Whole Body Melanoma Transcriptome Response in Medaka

Manfred Schartl¹², Yingjia Shen³, Katja Maurus¹, Ron Walter³, Chad Tomlinson⁴, Richard K. Wilson⁴, John Postlethwait⁵, Wesley C. Warren⁴*  

¹ Physiological Chemistry, University of Würzburg, Biozentrum, Am Hubland, 97074, Würzburg, Germany, ² Comprehensive Cancer Center, University Clinic Würzburg, Josef Schneider Straße 6, 97074, Würzburg, Germany, ³ Department of Chemistry and Biochemistry, 419 Centennial Hall, Texas State University, 601 University Drive, San Marcos, TX, 78666, United States of America, ⁴ McDonnell Genome Institute at Washington University, 4444 Forest Park Blvd., St Louis, MO, 63108, United States of America, ⁵ Institute of Neuroscience, University of Oregon, 1425 E. 13th Avenue, Eugene, OR, 97403, United States of America

* wwarren@genome.wustl.edu (WCW); phch1@biozentrum.uni-wuerzburg.de (MS)

Abstract

The incidence of malignant melanoma continues to increase each year with poor prognosis for survival in many relapse cases. To reverse this trend, whole body response measures are needed to discover collaborative paths to primary and secondary malignancy. Several species of fish provide excellent melanoma models because fish and human melanocytes both appear in the epidermis, and fish and human pigment cell tumors share conserved gene expression signatures. For the first time, we have examined the whole body transcriptome response to invasive melanoma as a prelude to using transcriptome profiling to screen for drugs in a medaka (Oryzias latipes) model. We generated RNA-seq data from whole body RNA isolates for controls and melanoma fish. After testing for differential expression, 396 genes had significantly different expression (adjusted p-value < 0.02) in the whole body transcriptome between melanoma and control fish; 379 of these genes were matched to human orthologs with 233 having annotated human gene symbols and 14 matched genes that contain putative deleterious variants in human melanoma at varying levels of recurrence. A detailed canonical pathway evaluation for significant enrichment showed the top scoring pathway to be antigen presentation but also included the expected melanocyte development and pigmentation signaling pathway. Results revealed a profound down-regulation of genes involved in the immune response, especially the innate immune system. We hypothesize that the developing melanoma actively suppresses the immune system responses of the body in reacting to the invasive malignancy, and that this mal-adaptive response contributes to disease progression, a result that suggests our whole-body transcriptomic approach merits further use. In these findings, we also observed novel genes not yet identified in human melanoma expression studies and uncovered known and new candidate drug targets for further testing in this malignant melanoma medaka model.
Introduction

While the incidence of most cancers is decreasing, malignant melanoma (MM) continues to increase by 3–7% per year in Caucasians and has a 5-year survival rate of only 20% [1]. Melanoma genome sequencing and exome sequencing of an astonishingly high heterogeneity of clinical samples have led to the identification of a large catalog of recurrent somatic variants [2], most notably BRAF and NRAS. However, patient outcomes from treatment with novel compounds are compromised by an almost 100% recurrence rate due to rapidly occurring drug resistance [3] and melanoma suppression remains unpredictable when targeting protein-coding genes identified in large cohort sequencing projects. A deeper exploration of cause-and-effect melanomagenesis data is urgently needed. At present, vertebrate melanoma models exist for zebrafish, medaka, platyfish, mouse, the Syrian hamster, Mongolian gerbil and opossum [4–7], each with its own advantages and disadvantages for utility in drug discovery and molecular mimicry of human melanoma.

In contrast to humans, mice and other rodent skin is devoid of epidermal melanocytes as they occupy the hair follicles and dermis, except for some in the skin of footpads, tail, ears and snout, whereas fish melanoma models mimic the appearance of melanocytes in the human epidermis. Another advantage appears to be the lack of hair cycling which periodically changes the profile of gene expression in skin and other organs, which complicates studies on tumor environment. Fish models available in zebrafish and medaka have additional advantages, including easy transgenesis, superior suitability for bioimaging analyses due to the transparency of embryos and larvae and importantly, their suitability for high throughput chemical screens [8, 9]. While single driver gene melanoma models, such as medaka and zebrafish, are initially genetically less complex than human melanomas, the expression signatures of zebrafish and medaka melanomas clearly show common features with human melanoma [10, 11]. Fish models have proven their value for melanoma research and have become well-accepted tools for furthering our knowledge and for developing better diagnostic and therapeutic tools [7, 12].

Several studies have examined gene expression changes in zebrafish, medaka, mouse and human melanomas [13–15]. A recent meta-analysis of human melanoma microarray data suggested that 17 genes show aberrant expression [15]. All approaches [13–15] thus far reported have dissected out melanoma tumor cells of various classifications, mostly from the metastatic disease, and compared results to normal skin or melanocytic nevi. None of these studies measured the whole-body transcriptome in established melanoma and compared results to normal individuals. This type of comparison is necessary because tumors also have effects on entire organisms; furthermore, tissue systems throughout the organism play a role in tumor growth and disease, especially the vasculature, endocrine, and immune systems [12, 16]. Melanoma is a highly invasive and metastatic tumor where already early on, small nests or single tumor cells colonize the whole body. This is further complicated by the fact that these microtumors are often non-pigmented and reside far away from the primary lesion, making them easy to overlook even by an experienced experimenter. Therefore, our strategy was to compare the transcriptomes of intact melanoma fish rather than dissecting the melanoma from the “healthy” part of the individual. This strategy allowed for the establishment of a melanoma transcriptional disease signature and identified disregulation of various canonical pathways that may indicate a systemic response to the spread of malignant melanocytes.

Material and Methods

Sequencing

A transgenic melanoma model in medaka (Oryzias latipes) was used to determine the transcriptional disease signature (TDS) of melanoma. In this model, the oncogenic receptor
tyrosine kinase Xmrk, which is encoded by the melanoma-inducing gene from the platyfish (Xiphophorus maculatus), is specifically expressed in pigment cells under control of the medaka mitf promoter [17]. The stably integrated transgene was introduced into an inbred strain (Carbio) of medaka by introgressive breeding. Fish were kept under standard conditions in the aquarium facility of the Biozentrum at the University of Würzburg and all studies were approved by the Institutional Review Board. For RNA-seq experimentation, total RNA was extracted from the intact bodies of size-matched (1cm total length) non-transgenic and transgenic fish 3–5 weeks old using an RNeasy Qiagen kit (Venlo, Netherlands) according to the manufacturers’ protocol. To compare gene expression between controls and melanoma fish, a biological sample was prepared and sequenced individually from four individual melanoma fish and four individual control fish, thus providing four independent biological replicates of each condition. We checked total RNA for quality on the Agilent Fragment Analyzer, then enriched for poly(A)+ RNA using the MicroPolyA Purist kit (Ambion, Carlsbad, CA). We used ScripSeq (Epicentre, Madison, WI) to generate strand-specific cDNA that was sequenced on the Illumina Hiseq2000 platform as 100 base paired-end reads (insert size of 400bp). All sequences are available under accession number PRJNA219474.

**Differential gene expression**

Across samples, an average of 34.5 million read pairs with average base quality >35.7 were obtained after filtering for duplicates and low quality bases. Sequences from each sample were mapped to medaka transcripts (Ensembl version 80) using TopHat2 [18]. A custom perl script, available on request, was used to count mapped reads per each transcript in each sample. DESeq2 was used to analyze read count data from the previous step and to identify differential expression levels based on an error model that uses the negative binomial distribution[19]. In this study for all control to tumor gene comparisons we used an adjusted p-value of <0.02 as our significance threshold for further biological inference.

**RNA-seq validation**

For confirmation of RNA-seq data, selected genes were assayed with qPCR from ten control and ten melanoma fish individually (siblings of fish used for the RNA-seq experiments and two melanoma fish that were also used for the RNA-seq). RNA was isolated from whole fish bodies using the RNeasy Mini Kit (Qiagen, Venlo, Netherlands) according to the manufacturers’ protocol. cDNA was prepared from total RNA using the RevertAid kit (Fermentas, Wall-tham, MA) with random hexamer primers. 25 ng of each cDNA was analyzed in triplicate in a 25 μl volume using a SYBR green-containing mastermix for 5 min at 95°C followed by 40 cycles of 95°C for 30 s, 60°C for 30 s and 72°C for 20 s in a mastercycler ep realplex (Eppendorf, Hamburg Germany). Expression of each gene was normalized to ef1a1 levels, a positive control gene, and fold changes were calculated versus one wild-type fish. All p-values were calculated by applying the Mann-Whitney-U-test.

**Pathway and transcription regulator enrichment**

To infer further functional information from malignant melanoma whole body transcriptomes, it was necessary to match each regulated medaka gene to its human ortholog and, when possible, affiliated human gene symbol. Of 396 differentially regulated medaka genes (adjusted p value < .02), we were able to obtain 233 medaka genes with matching human gene annotation, a prerequisite for many statistical tests of functional enrichment. We used a combination of human orthology retrieval through the Ensembl BioMart tools (a predefined set of multispecies orthologs) and manual alignment curation by BLAST to derive this set. It is recognized in
some cases that an additional copy of each gene could be present in the medaka genome because teleosts experienced an additional unique whole genome duplication event [20, 21]. The human gene symbols and log ratio values for each differentially expressed medaka gene were input into the Ingenuity core analysis software [22]. We then performed canonical pathway enrichment and upstream regulatory activation analyses. To measure statistical significance for over-represented canonical pathways, we used a right-tailed Fisher's exact test (p-value cutoff < .01). To ascertain transcription regulators that could explain observed gene expression changes, we relied on two test statistics 1) a p-value overlap score was calculated with a Fishers exact test, whereby a statistical overlap is calculated between differentially expressed genes and the known targets of the transcription regulators (p-value cutoff < .01) and 2) a activation z-score that attempts to infer activation states based on prior knowledge of regulation by the regulator compared to a random model of regulation, a z-score of greater than 2 or less than -2 were considered significant [22].

Drug interaction

To discover genes that could be targeted with known drugs, we input the 233 medaka expression candidates with human orthologs into the DGIdb database (http://dgidb.genome.wustl.edu) and performed a search with default screen parameters to find all gene to drug interactions using the antineoplastic filter [23].

Results

We used our transgenic medaka model (tg(mitf:xmrk)) for melanoma where the pigment cell-transforming oncogene xmrk [24] isolated from the platyfish Xiphophorus maculatus was placed under control of the medaka mitf (microphthalmia-associated transcription factor) promoter, which drives oncogene expression in melanocytes [17]. The mitf promoter confers strong but strictly pigment cell-specific expression to the tumor-inducing driver-oncogene xmrk with no ectopic or background expression [17, 25–27]. In the homogenous genetic background of the medaka strain used here (Fig 1A), transgenic juveniles start to develop highly invasive extracutaneous malignant melanoma at late larval stages (3–4 weeks of age) with almost 100% penetrance [17](Fig 1B). At later stages (7–8 weeks), these melanomas progress to metastatic disease with multiple large nodular and invasive tumors in the abdomen and
trunk musculature (Fig 1C). In the initial stage of melanoma development, the progression of the malignant disease is mostly apparent by invasion of tumor cells deeply into the trunk musculature, where they follow the space between single muscle bundles (Fig 1D). At multiple extracutaneous sites proliferation of melanoma cells that arrived from elsewhere leads to tumor nodules and lesion expansion (Fig 1E).

We generated RNA-seq data from whole body RNA isolates for four melanoma and four control siblings at 3–4 weeks of age, a stage when the tumors switch from localized lesions to progressive malignant disease. Transcript read counts >10 per individual in at least 50% of all fish (control and tumor) were generated for 19,985 of 25,434 Ensembl medaka genes. After read count normalization and statistical testing for differentially expressed genes [28], 396 of medaka genes were found to have significantly different expression levels (adjusted p-value < 0.02) between melanoma and control fish. Among all differentially expressed genes, a total of 233 matched human orthologs with descriptive gene annotation, which allowed for putative biological inference, were considered further for statistical enrichment tests (Table A in S1 File). Independent quantitative PCR validation of a sub-set of these regulated genes confirmed the direction of expression for a majority (6 of 8; Table B in S1 File). We purposefully chose genes of importance to melanoma biology (irf1, atm), immune systems (trl2, stat4) and others of unknown melanoma association but significantly regulated in our medaka model (stc1, slc24a5, epd1, gadd45a). The reason for non-validation is unknown, but several possibilities exist, including paralogs confounding probe hybridization kinetics.

Overall in our study, more genes were down-regulated (n = 138) than up-regulated (n = 95) in melanoma fish compared with controls (Table A in S1 File). Of this set, 14 matched human genes shown to harbor recurrent somatic mutations [29] (Table C in S1 File). The upper and lower log ratio expression extremes were +2.48 and -5.67 for fkbp5 and snord14c, respectively. The role of snord14c in melanoma progression is unknown, as snord14c is a member of the small nucleolar RNAs (snoRNAs) and is classified as a C/D box snoRNA that generally are linked to methylation [30]. In contrast, fkbp5 is a member of the immunophilin protein family, which play a role in immunoregulation and basic cellular processes involving protein folding and trafficking and has been recently linked to melanoma dissemination [31].

We sought to validate our whole body transcriptome model by examining expression directionality of known molecular participants of the melanocyte development and pigmentation signaling pathways. The expression levels of five gene members were significantly (adjusted p-value <.02) elevated or decreased: shp, creb, pi3k, plcg2, and tyrp1 (Fig 2). As in other vertebrates, medaka melanomas originate from neural crest-derived pigment synthesizing cells and elevated expression of mitf, a key transcriptional regulator in melanoma, has been functionally linked to the neoplastic phenotype in humans[32]. However, in the whole body transcriptome, the mitf gene was up regulated two-fold in the melanoma fish but did not reach our p-value < .02 significance threshold (adjusted p-value 0.19). Nonetheless, the pigmentation gene tyrp1 showed a elevated level of expression in the whole body of melanoma fish as a result of mitf upregulation (Fig 2; p-value <0.0001). Another upregulated gene is hypoxia inducible factor 3. In mammalian melanocytes induction of melanin pigmentation led to upregulation of hif1α expression and related pathways [33]. A well-established melanoma marker in dermatocology, pmel, was increased as well (adjusted p-value <0.00002). A striking and novel feature not noted earlier in expression studies using dissected fish melanoma was a general down-regulation of components of the innate immune system. Some other significant examples of note include: interferon response factor 1 (irf1), several MHC, effectors of the cytotoxic T-cell response including perforin (prf1), complement components (c3, c1qa) and cytokines (il12).

To discover programs of responses to invasive melanoma at its initial stage, we tested for significant participation among signaling and metabolic pathways, molecular networks, and
biological processes that are most perturbed in these juvenile melanoma fish [22]. A total of 46 pathways were considered further (p-value < 0.01; Table D in S1 File). One of the significantly scored pathways was melanocyte development and pigmentation (Fig 2; Table D in S1 File), again validating our intact-fish approach for understanding organism-wide changes to gene expression that are likely to be highly relevant for indirect correlation to human melanoma progression. Two of the other highest scoring pathways were antigen presentation (p-value 0.0000014) and cytotoxic T lymphocyte-mediated apoptosis (p-value 0.000251). Antigen presentation is a vital process of the immune system with many cellular participants acting as sentinels that oscillate between detecting self versus entities that disrupt body homeostasis, i.e., cancer. We found members b2m, cd74, hla-dqa1, hla-drb1, psmb9, and tap2 of this pathway to all be down-regulated in melanoma fish (Table A in S1 File). After manual curation, a total of 6 of the top 10 scoring pathways (p-value < 0.001; Table D in S1 File) were affiliated with the immune system. To our knowledge this phenomenon was not observed in previous studies of the fish melanoma models where dissected tumors were analyzed [11, 34, 35].
To expand our systemic knowledge of whole body melanoma transcriptomes, we sought to elucidate the putative upstream regulators of melanoma-expressed genes for probable states of activation or inhibition that may provide testable hypotheses of downstream effects on melanoma sustainability (Table 1). We used the inferred activation states of transcriptional regulators given the observed differential state of genes in our medaka melanoma model for this purpose [22]. A z-score statistic was used to predict expression states, activated or inhibited, that we set at -2 or +2 to reach significance. Given this threshold, we found 17 and 16 molecules that predict an inhibitory or activated effect of a given regulator, respectively, on various medaka genes expressed in this whole body melanoma model (Table 1). A cumulative list of all molecules that met this statistical criteria are provided for further review (see S2 File). The

| Upstream Regulator | Molecule Type         | Predicted Activation State | Activation z-score | Medaka melanoma gene network count |
|--------------------|-----------------------|----------------------------|--------------------|-----------------------------------|
| IFN-gamma          | cytokine              | Inhibited                  | -5.082             | 33                                |
| TNF                | cytokine              | Inhibited                  | -3.173             | 24                                |
| CD40LG             | cytokine              | Inhibited                  | -2.962             | 12                                |
| LPS                | chemical drug         | Inhibited                  | -2.774             | 24                                |
| IRF7               | transcription regulator| Inhibited                  | -2.63              | 7                                 |
| IL1B               | cytokine              | Inhibited                  | -2.626             | 17                                |
| IL15               | cytokine              | Inhibited                  | -2.583             | 11                                |
| STAT1              | transcription regulator| Inhibited                  | -2.476             | 12                                |
| LFNAR              | group                 | Inhibited                  | -2.449             | 6                                 |
| IRF1               | transcription regulator| Inhibited                  | -2.361             | 9                                 |
| CD40               | transmembrane receptor | Inhibited                  | -2.356             | 6                                 |
| IL2                | cytokine              | Inhibited                  | -2.253             | 12                                |
| Aldesleukin        | biologic drug         | Inhibited                  | -2.219             | 5                                 |
| RELA               | transcription regulator| Inhibited                  | -2.13              | 15                                |
| IFN-alpha          | group                 | Inhibited                  | -2.095             | 8                                 |
| Cisplatin          | chemical drug         | Inhibited                  | -2.064             | 10                                |
| Inosine            | chemical              | Inhibited                  | -2                 | 4                                 |
| Alefacept          | biologic drug         | Activated                  | 2                  | 4                                 |
| OTX2               | transcription regulator| Activated                  | 2                  | 4                                 |
| CDKN2A             | transcription regulator| Activated                  | 2                  | 5                                 |
| SB203580           | chemical—kinase inhibitor| Activated                  | 2.144              | 6                                 |
| LNS1               | other                 | Activated                  | 2.172              | 5                                 |
| PRDM1              | transcription regulator| Activated                  | 2.183              | 5                                 |
| Mibolerone         | chemical drug         | Activated                  | 2.2                | 5                                 |
| PPARG              | ligand-dependent nuclear receptor | Activated | 2.213 | 5                                 |
| MAPK1              | kinase                | Activated                  | 2.219              | 6                                 |
| APOE               | transporter           | Activated                  | 2.226              | 7                                 |
| Forskolin          | chemical toxicant     | Activated                  | 2.236              | 15                                |
| Akt                | group                 | Activated                  | 2.236              | 5                                 |
| RICTOR             | other                 | Activated                  | 2.236              | 5                                 |
| CD3                | complex               | Activated                  | 2.447              | 15                                |
| PPARA              | ligand-dependent nuclear receptor | Activated | 2.651 | 12                                |
| IL10               | cytokine              | Activated                  | 2.751              | 9                                 |

doi:10.1371/journal.pone.0143057.t001
three molecules of highest inferred inhibitory significance were \( \text{ifn} \gamma \), \( \text{cd40lg} \), and \( \text{tnf-alpha} \). Interestingly, we found that \( \text{ifn} \gamma \) is the top ranked upstream regulator of genes expressed in medaka developing melanoma (z-score -5.08). These putatively \( \text{ifn} \gamma \) regulated genes (33 total) span several overlapping molecular pathways as shown in a network based presentation (Fig A in S1 File). The majority (82%) of the 33 differentially expressed medaka genes in melanoma fish were down-regulated as part of our curated \( \text{ifn} \gamma \) regulatory network and all but one, \( \text{ehf} \), agree with expectations based on the Ingenuity curated expression database.

Thus far, fish have provided useful models for furthering our understanding of melanoma biology and basic molecular mechanisms and for screening of potential drugs that act directly on the tumor [36]. When all differentially regulated expression candidate genes with matching human gene orthologs (n = 233) were examined for known drug to gene interactions, we found 10 genes with 73 unique drug interactions [23], with the phosphatidylinositol kinase gene \( \text{pik3r6} \) contributing 21 unique drug interactions (Fig 3). In addition, \( \text{pik3r6} \) displays recurrent somatic mutations in human melanoma and would thus be considered a high priority target in future experiments designed to utilize drugs that target this gene in our melanoma fish and observe therapeutic outcomes [29].

Discussion

The effects of genetic background on human malignancy progression is an important but poorly understood phenomenon [37]. Our system provides a significant opportunity to explore this problem and already its investigation has led to the identification of several melanoma genes, including osteopontin [38], which now serves as a clinical marker for human disease progression [39]. Among fish models, medaka presents as a unique feature the availability of many highly inbred strains that allow one to study cancer formation in animals of the same or a variety of different genetic backgrounds. Furthermore, the \( \text{xmrk} \) transgene on different genetic backgrounds provides different types of melanoma [17] and transforms mouse and human melanocytes [26, 27] that result in highly similar biochemical and genetic changes [10, 29].
14], making this system ideal for identifying both genetic modifiers of tumor type and drugs tailored to different genotypes. The pigment lesions in our mitf:xmrk transgenic medaka at the time of RNA-seq sampling are clearly not just "local hyperpigmentation" in the dermis but an aggressive malignancy that is located predominantly at extracutaneous sites, mainly as invasive growth between the muscle fibers and bones. Melanoma cells contain black pigment, and due to the transparency of the juvenile medaka, the area covered by melanin inside the fish in the trunk is a good proxy for estimating the size of the tumor. Moreover, the single homologous gene driver and the short time window in which tumors fully invade the whole body excludes many types of secondary (passenger) mutations that might occur. To exploit this situation, we explored the use of whole body transcriptome profiles to catalog gene candidates for therapeutic intervention and to discover the body’s canonical pathway responses to highly invasive melanoma.

The whole-body transcriptomes established in our study from melanoma-bearing transgenic medaka readily detected changes in canonical pathways known from studies on human melanoma [40], but also show a clear differentiation from earlier RNA-seq analysis of melanoma biopsies from adult medaka fish [10]. We identified a considerable number of genes that were not detected earlier when analyzing tumors in isolation [10]. For instance, snord14c showed the strongest down-regulated state in the whole body melanoma transcriptome, yet its role in melanoma formation is not understood to our knowledge, and thus identifies a new putative candidate for the further understanding of the molecular mechanisms of melanoma progression, perhaps highlighting a role for RNA methylation. The fkbp5 gene is another example of a potential contributor to melanoma dissemination in this medaka expansion model. Recent studies highlighted fkbp5 in the transformation of transforming growth factor-β from a tumor suppressor to pro-metastatic aggressor, facilitating cancer cells ability to spread to distant sites from the original primary tumor site [31].

Our whole body set of differentially expressed genes, critically collected when melanoma is in invasive mode, allowed us to search for molecular pathways that are significantly enriched in the melanoma developing fish. Given the transgenic initiation of melanocyte malignancy in our model, it was predicted at least some known genes associated with this pathway would be differentially regulated. The high scoring melanocyte pathway clearly shows that our whole body transcriptome measures were sufficient to detect modulation of some member genes, such as tyrp1, over whole body signal noise. The increase of tyrp1 expression is interesting since melanin synthesis, melanogenesis and melanin pigment affect the behavior of normal and malignant melanocytes. In melanoma cells melanogenesis is related to shorter survival of patients [41]. More importantly, our results shine new light on the direct relevance of the mechanisms by which immune gene networks respond or fail to respond to the expanding population of malignant melanocytes and are likely highly relevant to human melanoma pathogenesis. As a consequence of melanocyte proliferation and spread, we hypothesize that intact body systems attempt to identify and kill malignant melanocytes, but too often this alarm to the immune system is insufficient to overcome a fast growth tumor trajectory, especially if cellular cytoskeleton dynamics are not optimum. A heightened immune response could be compromised by reported immunosuppressive properties of malignant melanocytes [42]. In fact, an important immune pathway, antigen presentation, was found to be enriched for several differentially expressed medaka member genes and we observed down-regulation for all in the melanoma state. This may indicate mechanisms of immune escape of the fish pigment cell tumor in its initial stage and resembles conditions that have been noted also in human melanoma [43]. Further expansion of our list of statistically significant genes regulated within the known immune signaling pathways is likely with deeper sequencing per whole body melanoma isolate.
The inability of the immune system to detect and react with sufficient intensity to check progressive melanoma has confounded scientists. Some evidence suggests that melanoma has an immunosuppressive capability [42, 44]. In particular, intermediates in melanin synthesis, e.g. L-DOPA have immunosuppressive capability [45, 46]. We found several genes in our differentially expressed gene set that, when examined at the network level, broadly suggest immune system suppression; it will be important, however, to differentiate the mechanisms of immunosuppressive or atypical inflammatory responses for these candidates. Within several immune pathway classifications, such as antigen presentation, dendritic cell maturation, and cytotoxic T lymphocyte-mediated apoptosis, we observe all but one gene showing decreased expression in medaka melanoma. Together with the finding that most genes involved in immune system pathways were down-regulated in this early progression phase of fish melanomas, this result suggests that immune evasion might be important for the invasive success of malignant melanoma. Interestingly, a soluble form of pmel can protect melanoma cells from antibody-mediated immunity [47], while the membrane-bound form activates the immune response via dendritic cells [48]. While the transcript isoform of this gene was highly up-regulated in our melanoma fishes, protein form responses will need to be measured to fully understand the immunomodulatory consequences. Of note, in this study we believe the immune suppressive state that we observed at a very early stage (3–4 weeks) of tumor formation excludes the explanation that immune evasion is due to additional acquired mutations; rather, it is an indigenous feature of melanoma cells.

Utilizing prior knowledge of expected directional relationships between transcriptional regulators of all types (drugs, transcription factors, small molecules, etc.) and downstream expressed genes is a powerful model for high-level melanoma gene network inference (Fig A in S1 File). Our analysis of known upstream regulators strongly suggests that melanoma fish experience a robust suppression of the ifnγ responsive gene network, the highest scoring network in this study. Recently, an unexpected connection between the immune cytokine ifnγ and skin pigmentation homeostasis was found [49] mediated through the transcription factor irf1, a gene that we find consistently downregulated in the melanoma developing medaka. High-dose interferon has been used for nearly two decades as adjuvant treatment of patients with melanoma who have undergone a complete surgical resection but who are considered to be at a high risk of relapse. Evidence leading to interferon therapy was based on significantly improved relapse-free survival and marginally improved overall survival. Subsequent large, randomized trials have not been able to reproduce a benefit in overall survival, yet interferon treatment continues to be a first option for those patients [50]. Stat5, which was first identified as a melanoma-relevant gene in our fish melanoma model [51], is now used in the clinic to elucidate mechanisms of resistance to interferon treatment in human patients [25]. Collectively, our network inferred immune-system gene candidates could provide points of entry for therapeutics to elevate the immune response of the host to progressing melanoma or conversely, to subdue putative immunosuppression by malignant melanocytes. Such molecular insight will be challenging without our whole-body approach.

Importantly, our whole-body expression screen identified candidates for future drug screens, in particular immunoregulatory molecules. Drug screen outcomes will require network or systems biology interpretation because some genes participate in multiple pathways, for example, phosphatidylinositol kinases act in leukocyte extravasation and antigen pattern recognition pathways. A total of 11 aberrant medaka melanoma gene expression candidates found in our study, some examples include aadac, dhdh, dusp13, and pik3r6, have experienced missense (amino acid substitution) mutations in human melanoma genomes [52]. Despite the unknown clinical relevance of these mostly missense mutations in melanoma patients, we suggest further investigation of any drugs known to target these proteins translated from
differentially expressed genes in our fish melanoma model, but not currently used to treat human melanoma, is needed to test transcript outcome in isolation and synergy, i.e., drug combinations [29]. We contend that a reduced transcript network, that indirectly mimics the transcriptional disease signature, could be used to cost effectively screen drugs in a high throughput system.

In summary, building genetic networks that adequately represent whole body transcriptome responses to invasive cancer is crucial to the exploration of secondary pathways for melanoma suppression. Cancer genome-sequencing studies have identified recurring mutational signatures in various cancers including melanoma, but the corresponding immunological signatures of tumors are highly variable and are far from well understood. Closing this knowledge gap is likely to be important, given that infiltration of immune cells into tumors is correlated with positive prognoses [53, 54]. A better understanding of the complex interactions between cancer cells and the immune system is likely to lead to improvements in current therapeutic approaches and to spur the development of novel therapies. Our study 1) illustrates how focusing only on the tumor itself might leave unknowns that are revealed when whole animals are taken into account; 2) shows how the tumor can affect non-tumor body systems; and 3) provides insights into new potential therapeutic targets. The whole body transcriptome based characterization of the medaka system can serve as model for further research on melanoma tumor immunology.

Supporting Information

S1 File. (DOCX)

S2 File. (XLS)

S3 File. (XLS)

Acknowledgments

This work was supported by NIH grant R24 RR032658-01 to WCW and 1R24OD018555 to JHP, MS, RW, and WCW and Deutsche Forschungsgemeinschaft through GRK1048 and TR17 and the Deutsche Krebshilfe through the Melanomverbund to MS.

Author Contributions

Conceived and designed the experiments: WCW MS JP RW. Performed the experiments: KM YS CT WCW. Analyzed the data: YS CT KM. Contributed reagents/materials/analysis tools: RKW. Wrote the paper: MS WCW RW JP.

References

1. Rigel DS. Cutaneous ultraviolet exposure and its relationship to the development of skin cancer. J Am Acad Dermatol. 2008; 58(5 Suppl 2):S129–32. Epub 2008/04/25. S0190-9622(07)02413-9 [pii] doi: 10.1016/j.jaad.2007.04.034 PMID: 18410798.
2. Berger MF, Hodis E, Hefferman TP, Deribe YL, Lawrence MS, Protopopov A, et al. Melanoma genome sequencing reveals frequent PREX2 mutations. Nature. 2012; 485(7399):502–6. Epub 2012/05/25. doi: 10.1038/nature11071 PMID: 22622578; PubMed Central PMCID: PMC3367798.
3. Hauschild A, Grob JJ, Demidov LV, Jouary T, Gutzmer R, Millward M, et al. Dabrafenib in BRAF-mutated metastatic melanoma: a multicentre, open-label, phase 3 randomised controlled trial. Lancet. 2012; 380(9839):358–65. Epub 2012/06/28. doi: 10.1016/S0140-6736(12)60868-X PMID: 22735384.
4. Slominski A, Paus R. Bomirski melanomas—a versatile and powerful model for pigment cell and melanoma research (review). Int J Oncol. 1993; 2(2):221–8. PMID: 21573540.

5. Chan J, Robinson ES, Atencio J, Wang Z, Kazianis S, Coletta LD, et al. Characterization of the CDKN2A and ARF genes in UV-induced melanocytic hyperplasias and melanomas of an opossum (Monodelphis domestica). Mol Carcinog. 2001; 31(1):16–26. PMID: 11398194.

6. Plonka PM, Slominski AT, Pajak S, Urbanska K. Transplantable melanomas in gerbils (Meriones unguiculatus). II: melanogenesis. Exp Dermatol. 2003; 12(4):356–64. PMID: 12930290.

7. Patton EE, Mitchell DL, Nairn RS. Genetic and environmental melanoma models in fish. Pigment cell & melanoma research. 2010; 23(3):314–37. Epub 2010/03/17. doi: 10.1111/j.1755-148X.2010.00693.x PMID: 20230482; PubMed Central PMID: PMC2881310.

8. Wittbrodt J, Shima A, Scharf M. Medaka—a model organism from the far East. Nat Rev Genet. 2002; 3(1):53–64. Epub 2002/02/02. doi: 10.1038/ng4074 [pii] PMID: 11823791.

9. Thernes V, Grabher C, Ristoratore F, Bournat F, Choulika A, Wittbrodt J, et al. I-SceI meganuclease mediates highly efficient transgenesis in fish. Mech Dev. 2002; 118(1–2):91–8. Epub 2002/09/28. SO925477302002186 [pii] PMID: 12351173.

10. Scharf M, Kneitz S, Wilde B, Wagner T, Henkel CV, Spanik HP, et al. Conserved expression signatures between medaka and human pigment cell tumors. PLoS One. 2012; 7(5):e37880. Epub 2012/06/14. doi: 10.1371/journal.pone.0037880 PONE-D-12-02914 [pii] PMID: 22693581; PubMed Central PMID: PMC3365055.

11. Dovey M, White RM, Zon LI. Oncogenic NRAS cooperates with p53 loss to generate melanoma in zebrafish. Zebrafish. 2009; 6(4):397–404. Epub 2009/12/04. doi: 10.1089/zeb.2009.0606 PMID: 19954345; PubMed Central PMID: PMC2943216.

12. Mione MC, Trede NS. The zebrafish as a model for cancer. Dis Model Mech. 2010; 3(9):517–23. Epub 2010/04/01. doi: 10.1242/dmm.004747 PMID: 20354112; PubMed Central PMID: PMC2931530.

13. Winnepenninckx V, Lazar V, Michiels S, Dessen P, Stas M, Alonso SR, et al. Gene expression profiling of primary cutaneous melanoma and clinical outcome. Journal of the National Cancer Institute. 2006; 98(7):472–82. Epub 2006/06/04. doi: 10.1093/jnci/dj630 PMID: 16597583.

14. Meierjohann S, Scharf M. From Mendelian to molecular genetics: the Xiphophorus melanoma model. Trends Genet. 2006; 22(12):654–61. Epub 2006/10/13. S0168-9525(06)00336-2 [pii] doi: 10.1016/j.tig.2006.09.013 PMID: 17034900.

15. Berger MF, Levin JZ, Vijayendran K, Sivachenko A, Adiconis X, Maguire J, et al. Integrative analysis of the melanoma transcriptome. Genome research. 2010; 20(4):413–27. Epub 2010/02/25. doi: 10.1101/gr.103697.109 PMID: 20179022; PubMed Central PMCID: PMC2847744.

16. Ngjio SF, von Scheidt B, Moller A, Smyth MJ, Teng MW. The interaction between murine melanoma and the immune system reveals that prolonged responses predispose to autoimmune. Oncoimmunology. 2013; 2(2):e23036. Epub 2013/03/26. doi: 10.4161/onci.23036 PMID: 23524369; PubMed Central PMCID: PMC3601172.

17. Scharf M, Wilde B, Laisney JA, Taniguchi Y, Takeda S, Meierjohann S. A mutated EGFR is sufficient to induce malignant melanoma with genetic background-dependent histopathologies. The Journal of investigative dermatology. 2010; 130(1):249–58. Epub 2009/07/18. doi: 10.1038/jid.2009.213 PMID: 19609310.

18. Kim D, Pertea G, Trapnell C, Pimentel H, Kelley R, Salzberg SL. TopHat2: accurate alignment of transcriptomes in the presence of insertions, deletions and gene fusions. Genome biology. 2013; 14(4):R36. doi: 10.1186/gb-2013-14-4-r36 PMID: 23618408; PubMed Central PMID: PMC4053844.

19. Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. Genome biology. 2014; 15(12):550. Epub 2014/12/18. doi: 10.1186/s13059-014-0550-8 PMID: 25516281; PubMed Central PMCID: PMC4302649.

20. Meyer A, Scharf M. Gene and genome duplications in vertebrates: the one-to-four (-to-eight in fish) rule and the evolution of novel gene functions. Current opinion in cell biology. 1999; 11(6):699–704. Epub 1999/12/22. PMID: 10607714.

21. Taylor JS, Braasch I, Frickey T, Meyer A, Van de Peer Y. Genome duplication, a trait shared by 22000 species of ray-finned fish. Genome research. 2003; 13(3):382–90. Epub 2003/03/06. doi: 10.1101/gr.640303 PMID: 12618368; PubMed Central PMID: PMC430266.

22. Kramer A, Green J, Pollard J Jr., Tugendreich S. Causal analysis approaches in Ingenuity Pathway Analysis. Bioinformatics. 2014. Epub 2013/12/18. doi: 10.1093/bioinformatics/btt703 PMID: 24336805.

23. Griffith M, Griffith OL, Coffman AC, Weible JV, McMichael JF, Spies NC, et al. DGIdb: mining the drug-gable genome. Nat Methods. 2013; 10(12):1209–10. Epub 2013/10/15. doi: 10.1038/nmeth.2689 PMID: 24122041; PubMed Central PMCID: PMC3851581.
24. Wittbrodt J, Adam D, Maitschek B, Mauler W, Rauf F, Telling A, et al. Novel putative receptor tyrosine kinase encoded by the melanoma-inducing Tu locus in Xiphophorus. Nature. 1989; 341(6241):415–21. Epub 1989/10/05. doi: 10.1038/341a0 PMID: 2797166.

25. Wellbrock C, Weisser C, Hassel JC, Fischer P, Becker J, Vetter CS, et al. STAT5 contributes to interferon resistance of melanoma cells. Curr Biol. 2005; 15(18):1629–39. Epub 2005/09/20. S0960-9822(05)00955-3 [pii]; doi: 10.1016/j.cub.2005.08.036 PMID: 16169484.

26. Laisney JA, Mueller TD, Scharl M, Meierjohann S. Hyperactivation of constitutively dimerized oncogenic EGF receptors by autocrine loops. Oncogene. 2013; 32(19):2403–11. Epub 2012/07/04. doi: 10.1038/onc.2012.267 PMID: 22751127.

27. Schaafhausen MK, Yang WJ, Centanin L, Wittbrodt J, Bosserhoff A, Fischer A, et al. Tumor angiogenesis is caused by single melanoma cells in a manner dependent on reactive oxygen species and NF-kappaB. Journal of cell science. 2013; 126(Pt 17):3862–72. Epub 2013/07/12. doi: 10.1242/jcs.125021 PMID: 23843609.

28. Anders S, McCarthy DJ, Chen Y, Okoniewski M, Smyth GK, Huber W, et al. Count-based differential expression analysis of RNA sequencing data using R and Bioconductor. Nat Protoc. 2013; 8(9):1765–86. Epub 2013/08/27. nprot.2013.099 [pii]; doi: 10.1038/nprot.2013.099 PMID: 23975260.

29. Ding L, Kim M, Kanchi KL, Lu C, Griffith M, et al. Clonal architectures and driver mutations in melanomas. PLoS One. 2014; 9(11):e111153. Epub 2014/11/14. doi: 10.1371/journal.pone.0111153 PMID: 25393105; PubMed Central PMCID: PMC4230926.

30. Bachellerie JP, Cavaille J, Huttenhofer A. The expanding snoRNA world. Biochimie. 2002; 84(8):775–80. Epub 2002/08/17. PMID: 12183442.

31. Yen J, White RM, Wedge DC, Van Loo P, de Ridder J, Capper A, et al. The genetic heterogeneity and mutational burden of engineered melanomas in zebrafish models. Genome biology. 2013; 14(10):R113. Epub 2013/10/24. doi: 10.1186/gb-2013-14-10-r113 PMID: 24148783; PubMed Central PMCID: PMC3983654.

32. Slominski A, Kim TK, Brozyma AA, Janjetovic Z, Brooks DL, Schwab LP, et al. The role of melanogenesis in regulation of melanoma behavior: melanogenesis leads to stimulation of HIF-1alpha expression and HIF-dependent attendant pathways. Arch Biochem Biophys. 2014; 563:79–93. doi: 10.1016/j.abb.2014.06.030 PMID: 24997364; PubMed Central PMCID: PMCPMC4221528.

33. Brozyna AA, Jozwicki W, Carlson JA, Slominski AT. Melanogenesis affects overall and disease-free survival in patients with stage III and IV melanoma. Hum Pathol. 2013; 44(10):2071–8. doi: 10.1016/j.humpath.2013.02.022 PMID: 23791398; PubMed Central PMCID: PMCPMC3783651.

34. Shimanovsky A, Jethava A, Dasanu CA. Immune alterations in malignant melanoma and current immunotherapy concepts. Expert opinion on biological therapy. 2013; 13(10):1413–27. Epub 2013/08/13. doi: 10.1517/14712598.2013.827658 PMID: 23990800.
43. Tucci M, Stucci S, Passarelli A, Giudice G, Dammacco F, Silvestris F. The immune escape in melanoma: role of the impaired dendritic cell function. Expert Rev Clin Immunol. 2014; 10(10):1395–404. doi: 10.1586/1744666X.2014.955851 PMID: 25225774.

44. Liu Q, Tomei S, Ascierio ML, De Giorgi V, Bedognetti D, Dai C, et al. Melanoma NOS1 expression promotes dysfunctional IFN signaling. The Journal of clinical investigation. 2014; 124(5):2147–59. Epub 2014/04/03. doi: 10.1172/JCI69651 PMID: 24691438; PubMed Central PMCID: PMC4001531.

45. Slominski A, Paus R, Mihm MC. Inhibition of melanogenesis as an adjuvant strategy in the treatment of melanotic melanomas: selective review and hypothesis. Anticancer Res. 1998; 18(5B):3709–15. PMID: 9854482.

46. Slominski A, Zmijewski MA, Pawelek J. L-tyrosine and L-dihydroxyphenylalanine as hormone-like regulators of melanocyte functions. Pigment cell & melanoma research. 2012; 25(1):14–27. doi: 10.1111/j.1755-148X.2011.00898.x PMID: 21834848; PubMed Central PMCID: PMCPMC3242935.

47. Maresh GA, Marken JS, Neubauer M, Aruffo A, Hellstrom I, Hellstrom KE, et al. Cloning and expression of the gene for the melanoma-associated ME20 antigen. DNA and cell biology. 1998; 18(5B):3709–15. PMID: 9854482.

48. Panelli MC, Riker A, Kammula U, Wang E, Lee KH, Rosenberg SA, et al. Expansion of tumor-T cell pairs from fine needle aspirates of melanoma metastases. Journal of immunology. 2000; 164(1):495–504. Epub 1999/12/22. PMID: 10605047.

49. Natarajan VT, Ganju P, Singh A, Vijayan V, Kirty K, Yadav S, et al. IFN-gamma signaling maintains skin pigmentation homeostasis through regulation of melanosome maturation. Proc Natl Acad Sci U S A. 2014; 111(6):2301–6. doi: 10.1073/pnas.1304988111 PMID: 24474804; PubMed Central PMCID: PMCPMC3926048.

50. Eggermont AM, Suciu S, Santinami M, Testori A, Kruit WH, Marsden J, et al. Adjuvant therapy with pegylated interferon alfa-2b versus observation alone in resected stage III melanoma: final results of EORTC 18991, a randomised phase III trial. Lancet. 2008; 372(9633):117–26. Epub 2008/07/16. doi: 10.1016/S0140-6736(08)61033-8 PMID: 18620949.

51. Morcinek JC, Weisser C, Geissinger E, Scharf M, Wellbrock C. Activation of STAT5 triggers proliferation and contributes to anti-apoptotic signalling mediated by the oncogenic Xmrk kinase. Oncogene. 2002; 21(11):1668–78. Epub 2002/03/16. doi: 10.1038/sj.onc.1205148 PMID: 11896598.

52. Hodis E, Watson IR, Kryukov GV, Arold ST, Imielinski M, Theurillat JP, et al. A landscape of driver mutations in melanoma. Cell. 2012; 150(2):251–63. Epub 2012/07/24. doi: 10.1016/j.cell.2012.06.024 PMID: 22817889; PubMed Central PMCID: PMCPMC3600117.

53. Coca S, Perez-Piqueras J, Martinez D, Colmenarejo A, Saez MA, Vallejo C, et al. The prognostic significance of intratumoral natural killer cells in patients with colorectal carcinoma. Cancer. 1997; 79(12):2320–8. Epub 1997/06/15. PMID: 9191519.

54. Galon J, Costes F, Sanchez-Cabo F, Kirilovsky A, Mlecnik B, Lagorce-Pages C, et al. Type, density, and location of immune cells within human colorectal tumors predict clinical outcome. Science. 2006; 313(5795):1960–4. Epub 2006/09/30. doi: 10.1126/science.1129139 PMID: 17008531.