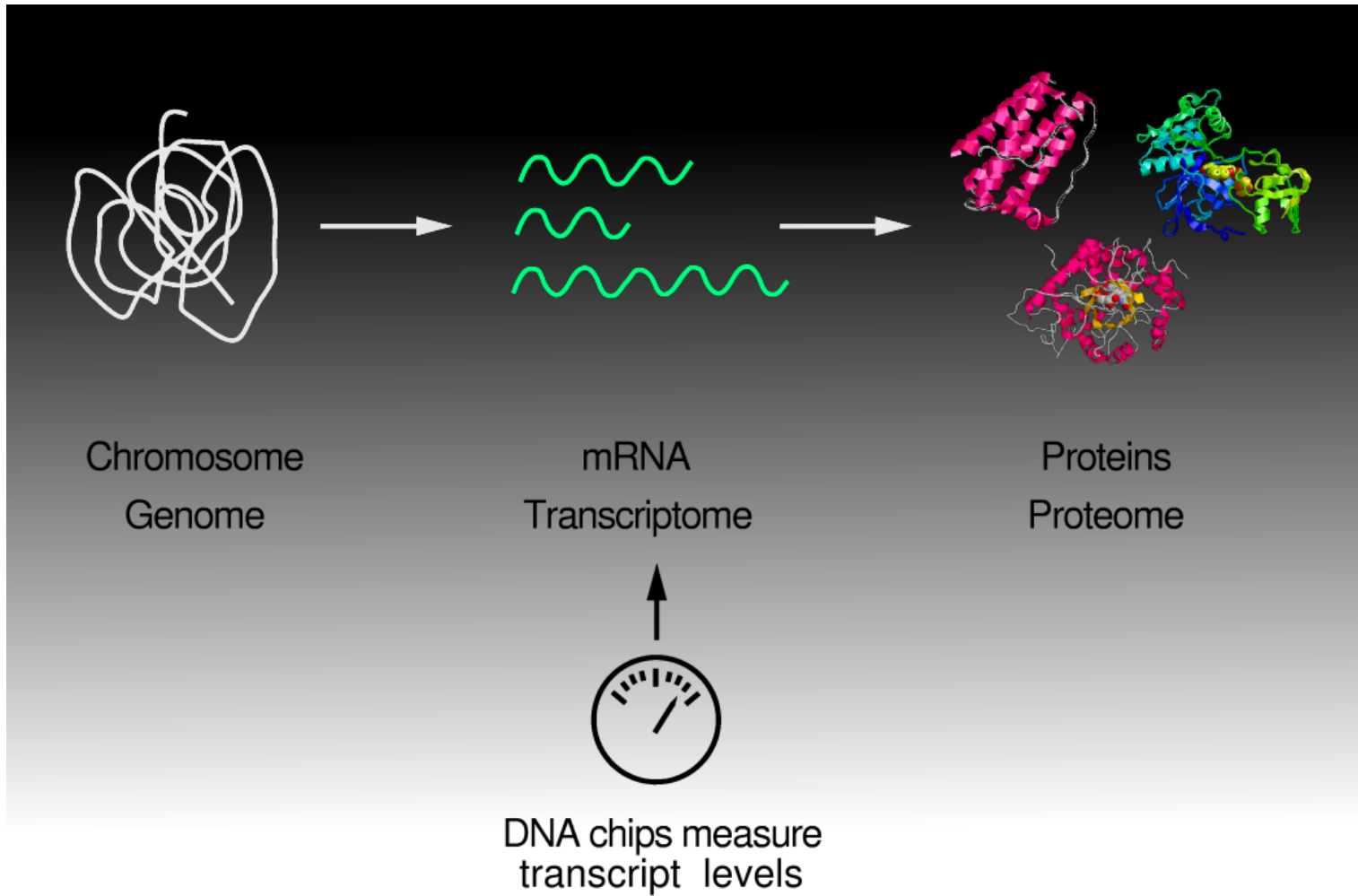
- There may be 100,000 different transcripts in human cells ($\pm 50,000$)
- We only have sound information on $\approx 12,000$ genes
- All cells have the same genome, but there are more than 200 cell types in a single organism
- Gene expression determines the cell type (neuron, lymphocyte, fibroblast etc.) and directs development of an organism (by spatial/temporal patterns)
- DNA chip technology promises to solve such unanswered questions

Basic Biology

- Genes contain construction information
- All structure and function is made up by proteins
- mRNA is sort of 'working copy', containing design of one protein
- mRNA is transferred to cytoplasm where protein is made



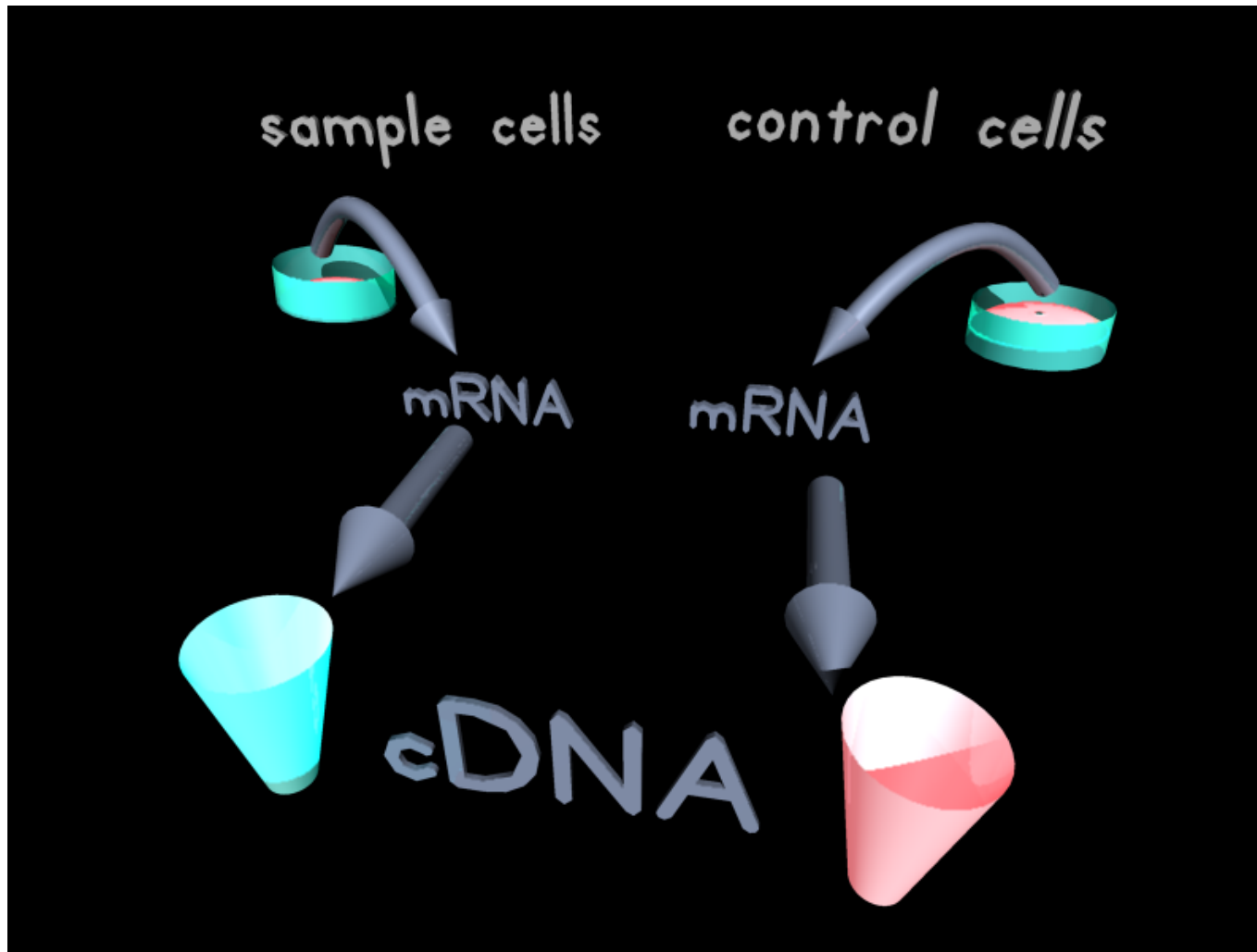
More Schematically ...



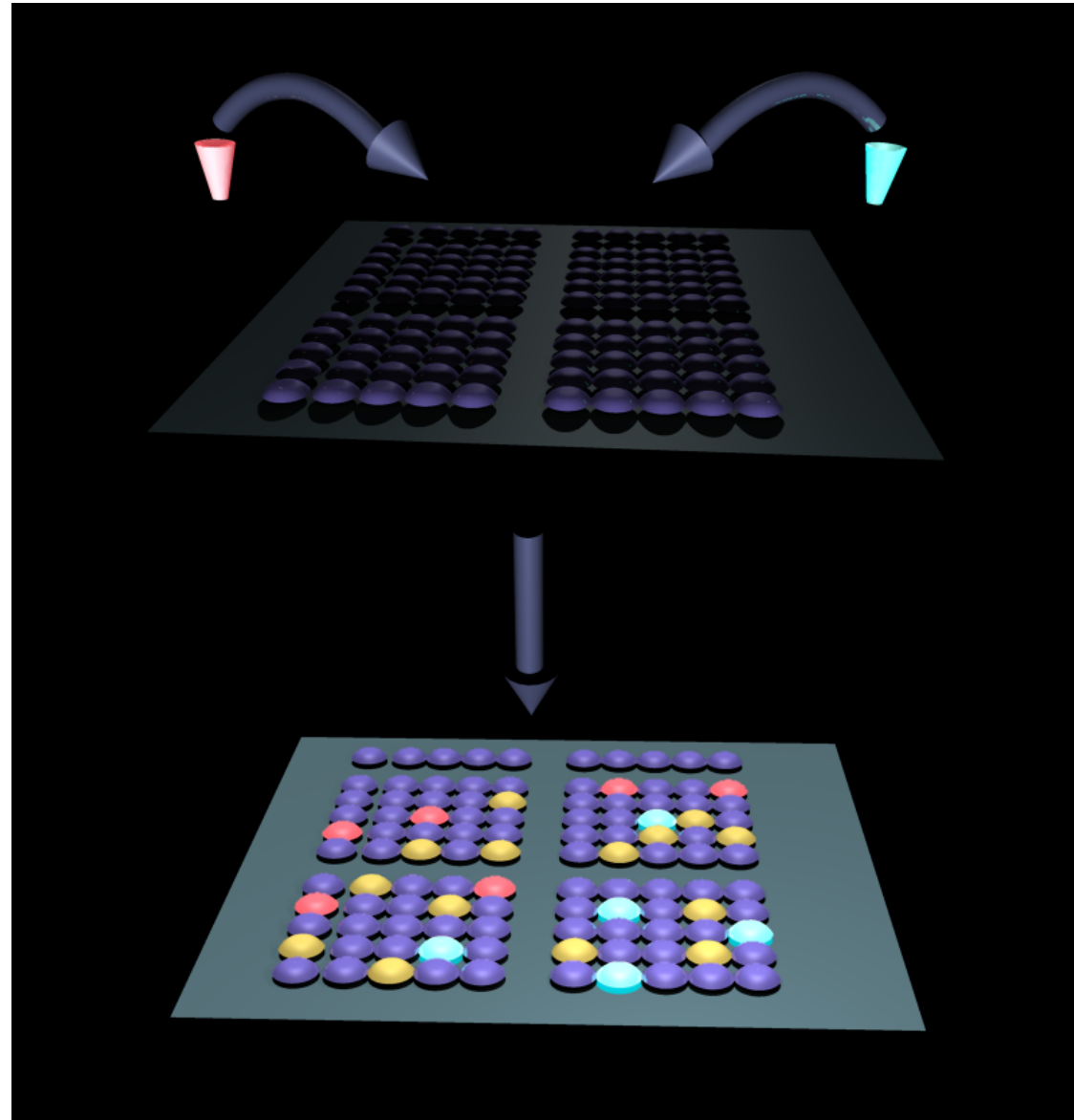
DNA Chip Technology 1

- Array: Small glass slide, contains 100s to 10,000s of DNA fragments ('spots') on few cm²
- Each DNA fragment will bind specifically a complementary DNA/RNA: 'Hybridization'
- 'Active' (transcribed) genes can be extracted from cells/tissues, labeled and hybridized to the array ⇒ 'active' genes will light up on the array

DNA Chip Technology 2



DNA Chip Technology 3



DNA Chip Technology 4

- Chip is read out by video camera
- Digitized image is analyzed by image analysis software

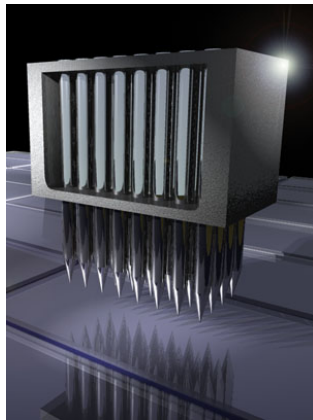
- Result: list of numbers

	R	G
spot1	1,346.2	1,575.8
spot2	100,326.1	30,872.0
spot3	987.1	177.2
spot4	(...)	(...)
(...)		

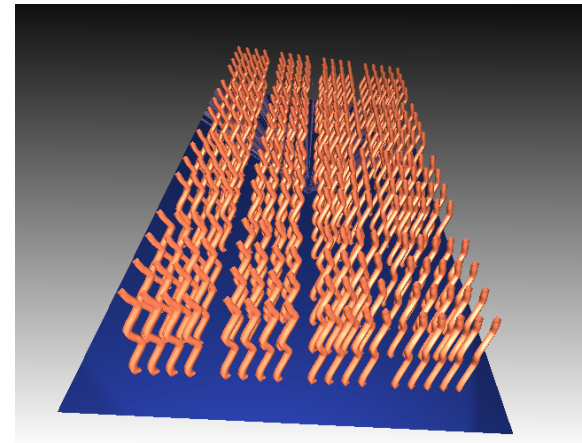
- N.B. the second column, 'G', is missing for one-color experiments

Competing Technologies

- Two systems: printed/‘spotted’ chips and on-chip synthesis



(A)



(B)

- For (A) mostly long DNA strands (500–3000 nt)
- For (B) only oligonucleotides (≤ 25 nt)

Printed/Spotted Chips

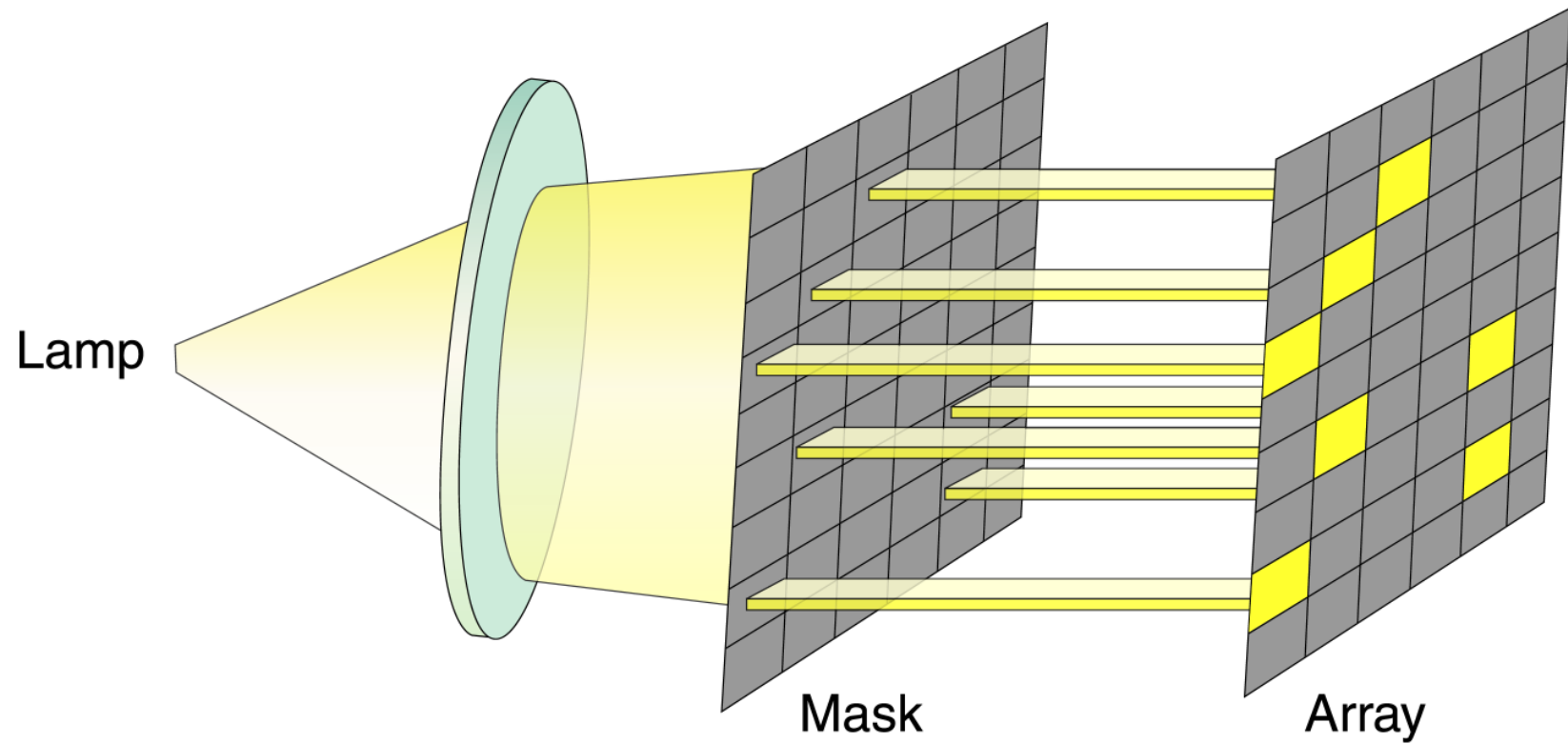
- Any DNA fragment can be put on the chip. Often, cDNA from libraries is used; to get a suitable amount for spotting, mass PCR has to be used. However, though performed less frequently, synthetic oligonucleotides can be used as well.■
- For cDNA chips, a new PCR reaction has to be performed for every batch of chips, these are quite different from each other with respect to the amount of DNA bound in a certain spot.■
- DNA fragments are transferred to the chip either by a spotting robot that transfers nanoliter quantities of liquid, or by an ink-jet like device.

- Spotted or printed chips are usually hybridized with two differently labeled mRNA preparations (i.e. their cDNA representation). By competitively hybridizing with two targets, the DNA amount in a single spot becomes less important (but not irrelevant!!).

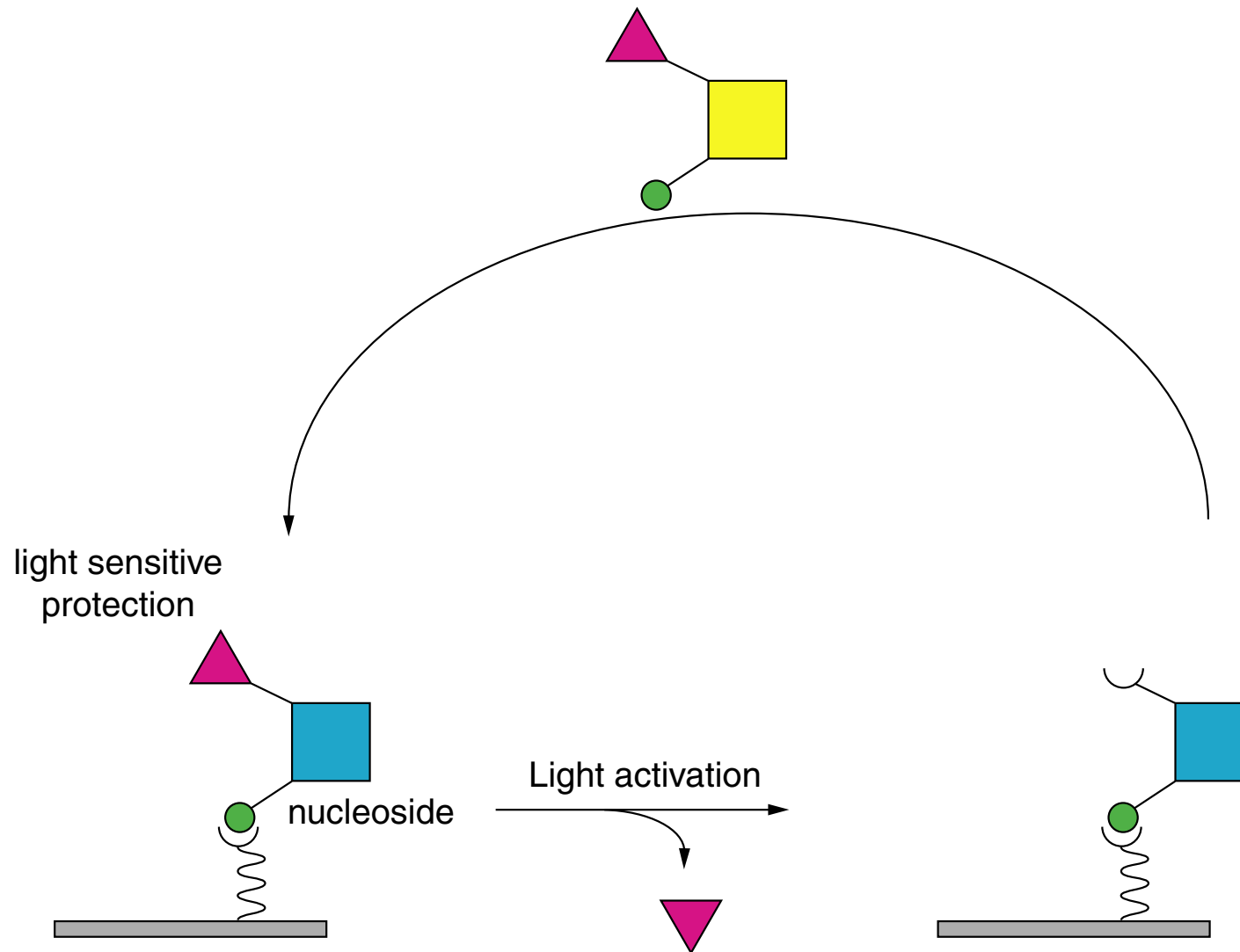
On-chip synthesis of oligonucleotides

- Oligonucleotides are built up on the chip surface by sequentially elongating the growing chain with a single nucleotide. To determine the sequence of the final oligonucleotides on each position of the chip, a process called *photolithography* is used.■
- As chemical yield of the stepwise elongation is limited, oligonucleotides can't be grown to more than 25 nt length.

Photolithography



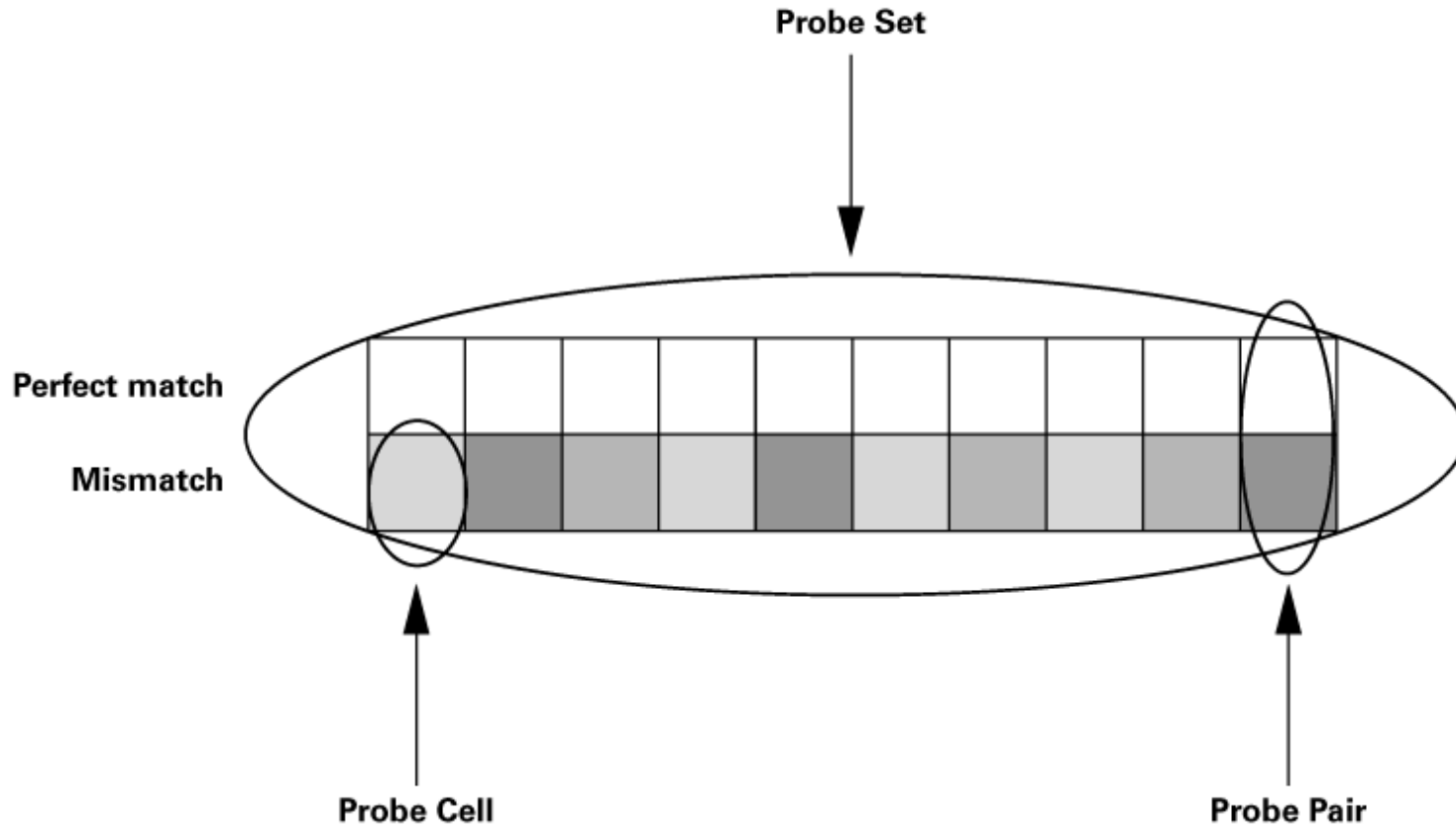
Light activated oligo synthesis



On-chip synthesis of oligonucleotides

- Oligonucleotides are built up on the chip surface by sequentially elongating the growing chain with a single nucleotide. To determine the sequence of the final oligonucleotides on each position of the chip, a process called *photolithography* is used.
- As chemical yield of the stepwise elongation is limited, oligonucleotides can't be grown to more than 25 nt length.■
- Hybridization to short oligos is quite unspecific, thus a number of them has to be used to probe for a single gene (usually 12–25).■
- Frequently, cross-hybridization occurs. To eliminate this effect, hybridization is compared with that of an oligo that bears a single mismatch.

Affymetrix probe set



cDNA vs. Oligo Chips

- long DNA strands are more specific than oligos:
 - ★ cDNA chips: 1 (2,3 identical) spots per gene
 - ★ oligo chips: many oligos per gene
- Oligo chips by on-chip synthesis:
Affymetrix GeneChip™:
 - ★ Single-color readout
 - ★ approx. 20 oligos per gene
 - ★ mismatched control for *every* oligo
 - ★ sophisticated weighting and averaging over 20 oligo pairs
 - ★ much of the information is proprietary



References

- J. Khan, M. Bittner, Y. Chen, P.S. Meltzer, and J. Trent (1999) DNA microarray technology: the anticipated impact on the study of human disease. *Biochim Biophys Acta* **1423**: M17–M28
- D.J. Duggan, M. Bittner, Y. Chen, and J.M. Trent (1999) Expression profiling using cDNA microarrays. *Nat Genet* **21** (Suppl): 10–14
- P. Hedge, R. Qi, K. Abernathy, C. Gay, S. Dharap, R. Gaspard, J.E. Hughes, E. Snesrud, N. Lee, and J. Quackenbush (2000) A concise guide to cDNA microarray analysis. *Biotechniques* **29**: 548–562
- A. Brazma and J. Vilo (2000) Gene expression data analysis. *FEBS Lett* **480**:17–24
- D.J. Lockhart and E.A. Winzeler (2000) Genomics, gene expression and DNA arrays. *Nature* **405**: 827–836
- S. Lampel and P. Lichter (2000) Nukleinsäurechips. *Medizinische Genetik* **12**: 287–289

Measures of expression

- For cDNA chips, mostly the *ratio of expressions* is used:

$$\text{ratio}_i = \frac{R_i}{G_i}$$

- The logarithm of the ratios is symmetric around ratio=1 (no change with respect to control condition):

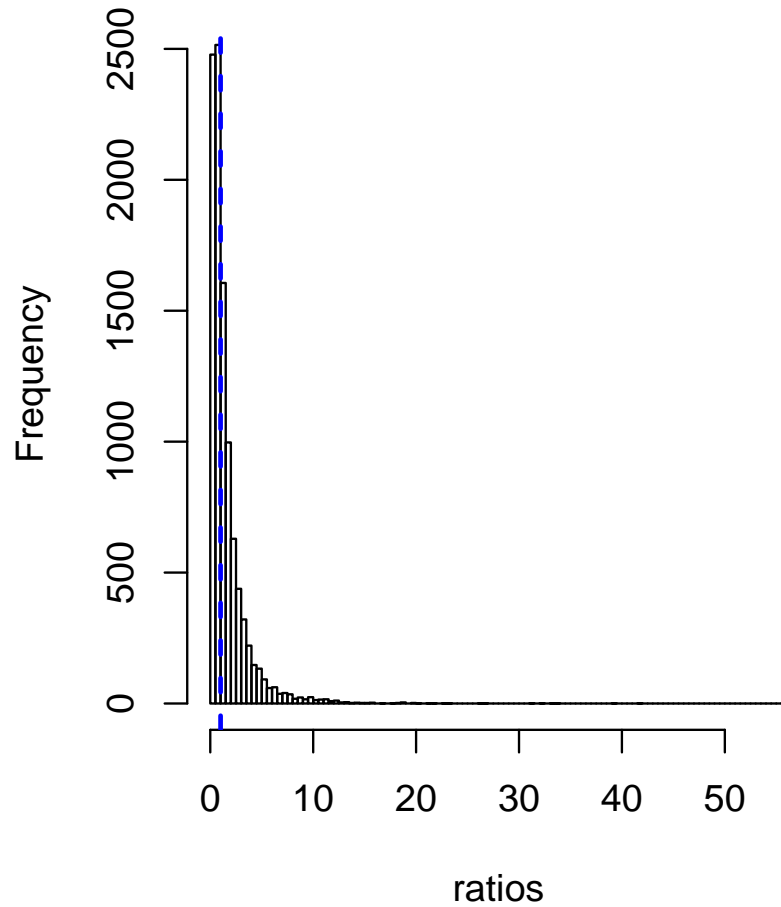
$$\text{logratio}_i = \log \frac{R_i}{G_i} = \log R_i - \log G_i$$

- Logratios to different bases of the logarithm (2, e , 10) are identical up to a constant factor:

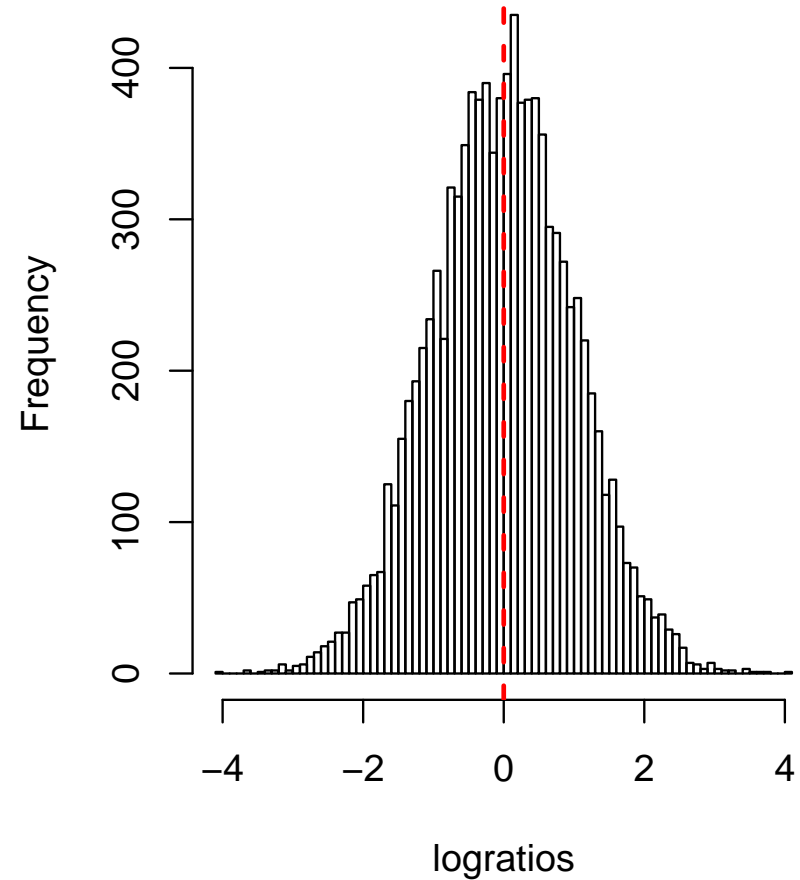
$$\log_2(x) = \log_{10}(x) \cdot \log_2(10)$$

Distribution of ratios/logratios

distribution of ratios



distribution of logratios



Measures of expression 2

- Ratios are independent of absolute signal intensity, i.e. $R_i/G_i = 20/10 = 2$ will give the same ratio as $R_i/G_i = 20,000/10,000$. Sometimes, values of M and A are used:

$$M = \log \frac{R_i}{G_i} \quad (\text{logratio})$$

$$A = 1/2 \log(R_i \cdot G_i) \quad (\text{average expression})$$

- For Affymetrix-type arrays, the signal intensities of the whole probe set have to be aggregated first. Affymetrix software (MAS) uses trimmed means:

$$\text{AvgDiff} = \frac{1}{|A|} \sum_{j \in A} (PM_j - MM_j), \quad A \subset N$$

Questions asked to microarray data: three case studies

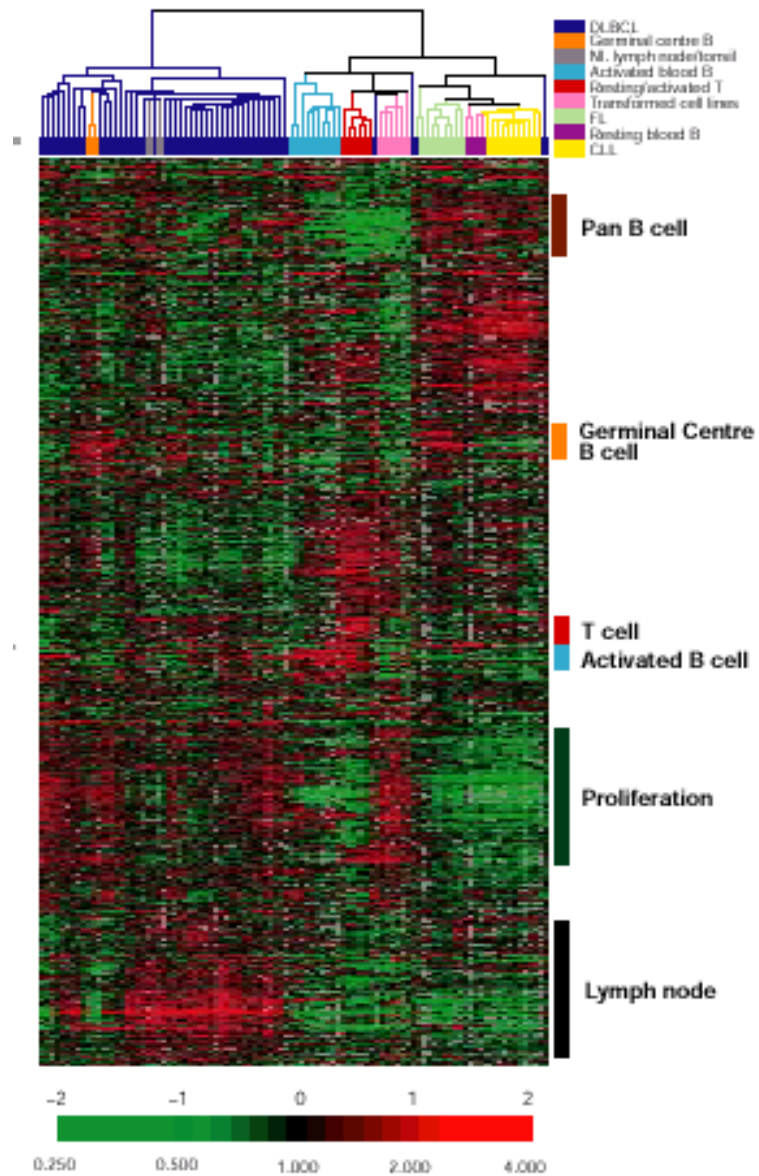
Alizadeh et al.: Lymphoma

- Study was published in *Nature* **403**:503–511 (2000)
- Gene expression profiling of Diffuse Large B-Cell Lymphoma (DLBCL)
- Lymphoma is a blood cancer where *peripheral* blood cells degenerate and divide without control
- DLBCL is an aggressive form of this disease, originating from B-lymphocytes. Overall 5-year survival is about 40%.
- Current clinical risk factors are not sufficient.

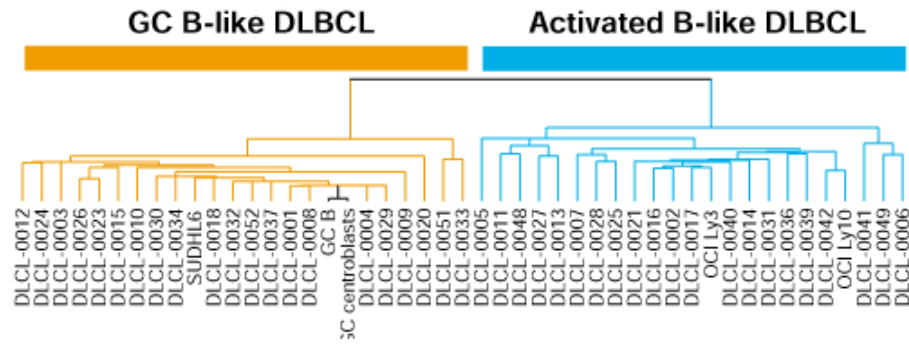
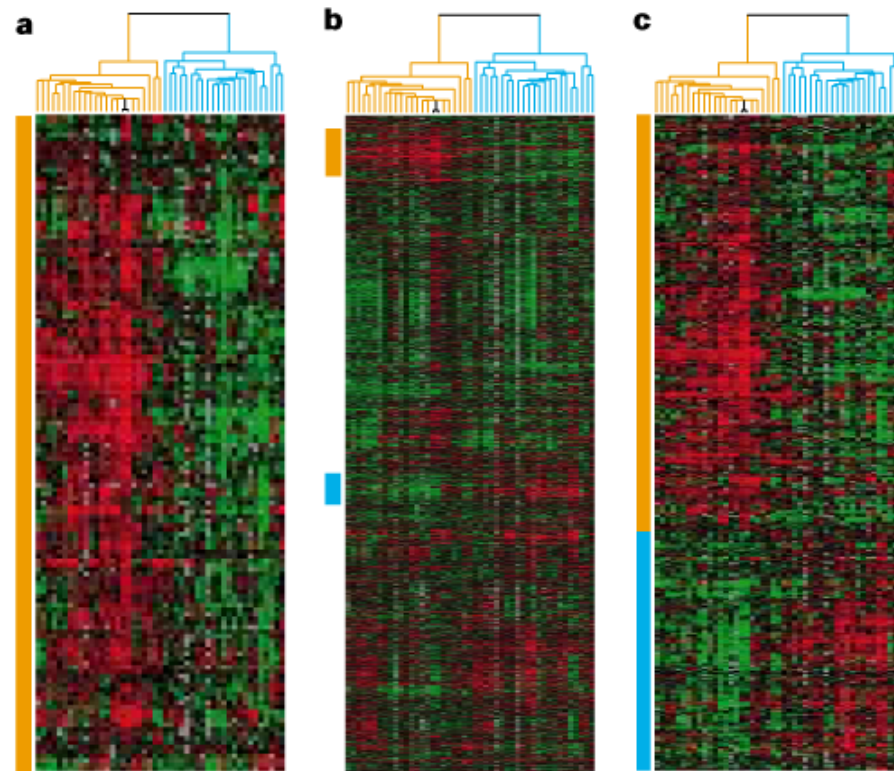
Alizadeh et al.: Methods

- A special cDNA chip was used, the *Lymphochip*
- spotted cDNA array of approximately 17,000 clones related to Lymphocytes
- 42 samples of DLBCL were analyzed, plus additional samples of normal B cells and of related diseases
- mRNA from these samples was competitively hybridized against control mRNA, stemming from a pool of lymphoma cell line mRNA preparations
- Data were analyzed by clustering

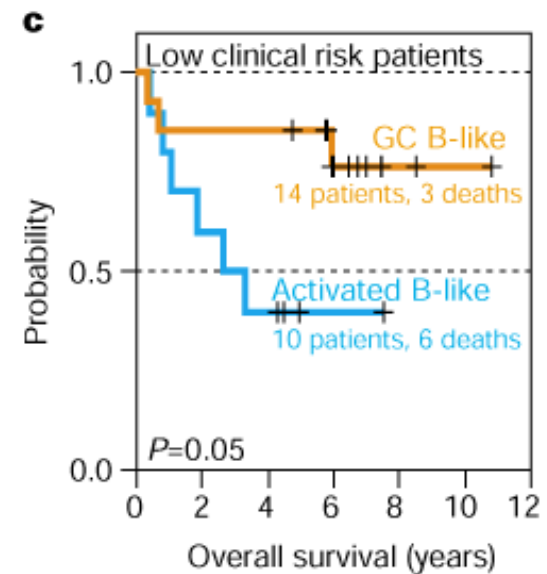
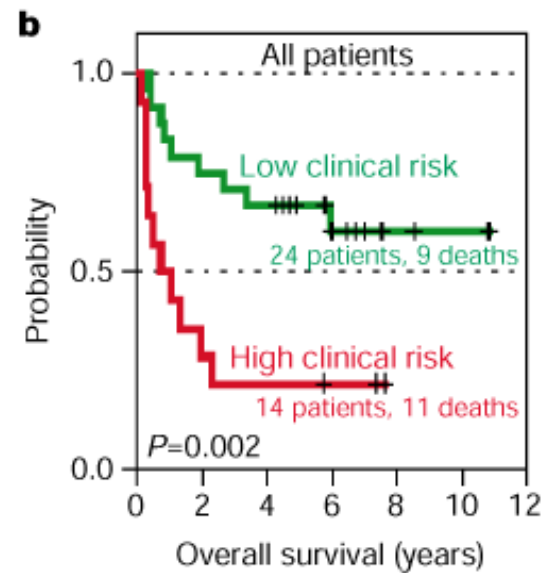
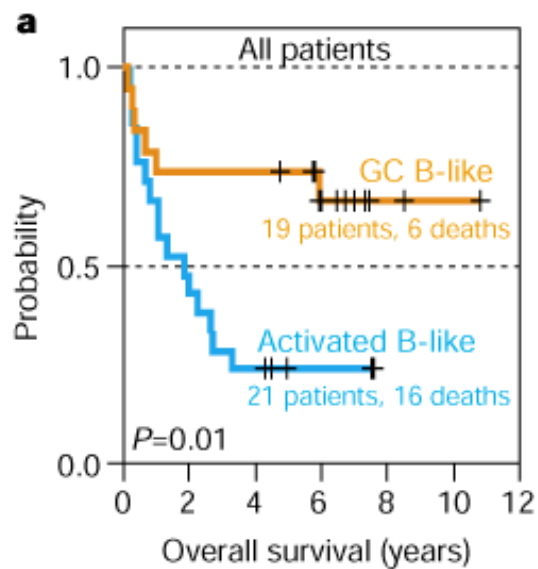
Alizadeh et al.: Results 1



Alizadeh et al.: Results 2



Alizadeh et al.: Results 3



Van't Veer et al.: Breast cancer

- published in *Nature* **415**:530–536 (2002)
- looks for prognostic markers in breast cancer
- two classes of patients: those with distant metastasis (other than in breast) within 5 years, and those without (also had negative lymph node status)
- In statistical thinking, this is a *classification* problem: given a set of *variables*, can we train a *classifier* such that it predicts for any new sample the *class* as correctly as possible?

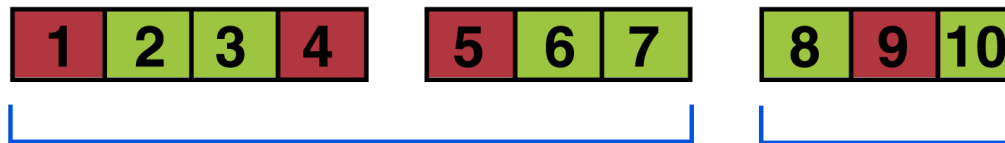
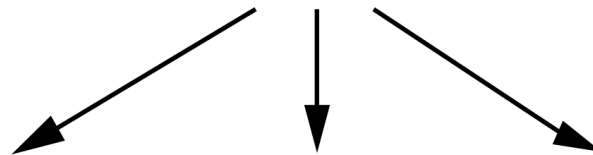
Van't Veer et al.: Methods

- A custom-made 25,000-clone chip was used; each feature contained a unique 60-mer oligonucleotide. This oligo was transferred to the chip by ink-jet like printing.
- The chips were hybridized competitively; the reference mRNA was obtained from a pool of patient mRNA (98 patients in total).
- Only data from certain genes (231) were used; finding out informative genes is called *feature selection* in machine learning.
- A home-made *ad hoc* classification method was used (no details given here). You can do better with established classification methods (taught later in this course).
- The model was validated by cross validation and by an independent test set.

Cross-validation

class A

class B



Training set

Test set

run 1



Test set

Training set

run 2



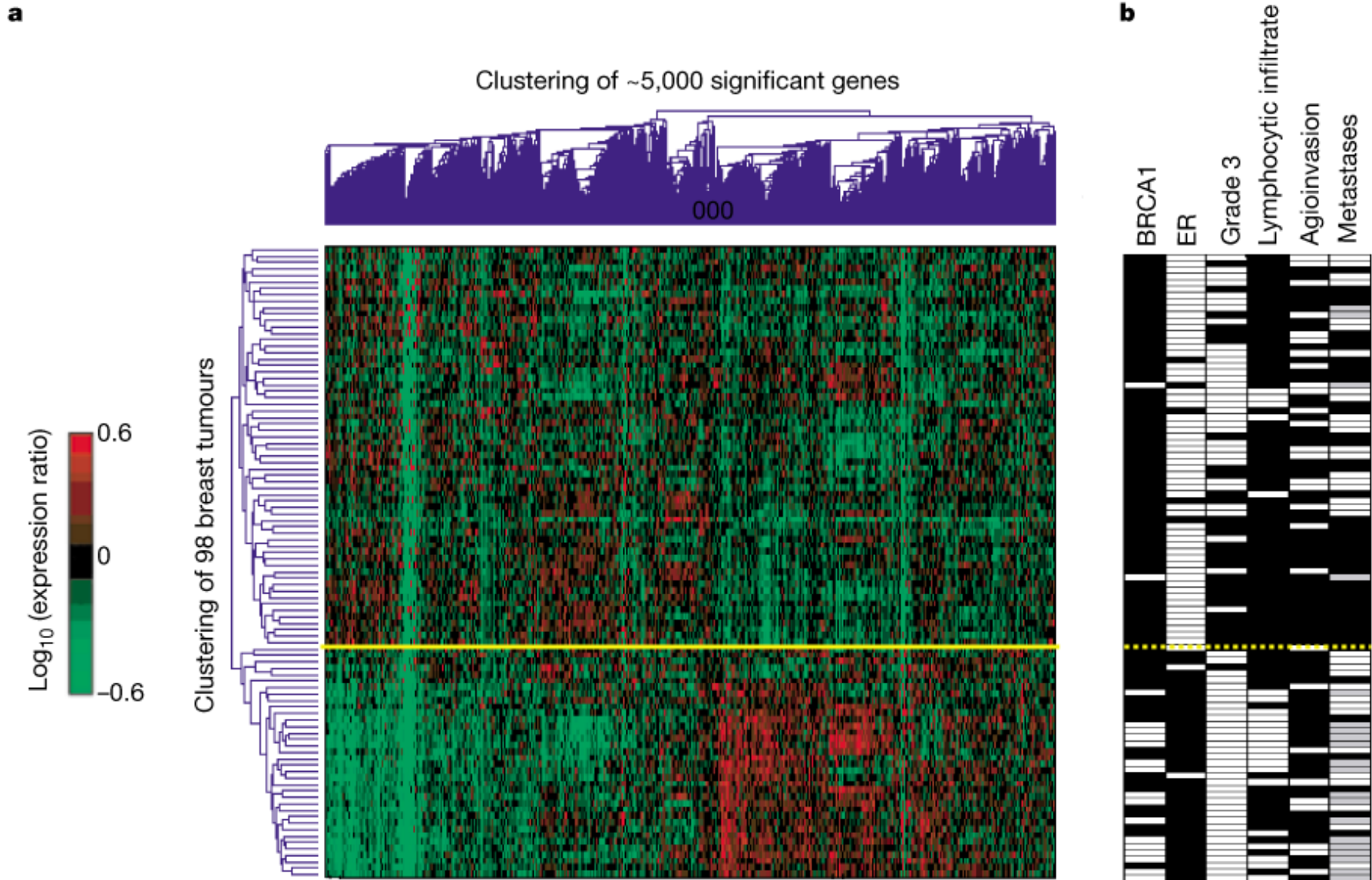
Training ...

Test set

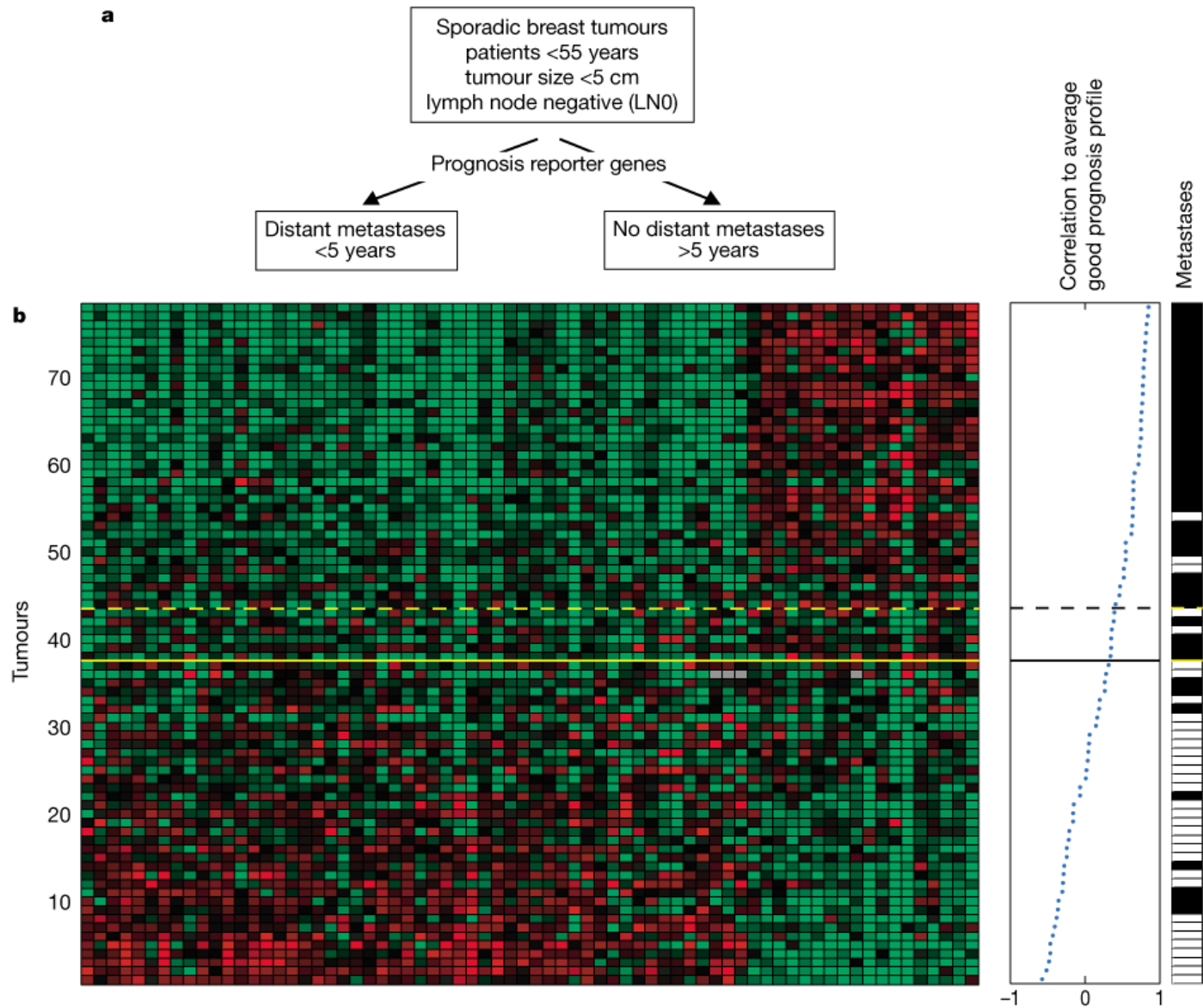
... set

run 3

Van't Veer et al.: Results 1



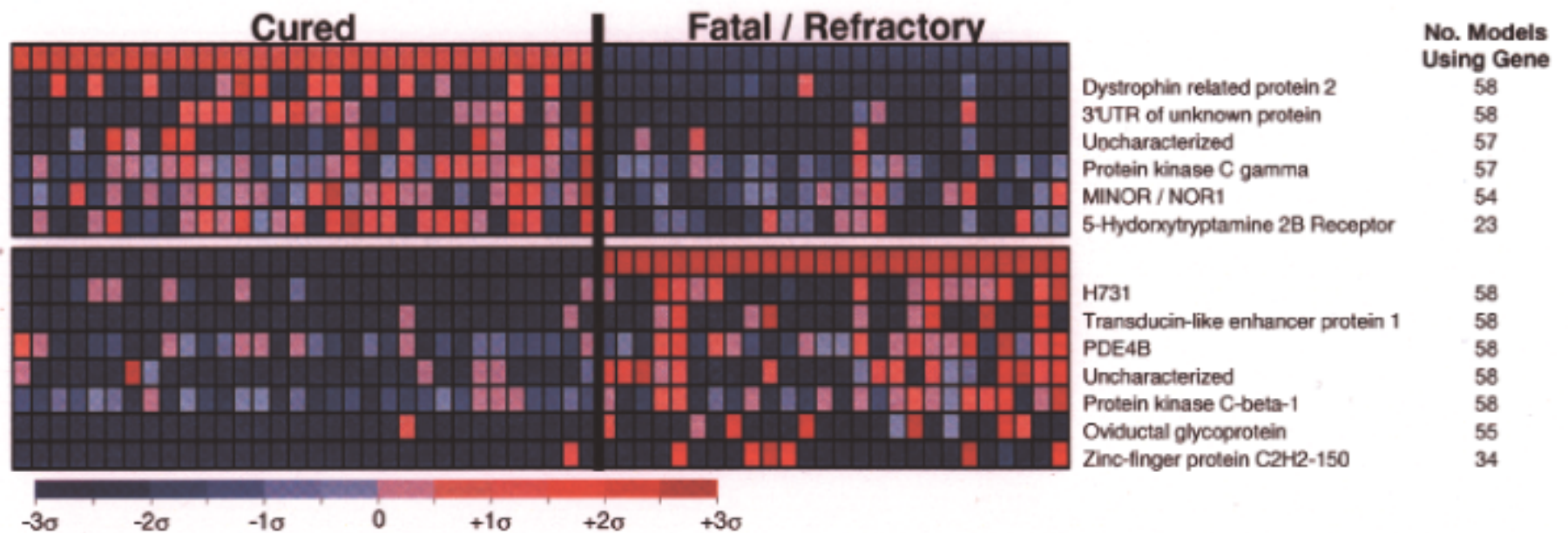
Van't Veer et al.: Results 2



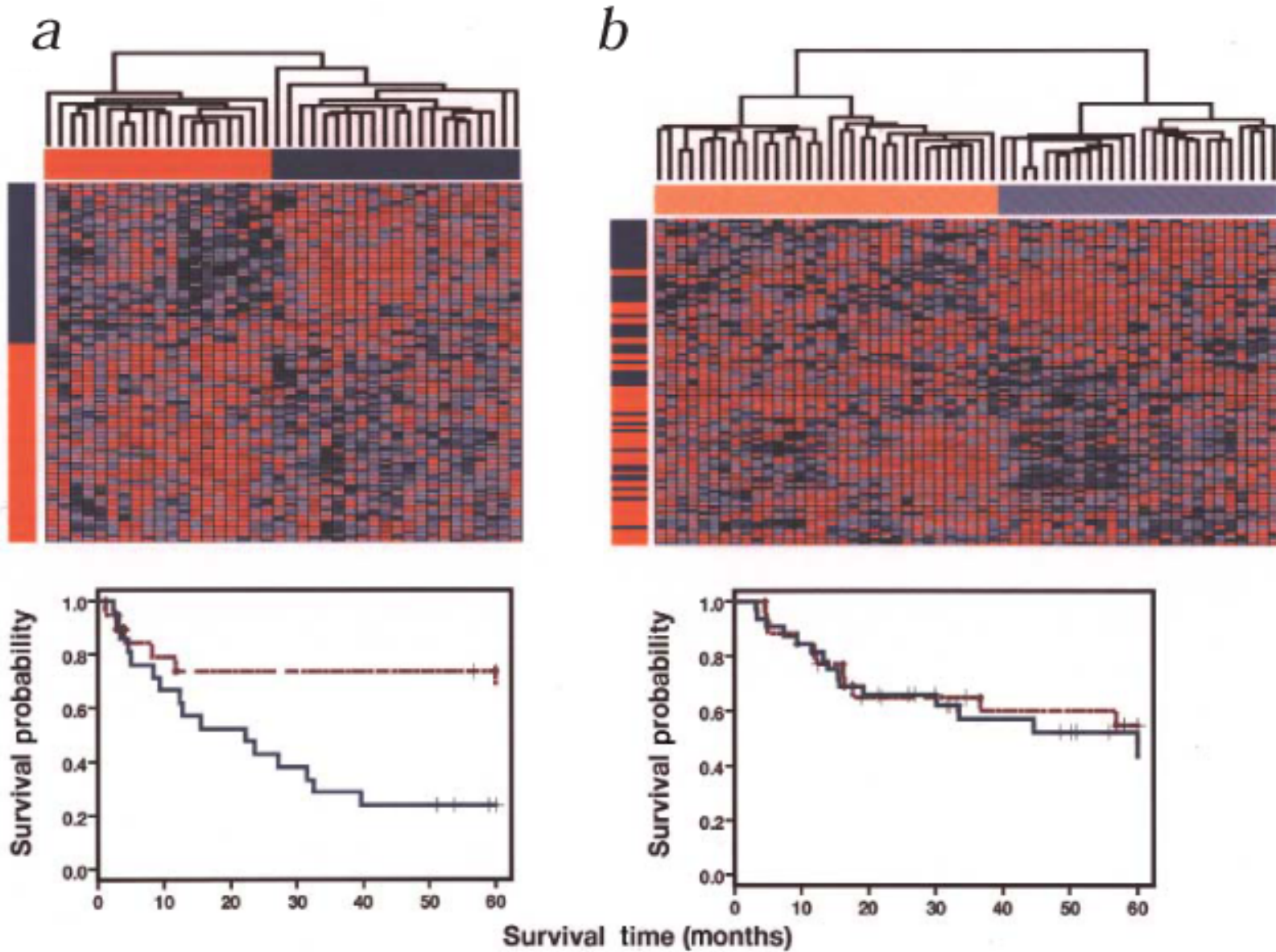
Shipp et al.: Lymphoma again

- published in *Nat. Med.* 8:68–74
- Same lymphoma (DLBCL) as in the study of Alizadeh et al. was investigated
- Samples from 58 patients with DLBCL were subjected to gene expression analysis
- Affymetrix chip was used (6,800 probe sets)
- A classification (supervised) approach was taken
- Results were compared with those of Alizadeh et al.

Shipp et al.: Results 1



Shipp et al.: Results2



Possible extension: Regression

- This was treated as a classification problem, i.e. there were distinct *classes* (cured vs. fatal) as *response variables*
- One could also use a *continuous* response variable: e.g. survival time, or the probability of being cured
- Fitting a model that predicts a continuous response is called *regression* in statistics (methods to be discussed later)