

THE FALSE PREMISE OF THE PROSTATE CANCER GENE BIOMARKER PARADIGM

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Abstract. Prostate tumor heterogeneity questions the value of the widely used gene biomarker paradigm for cancer diagnostic and therapy. We quantified the transcriptomic topologies and interplay of the toll-like receptors and chemokine signaling pathways in point biopsies of three cancer nodules and surrounding normal tissue from a surgically removed metastatic prostate. The analysis revealed that the cancer-related topology remodeling is different even between equally graded cancer nodules within the same tumor, pointing to distinct molecular mechanisms of the immune response. Our results invalidate the cancer biomarker paradigm and indicate the necessity to personalize gene anti-cancer therapy beyond individual patient to his/her cancer nodules' gene master regulators.

Keywords: chemokine signaling, gene master regulators, genomic fabric paradigm, immune response, toll-like receptor signaling.

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Introduction

Prostate cancer (PCa), the out-of-control grow of the prostate gland, is the leading cancer-cause of death for men, with about 1 in 8 men being diagnosed during their lifetime. The incidence of PCa depends on age [1], race/ethnicity [2], diet and lifestyle [3], environment [4] and exposure to radiation and carcinogens [5].

There are several signs and symptoms used to diagnose both early and advanced stages of PCa but the most precise to date is considered the biopsy whose result can be graded using the Gleason score [6]. Clinicians and researchers are using a

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wide spectrum of molecular PCa biomarkers collected from blood [7], urine [8], sputum [9] or direct prostate biopsies.

As of October 5th, 2025, 50,366 PubMed listed publications reported investigations about biomarkers of prostate cancer (e.g. [10, 11], out of which, 10,441 publications were directly related to PCa gene biomarkers. *A PCa gene biomarker is a gene expressed in the prostate gland whose altered DNA sequence [12], 3D configuration [13] or expression level [14] is widely believed as always responsible for the cancerization of the tissue.* Conversely, restoration of the normal sequence, spatial configuration or normal expression of the biomarker was hoped to provide the cure.

The PCa gene biomarkers were selected by automated meta-analyses of large datasets of DNA sequences included in publicly accessible databases like TCGA - The Cancer Genome Atlas [15] and NCI – National (USA) Cancer Institute GDC Data Portal [16]. For gene expression alteration one may search the NCBI-GEO (Gene Expression Omnibus) the public functional genomic data repository of the National (USA) Center for Biotechnology Information [17]. During the years, JacobasLab from Albert Einstein College of Medicine (NY, 2001-2013), New York Medical College (2013-2017) or Prairie View A&M University (2017-) contributed with a total of 86 GEO datasets, out of which 18 were related to different forms of cancer [18]. Clinicians are also using commercial bioassays to detect various kinds of gene mutations [19, 20] or so-called “transcriptomic signatures” [21] of specific forms of cancer.

The large number of publications about cancer biomarkers is a direct consequence of the hundreds of millions of dollars invested in research and billions in revenues for pharma industry and hospitals. Yet we are still far from a definite therapeutic solution. Here, I will prove that the quasi-unanimously accepted Gene Biomarker Paradigm (**GBP**) in PCa research and clinical applications, is not only biologically false and misleading but also therapeutically irrelevant. For this, I will be using own GEO deposited genomic data obtained by profiling the gene expression from three cancer nodules and the surrounding normal tissue located in a surgically removed prostate tumor. The falsity of the GBP will be proven by showing that instead of having the same biomarker gene(s) significantly regulated in the cancerous prostates of all men, the regulomes are different even between equally pathologically graded cancerous nodules from the tumor of the same individual.

What makes GBP unreliable for PCa diagnostic and irrelevant for PCa therapy?

The blamed biomarker is not the single affected gene, the NCI GDC Data Portal listing mutation in 710 protein coding and 1 lncRNA genes responsible for 557 certified PCa cases. However, given the stochastic nature of the chemistry of

DNA replication, 1/1000 bases are randomly mutated, so practically there is no unmutated gene at any time in every cell. In addition, together with the biomarker, expression of hundreds other genes are significantly regulated in cancer in unrepeatable combinations from person to person. Hence, practically no specificity for both mutations and “transcriptomic signature”. Moreover, the meta-analyses combined data collected by different labs, using different platforms and protocols and ignored the high (~30%) technical noise of the used genomic platforms. Nonetheless, the predictive values of the commercial cancer test kits [19 - 21] are disputable given that the number of possible combinatorial outcomes [22] substantially exceeds the number of living males on the Earth (sic!).

Beyond biological considerations, there are several methodological critical points of the GBP. First: the meta-analyses used to identify the biomarkers combined data collected by different labs, performed with various technologies and protocols and ignored the high (~30 - 40%) technical noise of the used genomic platforms. Second: the biomarkers were selected by comparing the average characteristics of a population of PCa patients with one of healthy-considered males, sometimes with the standard human reference genome [23] or pangenome [24]. Nevertheless, the most legitimate reference is the cancer-free tissue surrounding the cancer nodules as adopted by Iacobaslab [25-27] and other groups [28 - 31].

A very important observation is that the PCa biomarkers were selected from the most frequently altered genes among the men affected by this mortal disease. For instance, the five most frequently PCa altered genes listed by the NCI in the most recent data portal are: *IDH1* (76.80%), *TP53* (46.78%), *ATRX* (36.45%), *CIC* (21.44%), *FUBP1* (8.77%). If such a gene is among the most frequently altered, it means that it is one of the least protected by the cellular homeostatic mechanisms as it happens for the least important players in cell life cycle. Therefore, restoring its normal sequence or expression level would be of little consequence and so practically of no therapeutic value. Part of the above criticism was included in the Editorial of a recently published Reprint Book [32].

Materials and methods to test the Gene Biomarker Paradigm

Functional pathways

GBP was tested on the pathways constructed by the KEGG (Kyoto Encyclopedia Genes and Genomes) for toll-like receptor signaling [33], hereafter denoted as “**TLR**” and chemokine signaling [34], hereafter denoted “**CHR**”. TLRs are membrane bound receptors that play critical roles in pattern recognition and host defenses [35]. Activation of TLR signaling pathways might fight cancer by recognizing dangerous damage-associated molecular patterns [36]. Inflammatory-

immune response of the prostate to cancerization implies calling leukocytes to the site of cell depreciation using chemokines, small chemoattractant proteins that provides directional indications for the cell trafficking. The chemokines bind to chemokine (G-protein coupled) receptors expressed in the immune cells. The analyzed pathways shared 33 genes, out of which we quantified 28: *AKT1/2/3*, *CCL3/3L3*, *CCL4L2*, *CCL5*, *CHUK*, *CXCL9/10/11*, *IKBKB/G*, *MAP2K1*, *MAPK1/10*, *NFKB1*, *NFKBIA*, *PIK3CA/B/D*, *PIK3R1/2/3*, *RAC1*, *RELA*, *STAT1/2*.

Prostate samples and gene expression data

The experiments described in this study were conducted according to the guidelines of the Declaration of Helsinki. At the time of the experiment (2016), the study was part of Dr. D.A. Jacobas' project approved by the Institutional Review Boards (IRB) of the New York Medical College's (NYMC) and Westchester Medical Center (WMC) Committees for Protection of Human Subjects. The approved IRB (L11,376 from 2 October 2015) granted access to frozen cancer specimens from the WMC Pathology Archives and depersonalized pathology reports, waiving the patient's informed consent.

The specimens were taken from three adenocarcinoma nodules, hereafter denoted as "A", "B", "C", and the surrounding normal tissue "N", all collected from the surgically removed prostate tumor of a 65y old black man. Nodule "A", located in the left posterior quadrant was received a total Gleason score 9/10 (4 + 5), while the other two nodules were equally graded 8/10 (4 + 4). Every point biopsy was split in four and total RNA was extracted from each quarter, purified, reversed transcribed, fluorescently labeled and hybridized with Agilent-026652 Whole Human Genome Microarray 4x44K v2 [37]. The hybridization protocol and the raw data can be retrieved from the public GEO repository [38].

Gene expression level and variation

According to our standard protocol [39], the expression level of a properly quantified gene "i" in all samples of a particular condition/region "c" were normalized to the median gene and averaged (AVE) over the four biological replicas of that region.

$$\forall i \ \& \ \forall c \rightarrow AVE_i^{(c)} \equiv \frac{1}{4} \sum_{k=1}^4 \left(\frac{\alpha_{i,k}^{(c)}}{\langle \alpha_{i,k}^{(c)} \rangle} \right) \text{ with } a_i^{(c)} = \frac{1}{4} \sum_{k=1}^4 \alpha_{i,k}^{(c)} \quad (1)$$

where: $\frac{\alpha_{i,k}^{(c)}}{\langle \alpha_{i,k}^{(c)} \rangle}$ is the net fluorescence (background subtracted foreground) of spot probing that gene in replica k (=1,2,3,4) of region c.

Moreover, for each gene, we determined also the Relative Expression Variation (REV) across the biological replicas:

$$REV_i^{(c)} \equiv \frac{\sigma_i^{(c)}}{2AVE_i^{(c)}} \left(\sqrt{\frac{r_i}{\chi^2(\beta; r_i)}} - \sqrt{\frac{r_i}{\chi^2(1-\beta; r_i)}} \right) \quad (2)$$

REV is the midinterval chi-square estimate of the coefficient of variation.

Gene expression regulation

Expression of a gene is considered significantly regulated in a cancer nodule with respect to the normal surrounding tissue of the tumor if it satisfies the composite criterion of the absolute expression ratio “ $|x|$ ” exceeding the cut-off “CUT” and the less than 0.05 p-value of the heteroscedastic t- test of means’ equality. “CUT” is computed for each gene to account for the combined effects of the technical noise of the used platform and the variability across the biological replicates:

$$\begin{cases} |x_i^{(cancer/normal)}| > CUT_i^{(cancer/normal)} \\ p_i^{(cancer/normal)} < 0.05 \end{cases} \quad (3)$$

Where:

$$x_i^{(cancer/normal)} = \begin{cases} \frac{AVE_i^{(cancer)}}{AVE_i^{(normal)}}, & \text{if } AVE_i^{(cancer)} > AVE_i^{(normal)} \\ -\frac{AVE_i^{(normal)}}{AVE_i^{(cancer)}}, & \text{if } AVE_i^{(cancer)} \leq AVE_i^{(normal)} \end{cases} \quad (4)$$

$$CUT_i^{(cancer/normal)} = 1 + \sqrt{\frac{2}{100} \left((REV_i^{(cancer)})^2 + (REV_i^{(normal)})^2 \right)} \quad (5)$$

Results

Regulation of the toll-like receptors signaling (TLR) pathway

We profiled 76 out of 109 TLR KEGG-identified genes. Transcripts of missing genes were either not probed by the Agilent microarray platform, or their amount was below the detectable limit, or were hybridized to microarray spots with either corrupted or saturated pixels in at least one sample. Figure 1 presents the significantly regulated TLR signaling [33] genes in the cancer nodules “A”, “B” and “C” with respect to “N”.

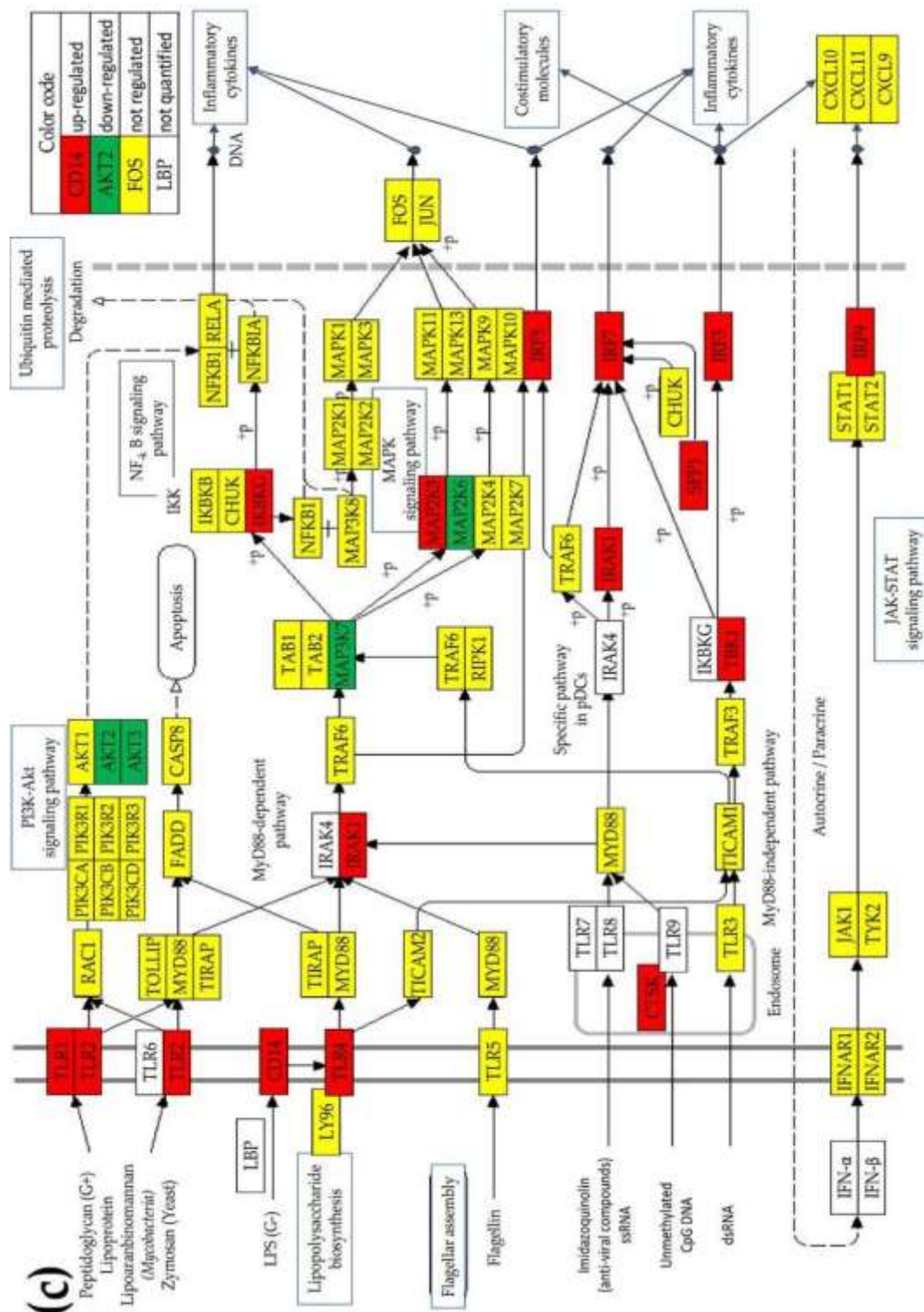
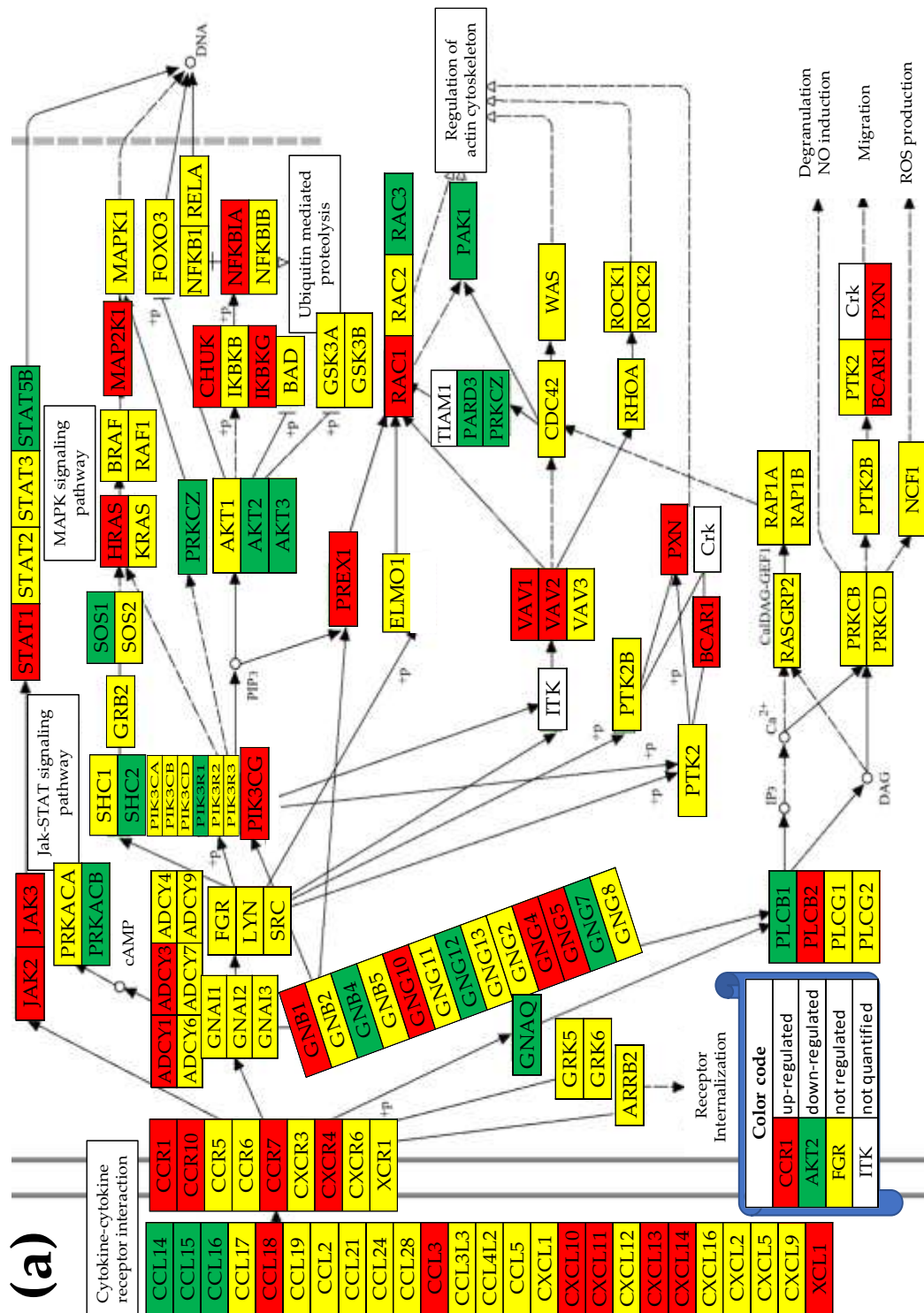


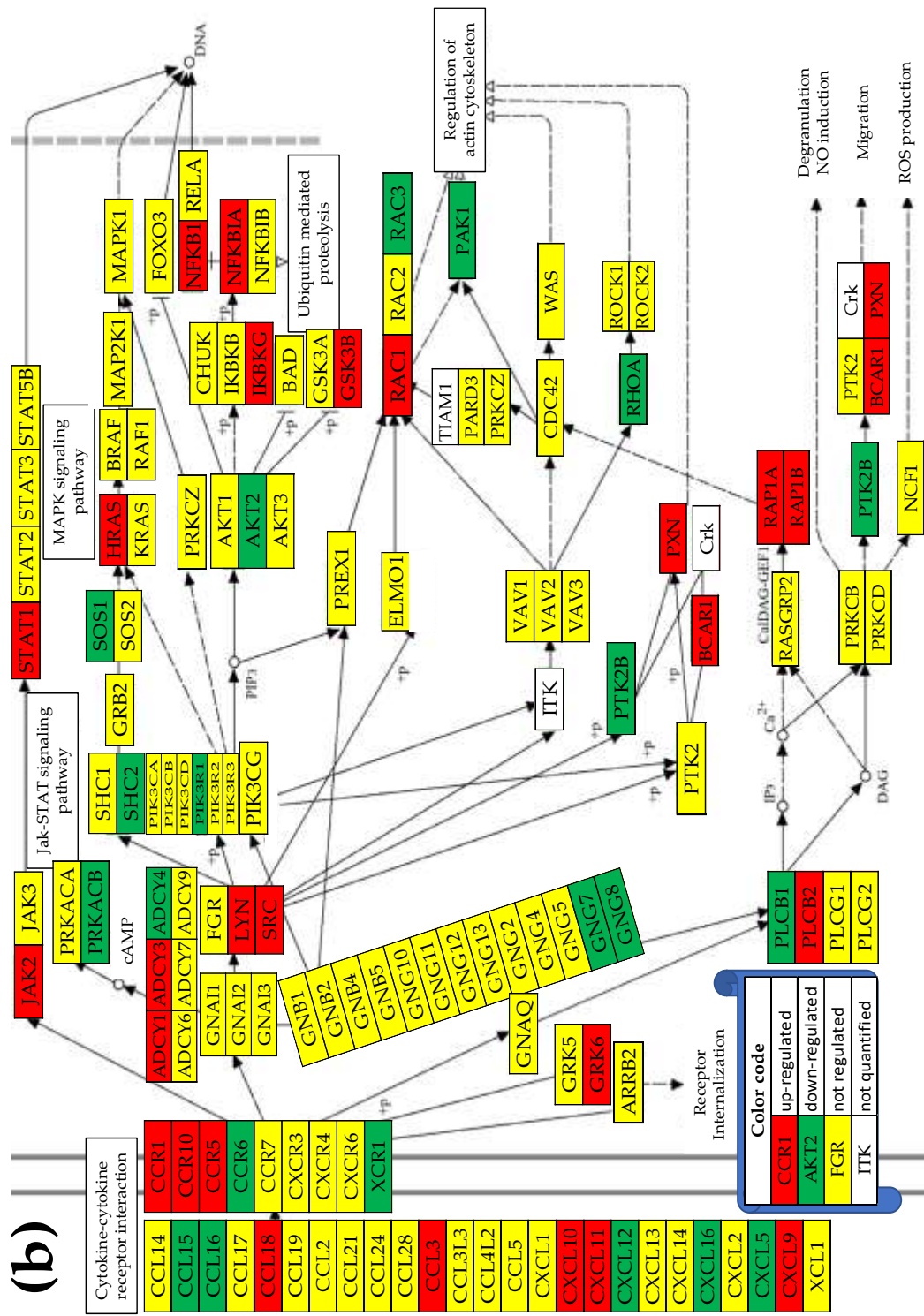
Figure 1: Significant expression regulation of the Toll-like receptors signaling pathway genes in cancer nodules “A” (a), “B” (b), and “C” (c) compared to **normal surrounding tissue “N”**. Regulated genes: *AKT2/3* (v-akt murine thymoma viral oncogene homolog 2/3), *CCL3* (chemokine (C-C motif) ligand 3), *CD14* (CD14 molecule), *CHUK* (component of inhibitor of nuclear factor kappa B kinase complex), *CTSK* (cathepsin K), *CXCL9/10/11* (chemokine (C-X-C motif) ligand 9/10/11), *IFNAR1* (interferon alpha and beta receptor 1), *IKBKG* (inhibitor of kappa light polypeptide gene enhancer in B-cells, kinase gam-ma), *IRAK1* (interleukin-1 receptor-associated kinase 1), *IRF3/5/7/9* (interferon regula-tory factor 3/5/7/9), *JAK1* (Janus kinase 1), *MAP2K1/2/3/4/6/7* (mitogen-activated pro-tein kinase kinase 1/2/3/4/6/7), *MAP3K7* (mitogen-activated protein kinase kinase kinase 7), *MAPK9/10/11/13* (mitogen-activated protein kinase 9/10/11/13), *MYD88* (myeloid differentiation primary response 88), *NFKB1* (nuclear factor of kappa light polypeptide gene enhancer in B-cells 1), *NFKBIA* (nuclear factor of kappa light polypeptide gene enhancer in B-cells inhibitor, alpha), *PIK3R1* (phosphoinositide-3-kinase, regulatory subunit 1 (alpha)), *RAC1* (ras-related C3 botulinum toxin substrate 1 (rho family, small GTP binding protein Rac1)), *SPP1* (secreted phosphoprotein 1), *STAT1* (signal transducer and activator of transcription 1, 91kDa), *TBK1* (TANK-binding kinase 1), *TICAM1* (toll-like receptor adaptor molecule 1), *TLR1/2/3/4/5* (toll-like receptor 1/2/3/4/5), *TNF* (tumor necrosis factor), *TRAF6* (TNF receptor-associated factor 6, E3 ubiquitin protein ligase).

Of note is that even the equally graded nodules “B” and “C” exhibit differences in the gene regulation, proving their transcriptomic uniqueness. Thus, out of the 76 quantified genes, 17 (22.37%) were up- and 10 (13.16%) were down-regulated in “B”, while 13 (17.11%) were up- and 4 (5.26%) were down-regulated in “C”. Nodule “A” had 20 (26.32%) up- and 9 (11.84%) down-regulated genes. Although the three cancer nodules were differently regulated, no TLR gene was oppositely regulated in two nodules. Interesting, *CTSK*, a promoter of the cancer cell proliferation [40] and *CD14*, involved in tumor development [41], were upregulated in all cancer nodules.

Regulation of the chemokine signaling (CHS) pathway

Figure 2 presents the statistically significant regulation of the profiled 135 out of the 191 KEGG-identified CHS [34] genes in the cancer nodules “A”, “B” and “C” with respect to the normal tissue “N”. Of note is that even the equally graded nodules “B” and “C” exhibit differences in the gene regulation, proving the tumor transcriptomic heterogeneity. Thus, 24 (17.78%) were up- and 21 (15.56%) were down-regulated in “B”, while 25 (18.52%) were up- and 8 (5.93%) were down-regulated in “C”. Nodule “A” presented 32 (23.70%) up- and 20 (14.81%) down-regulated genes. Although the significant regulation profiles were different among the three cancer nodules, no CHS gene was oppositely regulated in two nodules.





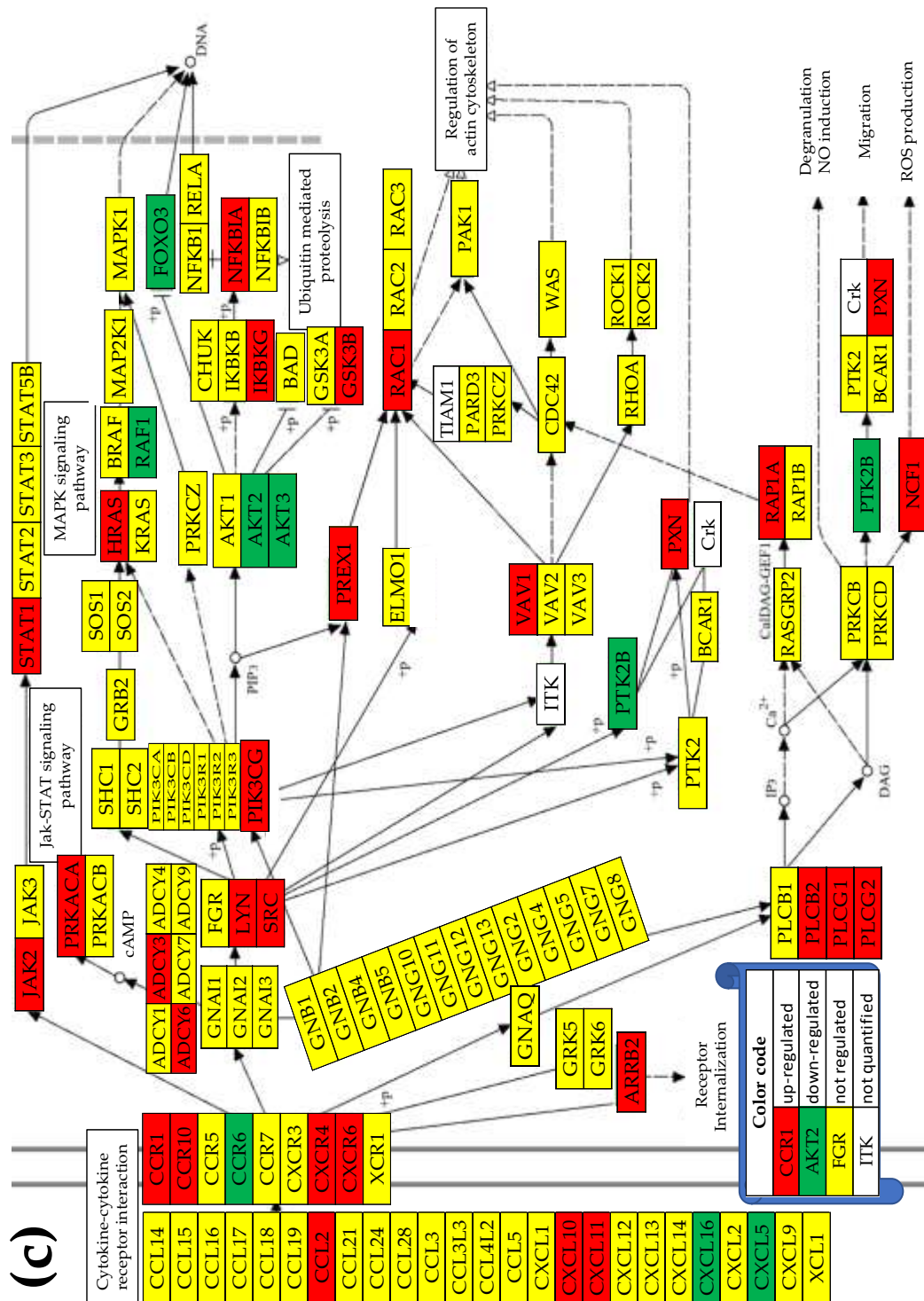


Figure 2: Significant expression regulation of chemokine signaling pathway genes in the cancer nodules “A” (a), “B” (b), and “C” (c) compared to normal tissue “N”. Regulated genes: *ADCY1/3/4/6* (adenylate cyclase 1/3/4/6), *AKT2/3*, *ARBB1/2* (arrestin, beta 1/2), *BCAR1* (BCAR1 scaffold protein, Cas family member), *CCL2/3/14/15/16/18* (chemokine (C-C motif) ligand 2/3/14/15/16/18), *CHUK*, *CXCL5/9/10/11/12/13/14/16*, *FOXO3* (forkhead box O3), *GNAQ* (guanine nucleotide binding protein (G protein), q polypeptide), *GNB1/4* (guanine nucleotide binding protein (G protein), beta polypeptide 1/4), *GNG4/5/7/8/10/12* (guanine nucleotide binding protein (G protein), gamma 4/5/7/8/10/and 12), *GRK6* (G protein-coupled receptor kinase 6), *GSK3B* (glycogen synthase kinase 3 beta), *HRAS* (Harvey rat sarcoma viral oncogene homolog), *IKBKKG*, *JAK2/3* (Janus kinase 2/3), *LYN* (LYN proto-oncogene, Src family tyrosine kinase), *MAP2K1*, *MAPK10*, *NFKB1*, *PAK1* (p21 protein (Cdc42/Rac)-activated kinase 1), *PARD3* (par-3 family cell polarity regulator), *PIK3CG* (phosphatidylinositol-4,5-bisphosphate 3-kinase, catalytic subunit gamma), *PIK3R1*, *PLCB1/2* (phospholipase C, beta 1/2), *PLCG1/2* (phospholipase C, gamma 1/2), *PREX1* (phosphatidylinositol-3,4,5-trisphosphate-dependent Rac exchange factor 1), *PRKACA/B* (protein kinase, cAMP-dependent, catalytic, alpha/beta), *PRKCZ* (protein kinase C, zeta), *PTK2B* (protein tyrosine kinase 2 beta), *PXN* (paxillin), *RAC1/3*, *RAF1* (Raf-1 proto-oncogene, serine/threonine kinase), *RAP1A/B* (member of RAS oncogene family), *RHOA* (ras homolog family member A), *SHC2* (SHC adaptor protein 2), *SRC* (SRC proto-oncogene, non-receptor tyrosine kinase), *STAT1/5B*, *VAV1/2* (vav 1/2 guanine nucleotide exchange factor), *XCL1* (chemokine (C motif) ligand 1), *XCRI* (chemokine (C motif) receptor 1).

Discussion

The substantial tumor transcriptomic heterogeneity, even between equally graded nodules, as illustrated for gene expression regulation in Figures 1bc and 2bc, questions the validity of the gene biomarker paradigm for the prostate cancer. The differential significant regulation patterns observed in nodules “B” and “C” not only as affected genes but also as percentages (22.37% vs. 17.11% up-regulated and 13.16% vs. 5.26% down-regulated TLR genes, respectively) emphasize the importance of considering intra-tumoral heterogeneity in therapeutic strategies [42 – 45]. For instance, while three TLR signaling mitogen-activated kinases (*MAP2K2*, *MAP2K3*, *MAPK13*) were upregulated and five (*MAP2K4*, *MAP2K6*, *MAPK9*, *MAPK10*, *MAPK11*) were down-regulated in “B”, only *MAP2K3* was upregulated and *MAP2K6* was down-regulated in “C” (Fig.1). The differential regulation of the mitogen-activated kinases in the two equally graded nodules indicates that participation of the MAPK signaling [46] to the “development, invasion, metastasis, and drug resistance” [47] of cancer cells is not uniform even across the same prostate. Also, while four CHS signaling guanine nucleotide-binding proteins (*GNB1*, *GNG10*, *GNG4*, *GNG5*) were up-regulated and three (*GNB4*, *GNG12*, *GNG7*) were down-regulated in “A”, only two genes (*GNG7*, *GNG8*) were down-regulated in “B” and no gene was up- or down-regulated in “C” (Fig.2).

Conclusions

1 Although limited to one case of prostate cancer, the present study proved that beyond similarities, not only every man develops a unique form of prostate cancer but even within the same tumor each cancer nodule has its own “transcriptomic personality”. The gene expression differences occurring even between equally pathologically graded cancer nodules in the same tumor makes unlikely the transcriptomic uniformity across distinct individuals as predicted by GBP. Despite here we investigated only two (albeit major) functional pathways, most likely, inter-nodule transcriptomic differences extend to all other components of the immune system.

2 GBP was invalidated also by our analysis of: mTOR signaling, metabolic pathways and expression coordination of *AKT2* with its partners central to the KEGG-constructed PCa pathway [48] in another prostate tumor [49]. Similar conclusion was obtained by us for the P53-signaling, apoptosis, block of differentiation, evading apoptosis, immortality, insensitivity to anti-growth signals, proliferation, resistance to chemotherapy, and sustained angiogenesis [50]. We found also that the immortalized PCa cell lines LNCaP and DU145 exhibited substantial differences in the gene expression compared to the standard prostate tissue [51, 52].

3 Altogether, the observed uniqueness of the gene expression profiles requires development of personalized gene therapy approach that has to go even to cancer nodule level, targeting the gene master regulators as proposed and tested by us on prostate, kidney and thyroid cancers [53 - 55].

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