

Normal human variation – We don't know remotely enough about the nature and origins of normal variation in “biomarkers” and virtually every important parameter for early disease detection.

Pre-clinical “occult” natural history of cancer – we don't know what we need to detect to make a difference.

Rigorous comprehensive description of source of samples – Sloppy standards. Can't trust results without this.

Sample divisibility – trade-offs between use and preservation of samples.

Sample handling and processing is an important source of extrinsic/confounding variation, but...

**Most** of the variation, especially the variation that can introduce unrecognized bias, is due to factors that act **before** sample collection – “Biology” (much larger parameter space, harder or impossible to enforce control).

Illustrative example....

# Diagnostic Markers for Early Detection of Ovarian Cancer

Irene Visintin,<sup>1</sup> Ziding Feng,<sup>2</sup> Gary Longton,<sup>2</sup> David C. Ward,<sup>3</sup> Ayesha B. Alvero,<sup>1</sup> Yinglei Lai,<sup>4</sup> Jeannette Tenthorey,<sup>1</sup> Aliza Leiser,<sup>1</sup> Ruben Flores-Saaib,<sup>5</sup> Herbert Yu,<sup>6</sup> Masoud Azori,<sup>1</sup> Thomas Rutherford,<sup>1</sup> Peter E. Schwartz,<sup>1</sup> and Gil Mor<sup>1</sup>

**Abstract** **Purpose:** Early detection would significantly decrease the mortality rate of ovarian cancer. In this study, we characterize and validate the combination of six serum biomarkers that discriminate between disease-free and ovarian cancer patients with high efficiency.

**Conclusions: We describe the first blood biomarker test with a sensitivity of 95.3% and a specificity of 99.4% for the detection of ovarian cancer. This novel multiplex platform has the potential for efficient screening in patients who are at high risk for ovarian cancer.**

## **Diagnostic Markers for Early Detection of Ovarian Cancer**

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### **Sample collection**

Ten mL of peripheral blood was drawn from subjects using standardized phlebotomy procedures (11). Within 2 to 4 hours of collection, samples were processed using guidelines set by the National Cancer Institute Inter-Group Specimen Banking Committee and stored at -80°C in the Tissue/Sera Bank of the Discovery to Cure program.

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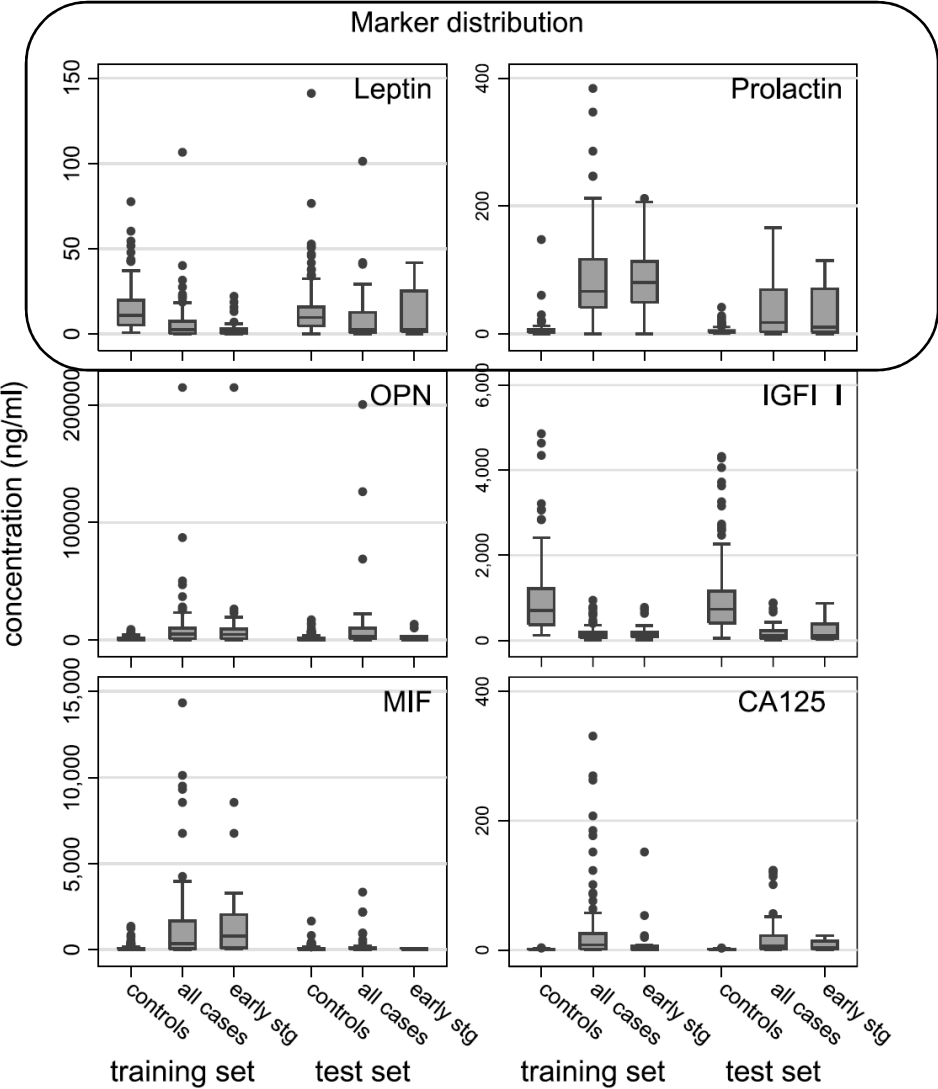
**Abstract** **Purpose:** Early detection would significantly decrease the mortality rate of ovarian cancer. In this study, we characterize and validate the combination of six serum biomarkers that discriminate between disease-free and ovarian cancer patients with high efficiency.

*Ovarian cancer group.* The disease group ( $n = 156$ ) includes women with newly diagnosed ovarian cancer (pelvic mass). All samples were collected previous to diagnosis at the gynecologic oncology clinic. Of

*Control group.* The healthy control group ( $n = 362$ ) included age-matched healthy individuals who came for a regular gynecologic examination. These individuals did not have a diagnosis of any type of cancer, were not genetically predisposed to develop ovarian cancer, and were disease free at least 6 months after sample collection. A total of

# Diagnostic Markers for Early Detection of Ovarian Cancer

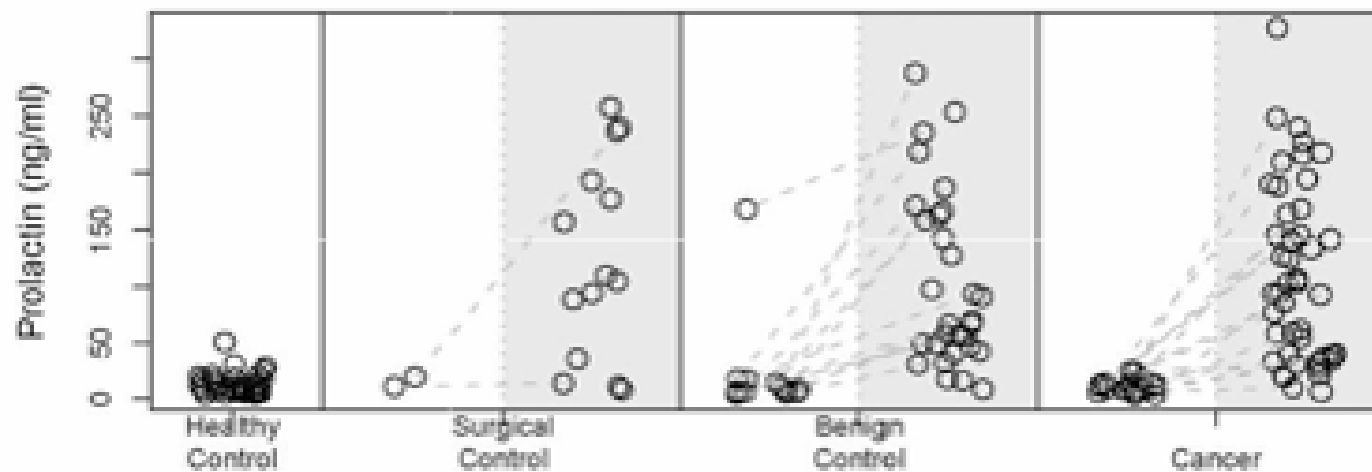
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# Effects of Blood Collection Conditions on Ovarian Cancer Serum Markers

Jason D. Thorpe<sup>1\*</sup>, Xiaobo Duan<sup>2</sup>, Robin Forrest<sup>1</sup>, Kimberly Lowe<sup>1</sup>, Lauren Brown<sup>1</sup>, Elliot Segal<sup>2</sup>, Brad Nelson<sup>3</sup>, Garnet L. Anderson<sup>4</sup>, Martin McIntosh<sup>1</sup>, Nicole Urban<sup>1</sup>

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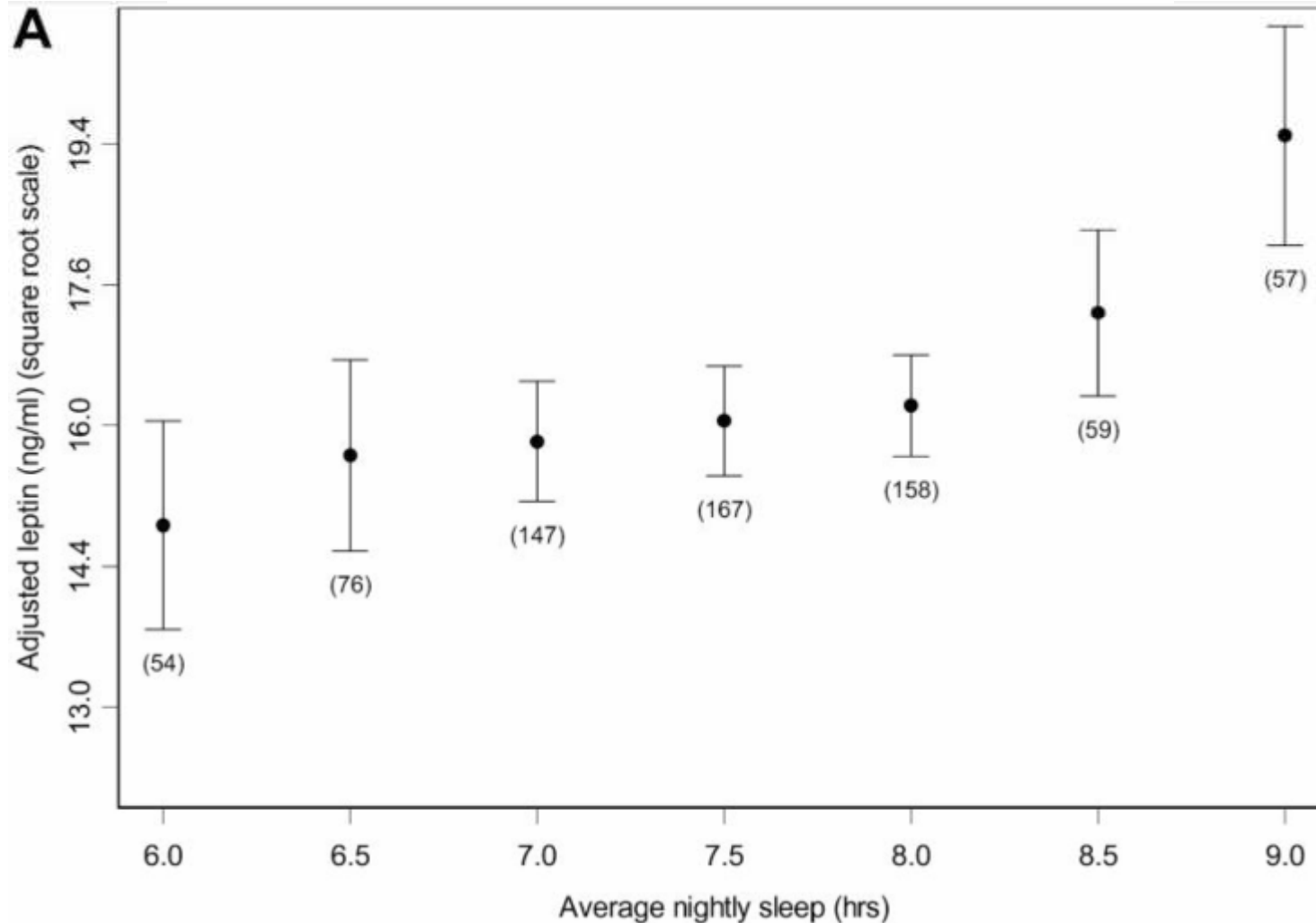
Prolactin	Blood Collection Conditions				
	At Clinic	(Reference)			
	In Surgery	93.23	8.82	<0.005	
Case/Control Group					
	Healthy Control	(Reference)			
	Surgical Control	15.23	20.28	0.45	
	Benign Control	0.45	11.49	0.97	
	Ovarian Cancer	2.37	5.97	0.69	

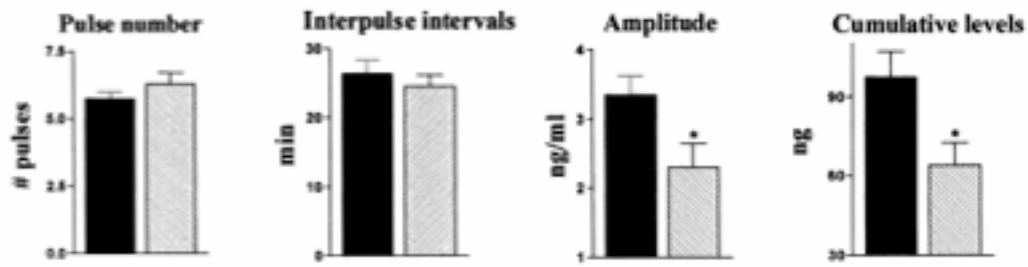
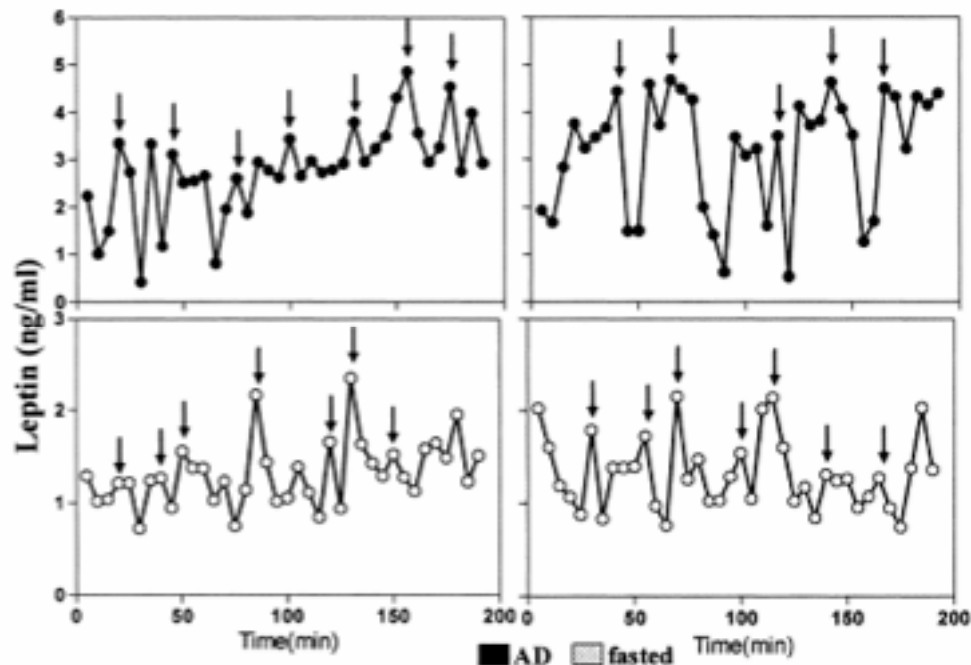


# Short Sleep Duration Is Associated with Reduced Leptin, Elevated Ghrelin, and Increased Body Mass Index

Shahrad Taheri<sup>1□</sup>, Ling Lin<sup>1</sup>, Diane Austin<sup>2</sup>, Terry Young<sup>2</sup>, Emmanuel Mignot<sup>1\*</sup>

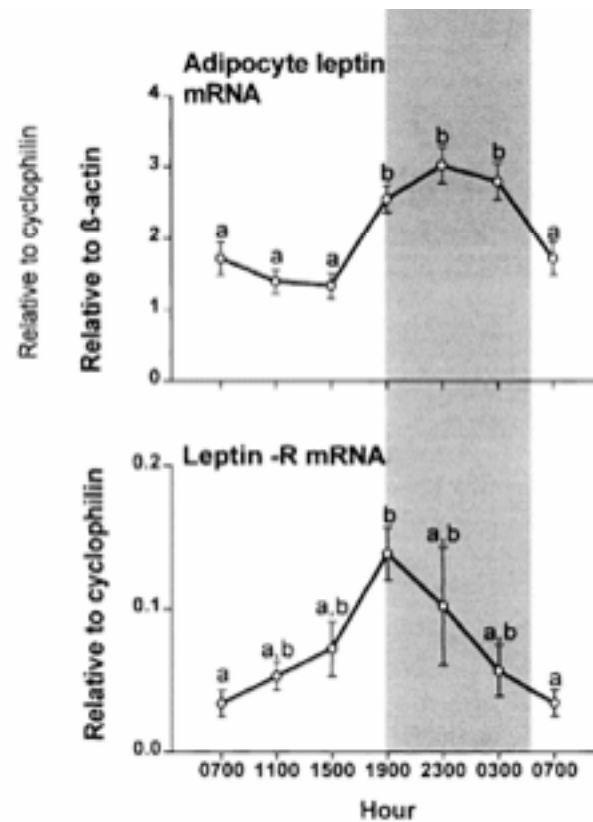
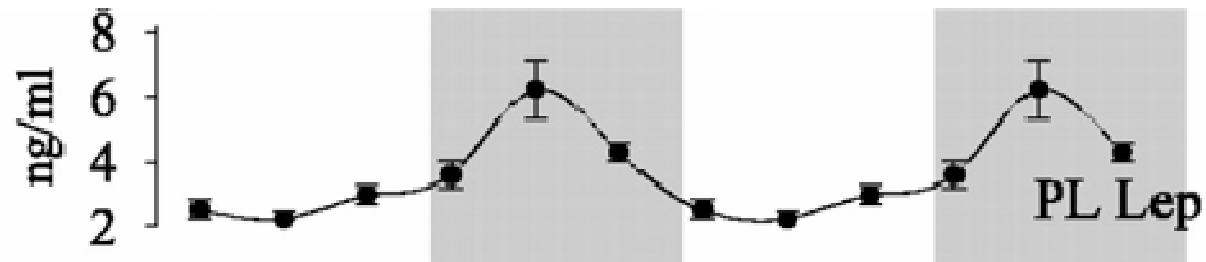
**1** Howard Hughes Medical Institute, Stanford University, Palo Alto, California, United States of America, **2** Department of Population Health Sciences, University of Wisconsin, Madison, Wisconsin, United States of America

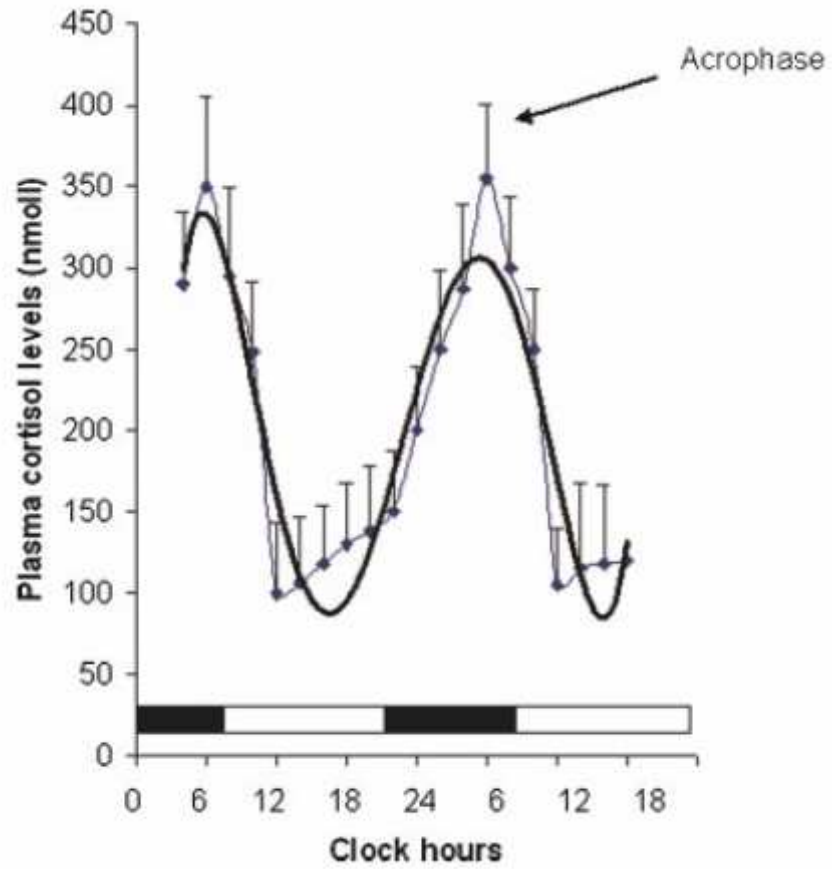




Fasting/eating  
dependent,  
ultradian  
variation in  
serum leptin

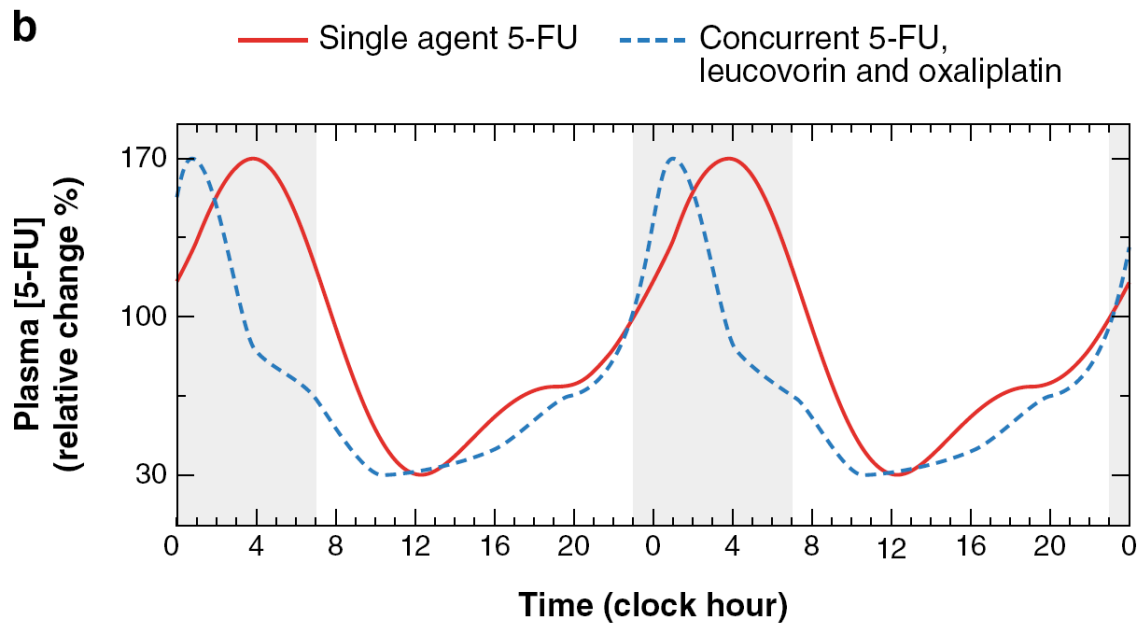
Circadian variation in plasma leptin and peripheral expression



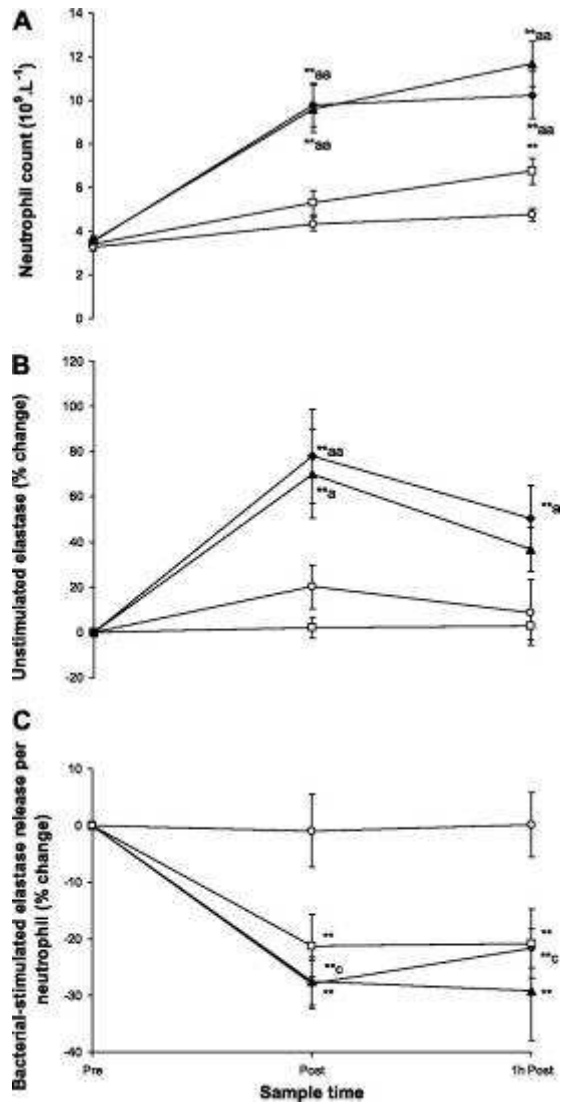


Circadian  
variation in  
serum cortisol

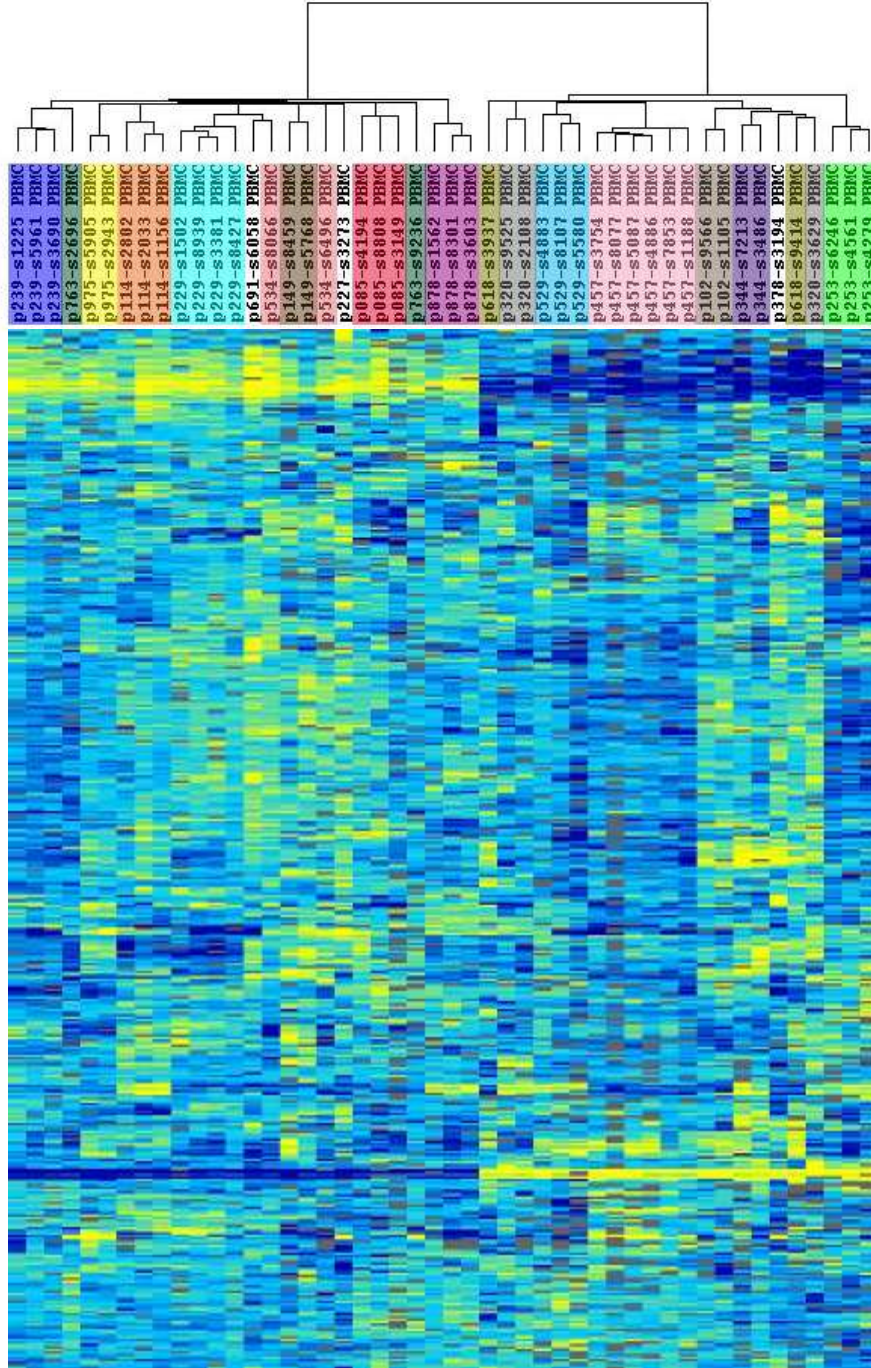
# Circadian variation in 5FU metabolism



# Neutrophil counts and exercise

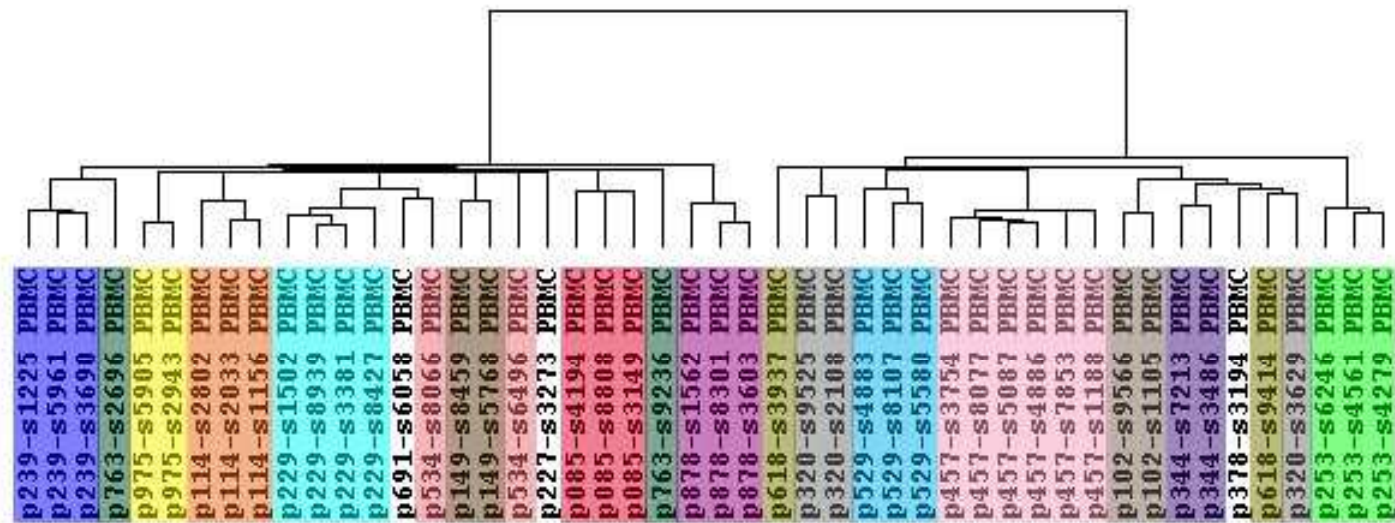


Ca. 3000 genes



Multidimensional variation in gene expression patterns observed in peripheral blood samples from "normal" volunteers

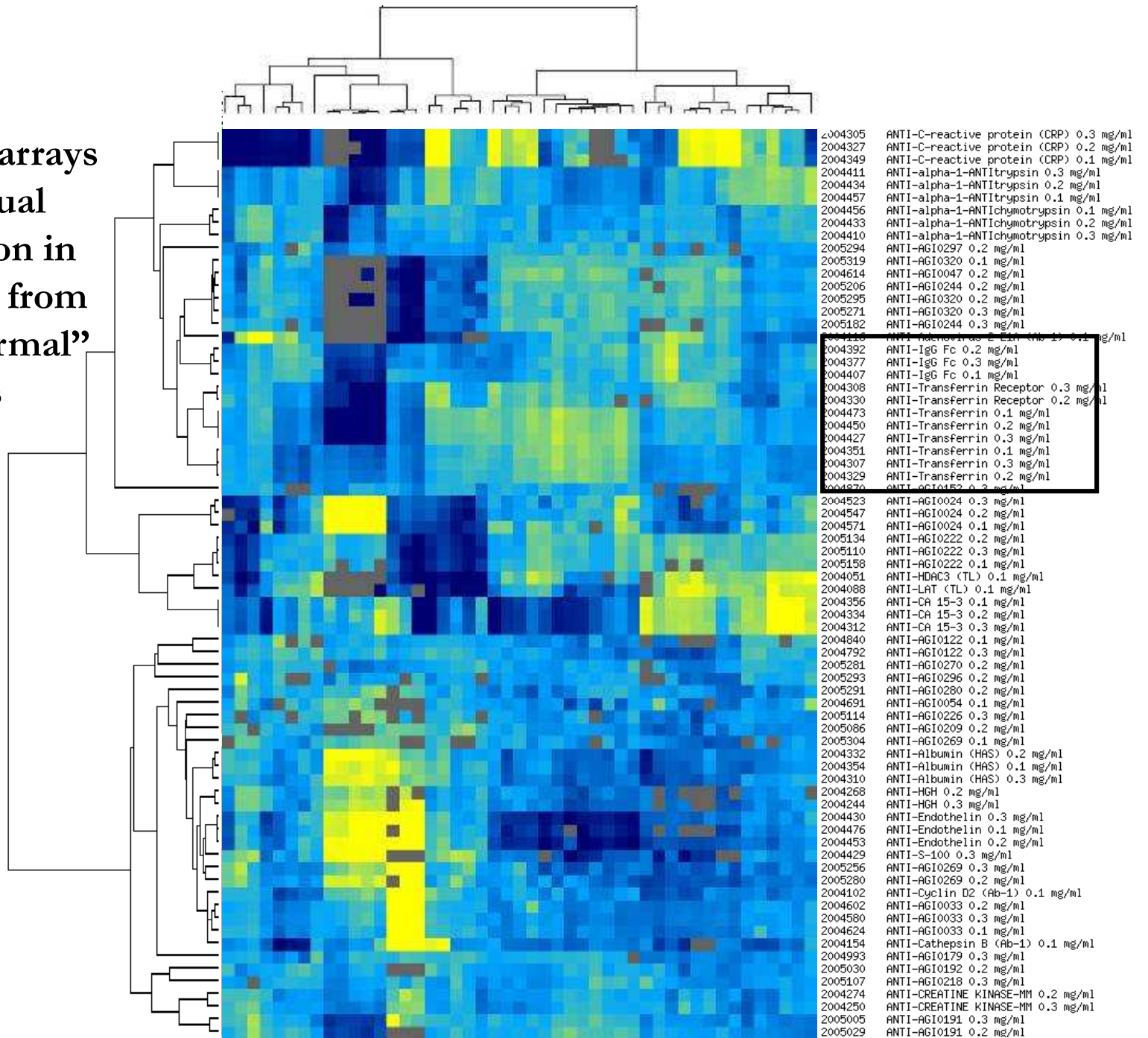
Addie Whitney  
Kate Rubins  
Jen Boldrick  
Max Diehn  
Ash Alizadeh  
David Relman



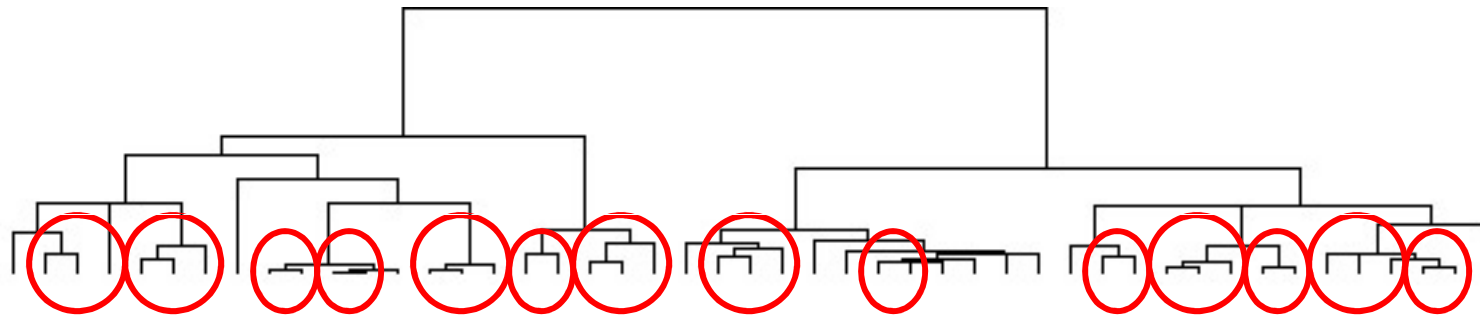
Distinctive, individual-specific features of gene expression patterns in peripheral blood cells persist over at least weeks



Antibody microarrays  
inter-individual  
protein variation in  
serum samples from  
“apparently normal”  
volunteers



**Most replicate serum samples  
cluster together as nearest neighbors**



Important potential confounding factors for any molecular marker include:

Interindividual genetic variation  
Intercurrent illness or physiological stress  
Psychological stress  
Sleep  
Nutrition  
Medication  
Time of day  
?

We lack a basic interpretative framework for molecular (and anatomic/histological) variation.

For any observation that might be interpreted as evidence of disease, what are all the factors that can influence it. (Differential diagnosis of molecular/anatomic/histological variation – in quantitative, probabilistic terms).

Disease is defined as a distinct deviation from the range of normal variation and detection and diagnosis of disease implicitly depends on knowing the scope and boundaries of “normal” variation. **Knowing what “normal” can look like is the foundation of medical diagnosis.**

But....

We know astonishingly little about the “normal human”

A **miniscule** fraction of molecular studies (eg. gene expression) have been devoted to defining what normal cells, tissues and fluids can look like, and how the variation relates to genetic, environmental and physiological factors.

We need a systematic characterization of normal human phenotypic variation

Anatomy

Histology

Expression patterns

Molecular profiles of cells, tissues, fluids

Microbial flora

-links to genotype and environmental factors.

## Biospecimens

We need to launch a deliberate thoughtful attempt to collect tissues, fluids and associated data from the “general population”.

On a **very large** scale – need to map out (rigorously and quantitatively) the gamut of variation.

Needs to be large scale because a lot of what appears to be pathological deviation may be outer limits of normal – but we don’t know. Knowing the rare non-pathological variants is critical for screening and early detection of low-prevalence disease (most cancers) and diagnostic interpretation of apparent anomalies.

# Autopsies!

Systematic, concerted data collection, not small series and case reports. High resolution imaging, meticulous histopathology.

How prevalent are occult neoplasms or other pathological variations in the population?

Autopsies of “normal” people are a tall order.

?Transplant donors

?Medical examiners

Not easy – but necessary! Need to be  
resourceful



## **Pre-clinical natural history of cancer**

Critical for rational early detection – Duh! ‘

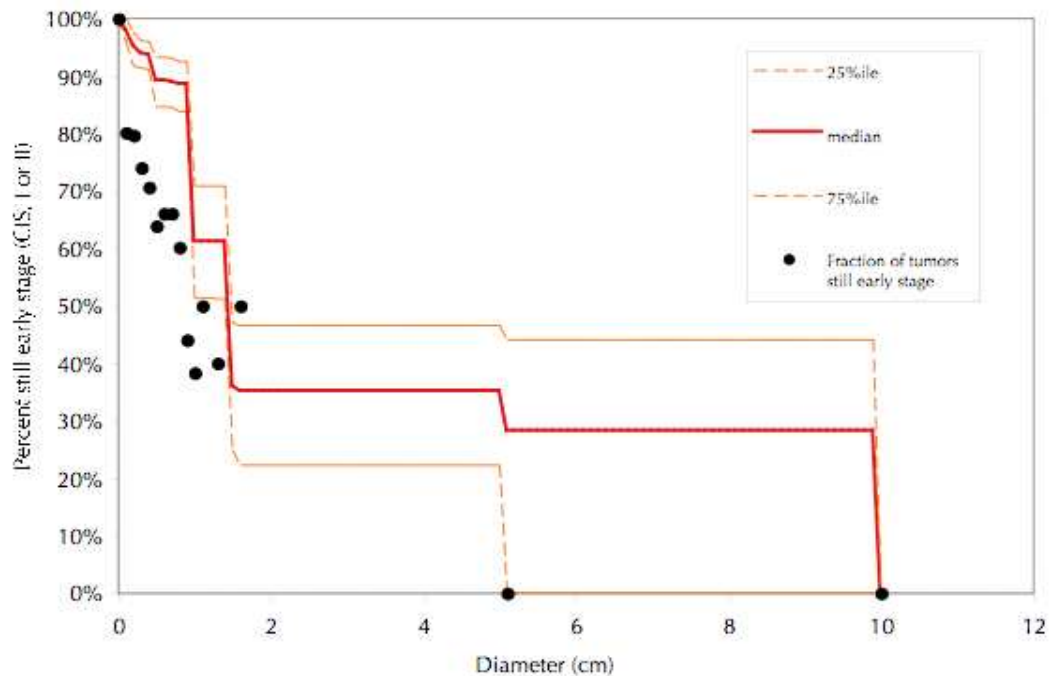
We need to know what we need to detect to make a difference!

Very difficult challenge, needs to be a priority.

Prevalent assumptions are commonly unfounded and very likely wrong.

Example: Serous ovarian cancers progress to advanced stage at a size 1000 times smaller than the median clinical early stage ovarian cancer

Kaplan-Meier analysis of serous "ovarian" cancer progression to Stage III or IV as a function of tumor size



Palmer and Brown, unpublished

## **Balancing use and preservation of precious biospecimens.**

Sample divisibility independent of prior aliquoting design  
(eg., sliceable, resealable tubes of frozen fluids).

Research article

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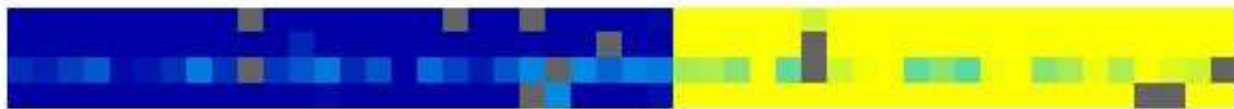
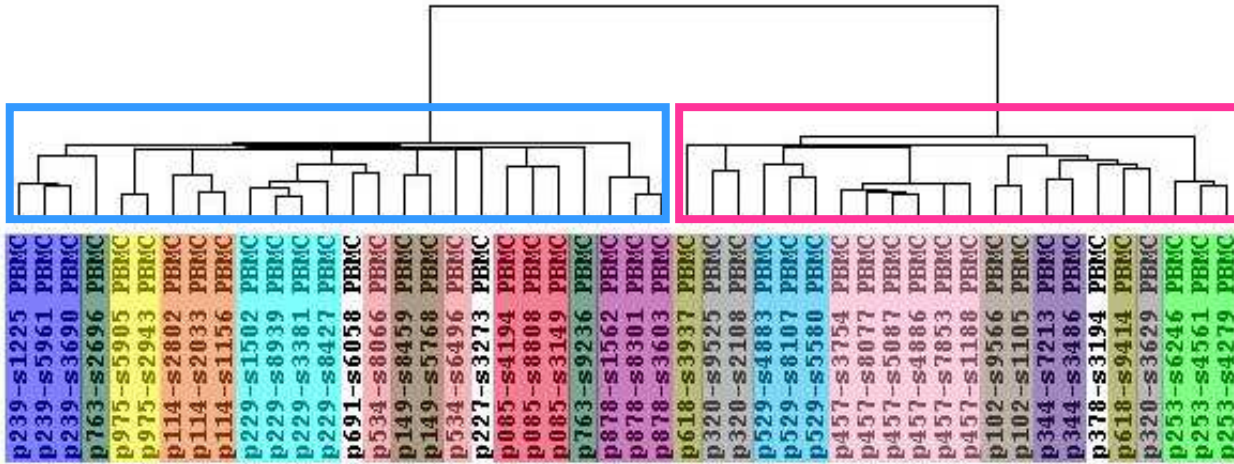
## **Early detection of breast cancer based on gene-expression patterns in peripheral blood cells**

Praveen Sharma<sup>1</sup>, Narinder S Sahni<sup>1</sup>, Robert Tibshirani<sup>2</sup>, Per Skaane<sup>3</sup>, Petter Urdal<sup>4</sup>, Hege Berghagen<sup>1</sup>, Marianne Jensen<sup>1</sup>, Lena Kristiansen<sup>1</sup>, Cecilie Moen<sup>1</sup>, Pradeep Sharma<sup>1</sup>, Alia Zaka<sup>1</sup>, Jarle Arnes<sup>5</sup>, Torill Sauer<sup>6</sup>, Lars A Akslen<sup>5</sup>, Ellen Schlichting<sup>7</sup>, Anne-Lise Børresen-Dale<sup>8</sup> and Anders Lønneborg<sup>1</sup>

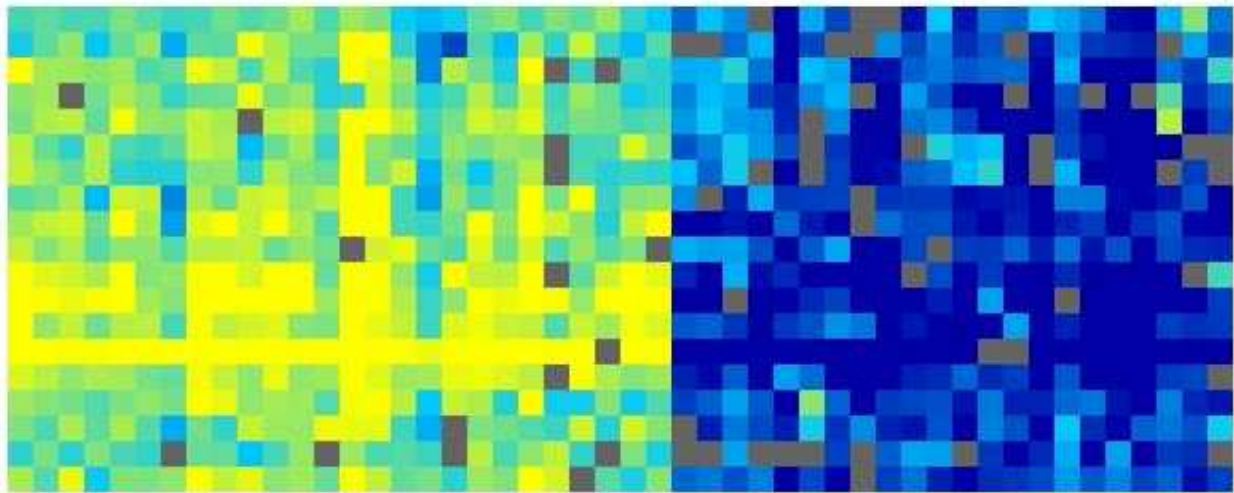
<sup>1</sup>DiaGenic ASA, Oslo, Norway



# Sex-specific



XIST



DKFZP434I143  
 hypothetical prot  
 Hs.205080 ESTs  
 Hs.105516 EST  
 Homo sapiens linc  
 Homo sapiens linc  
 Homo sapiens mRNA  
 EIF1AY  
 eukarvotic transl  
 Homo sapiens mRNA  
 PAH  
 phenylalanine hvd  
 Human DNA sequenc  
 RPS4Y  
 ribosomal protein  
 SMCY  
 SMC (mouse) homol  
 RPS4Y  
 ribosomal protein  
 UTY  
 ubiquitously tran  
 EIF1AY  
 eukarvotic transl  
 Homo sapiens linc  
 USP9Y  
 #ubiquitin specif



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## Epidemiology of the Insulin-like Growth Factor System in Three Ethnic Groups

J. K. Cruickshank,<sup>1</sup> A. H. Heald,<sup>2</sup> S. Anderson,<sup>1</sup> J. E. Cade,<sup>1</sup> J. Sampayo,<sup>2</sup> L. K. Riste,<sup>1</sup> A. Greenhalgh,<sup>1</sup> W. Taylor,<sup>3</sup> W. Fraser,<sup>3</sup> A. White,<sup>2</sup> and J. M. Gibson<sup>2</sup>

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IGF-II showed a strong inverse association with African-Caribbean ( $\beta = -0.264$ ,  $p = 0.001$ ) and Pakistani ( $\beta = -0.240$ ,  $p < 0.0001$ ) ethnicity compared with European ethnicity