

# Candidate p1RCC biomarkers and environmental factors influencing their expression levels

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#### **Problem Definition**

- Papillary renal-cell carcinoma (pRCC), comprises 15 20% of all kidney cancers. It occurs in the cells lining the small tubules in the kidney that filter waste from the blood and make urine.
- pRCC has two subtypes (p1RCC & p2RCC) based on histologic, cytogenetic, and gene expression differences.
- Little is known about the genetic basis of sporadic papillary renal-cell carcinoma, and no effective forms of therapy for advanced disease exist.
- The goal of this analysis is to find candidate diagnostic biomarkers and treatment regimen for p1RCC.

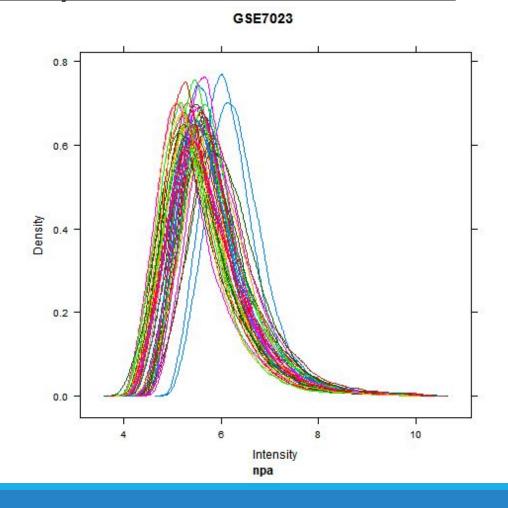
# Microarray gene expression data was downloaded for p1RCC and normal kidney tissue

- Data source: NCBI GEO database.
  <a href="https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE7023">https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE7023</a>
- Platform: Affymetrix GeneChip Human Genome U133 Plus 2.0 Array
- 46 cases
  - 19 p1RCC
  - 12 Normal Tissue
  - (also 16 p2RCC samples that were not used)
- 54,675 probes
- Target: Classify Normal Tissue (0) vs. p1RCC (1)

## Initial view using no normalization or background correction shows data consistency

Raw data (CEL files) from 47 microarrays were processed with:

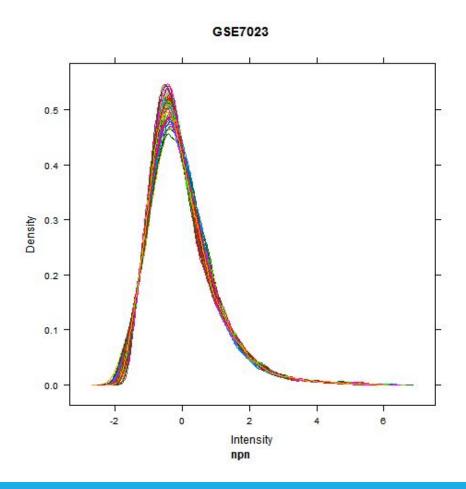
 Only using perfect match (PM) probes with median summarization



# Using normalization within each array but not between arrays

Raw data (CEL files) from all 47 microarrays were processed with:

- No background correction
- Only perfect match (PM)
- Median summarization
- Within-array normalization
- No between-array normalization

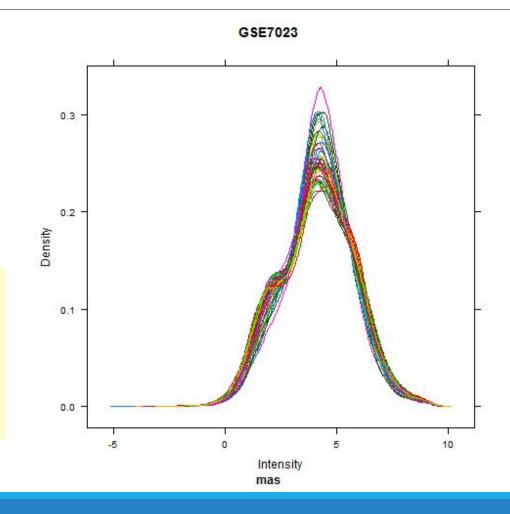


## Adding between-array normalization (MASS method) was not as useful

Raw data (CEL files) from 47 microarrays were processed and **normalized** using the **MAS5** method.

Using MAS5 or any between-array normalization methods can introduce artifacts. Please see the supplementary document:

A Case Against Microarray Data Normalization.pdf



### For meaningful predictive modeling, we selected features capturing the most variation between p1RCC and normal tissue

- Predictive modeling using 54,675 probes is not practical and even counterproductive (due to large number of collinear probes).
- Subsets of genes capturing the most variation can be located using Mahalanobis distance (in standard deviations), t-test and ANOVA.
- In each of 10 cross-validation passes of 80% data was randomly assigned to a trainset and 20% to a testset (before any statistical analysis or predictive modeling).
- For each trainset, top 100 probes/genes showing the most variation in expression with the test set were used to find the most predictive candidate biomarkers.

# Reliability Test: Ratio of overlap should not be affected by different pre-processing methods

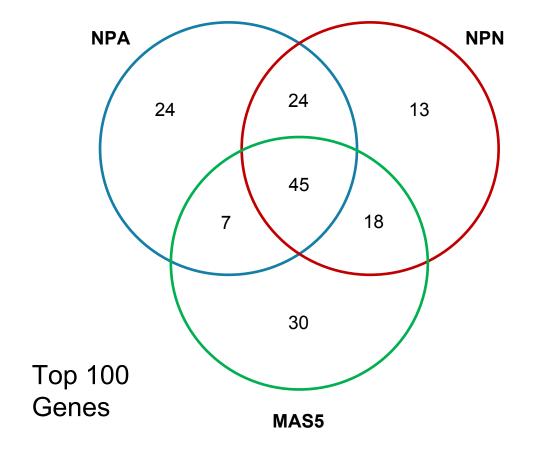
There is a relatively significant overlap between Top 100 genes from three different preprocessing methods.

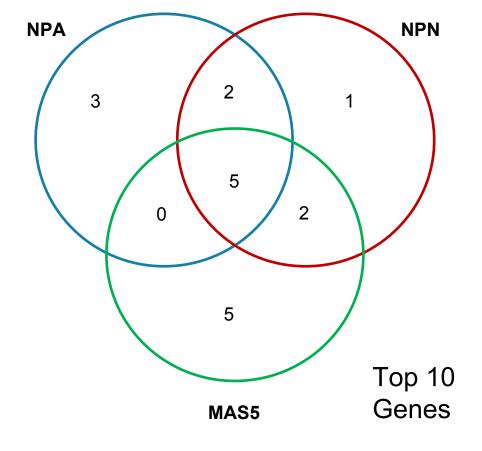
When we use Top 10 genes the ratio of overlap does not change between NPA-NPN and MAS5-NPA but it drops between MAS5-NPN.

	MAS5	NPA	NPN
MAS5	100	52	63
NPA	52	100	69
NPN	63	69	100

	MAS5	NPA	NPN
MAS5	10	5	3
NPA	5	10	7
NPN	3	7	10

# Reliability Test: Overlap ratio should not be affected by different pre-processing methods





Top predictive genes and selected pathways offer insight on hypothetical biomarker candidates (Comparing p1RCC and Normal Tissue)

NPA	NPN	MAS5
KNG1	KNG1	TMPRSS2
UMOD	UMOD	KNG1
KCNJ1	PTGER3	RALYL
HS6ST2	TFAP2B	FHL1
FHL1	KCNJ1	MUC15
PTGER3	HS6ST2	DMRT2
DMRT2	DMRT2	LRRK2
HSD11B2	KCNJ10	IRX2
EMCN	EMCN	PTGER3
LRRK2	FHL1	UMOD

# Can annotations on our top predictive genes offer insights on Bill's history?

Gene	Name	Pathway/Notes
KNG1	kininogen 1	Complement and coagulation cascades
PTGER3	prostaglandin E receptor 3	Calcium signaling pathway
DMRT2	doublesex and mab-3 related transcription factor 2	Sequence-specific DNA binding
UMOD	uromodulin	Most abundant protein in normal urine
FHL1	Four And A Half LIM Domains 1	Tumor suppressor gene on X chromosome. Points to JAK-STAT pathway

# Modeling (Normal Tissue & p1RCC)

- Binary Classification
  - Linear Discriminant Analysis
  - XGBoost
  - Both provided 100% accuracy with 10-fold cross-validation

- KNG1 provides widest discrimination gap between normal and p1RCC tissue with the highest predictability power.
- Other genes (like PTGER3, UMOD, DMRT2) also show 100% accuracy.

# Could decreased KNG1 expression offer an insight into Bill's medical history?

1. Got warfarin/coumadin for diagnosis of deep vein thrombosis.

. . .

2. Symptoms returned. Went back & found:

A. 7 cm mass left kidney

B. Cerebral meningioma

C. Spots in lung

. . . .

Meningioma hasn't grown.

Chest spots haven't grown.



KNG1 uses alternative splicing to generate two different proteins: High MWt kininogen (HMWK) and MWt kininogen (LMWK). **HMWK** is **essential for blood coagulation** and assembly of the kallikrein-kinin system.

Chromosome 3 (3q26) Translocations/Deletion as Risk Factors for RCC

# Uromodulin should be investigated as a candidate early-diagnosis biomarker for p1RCC

- UMOD can distinguish Normal Tissue from p1RCC with 100% accuracy.
- Uromodulin (encoded by UMOD; also known as Tamm-Horsfall protein) is the most abundant protein in mammalian urine under physiological conditions.

To explore: Is UMOD also a good **urine-based biomarker** for p1RCC?

### Tumorigenicity and FHL1

### Effect of chemicals on expression of top 10 genes: exposure to benzopyrene and several other agents enhances FHL1 expression

	Substance	Up ▼	Dn	FHL1	LRRK2	APBB1IP	TPI
	valproic acid	44	11	Ax Up	Up Mi	Ax Mi Up	
	benzo[a]pyrene	25	16	Up	Dn	Dn Dn Up	Dn
	trichostatin A	24	7		Up		
	aflatoxin B1	22	10	Dm	Dm	Ax	Ax
	cyclosporin A	21	21			Dn	
	2,3,7,8-tetrachlorodibenzodioxi	19	12	Mi Ax Dn Dn	Ax	Dn	MilA
	all-trans-retinoic acid	18	9	Dn			
	phenylmercury acetate	18	11	Dn Mi			U
	17beta-estradiol	17	13	Mi Dn Up Mi			Mi
	bisphenol A	16	31	lm Dn Dn		Dn Ax	
	nickel atom	16	15	Dn Dn	Dn	Up	
	formaldehyde	16	5				
	methyl methanesulfonate	16	14				
	quercetin	15	8	Up			
	6-propyl-2-thiouracil	15	11	Up	Dn	Up	
	ethyl methanesulfonate	14	7		Up		
	butanal	13	15		Dn	Dn	
	silicon dioxide	13	13				
	nickel sulfate	13	15			de	
	tetrachloromethane	13	5				
	N-nitrosodimethylamine	12	1	Up		Up	
					1.33		>



### Thank You!

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