Perturbation of Defense Pathways by Low-Dose Arsenic Exposure in Zebrafish Embryos

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BACKGROUND: Exposure to arsenic is a critical risk factor in the complex interplay among genetics, the environment, and human disease. Despite the potential for in utero exposure, the mechanism of arsenic action on vertebrate development and disease is unknown.

OBJECTIVES: The objective of this study was to identify genes and gene networks perturbed by arsenic during development in order to enhance understanding of the molecular mechanisms of arsenic action.

METHODS: We exposed zebrafish embryos at 0.25–1.25 hr postfertilization to 10 or 100 ppb arsenic for 24 or 48 hr. We then used total RNA to interrogate genome microarrays and to test levels of gene expression changes by quantitative real-time polymerase chain reaction (QPCR). Computational analysis was used to identify gene expression networks perturbed by arsenic during vertebrate development.

RESULTS: We identified a set of 99 genes that responded to low levels of arsenic. Nineteen of these genes were predicted to function in a common regulatory network that was significantly associated with immune response and cancer (p < 10^{-41}). Arsenic-mediated expression changes were validated by QPCR.

CONCLUSIONS: In this study we demonstrated that arsenic significantly down-regulates expression levels of multiple genes potentially critical for regulating the establishment of an immune response. The data also provide molecular evidence consistent with phenotypic observations reported in other model systems. Additional mechanistic studies will help explain molecular events regulating early stages of the immune system and long-term consequences of arsenic-mediated perturbation of this system during development.

KEY WORDS: arsenic, Comparative Toxicogenomics Database, embryonic development, gene regulatory networks, immunity, microarray, toxicogenomics, transcriptional profiling, zebrafish. Environ Health Perspect 117:981–987 (2009). doi:10.1289/ehp.900555 available via http://dx.doi.org/ [Online 22 February 2009]

Arsenic represents a global environmental health threat and a known human carcinogen. More than 100,000 individuals in New England are exposed to arsenic levels in drinking water that exceed federal standards [U.S. Environmental Protection Agency (EPA)], and 500 million people worldwide are at risk of exposure from contaminated groundwater (Mead 2005). Studies linking arsenic to adverse human health effects such as lung, bladder, and skin cancer were instrumental in motivating recent U.S. policy changes lowering maximum allowable limits in drinking water from 50 ppb to 10 ppb (U.S. EPA 2006). However, the effects of exposure to low levels of arsenic also remain unclear.

Exposure to very low levels of arsenic (<10 ppb) is linked to adverse biological effects, including endocrine disruption and alteration in cell cycle kinetics, cell signaling, and the proliferative response (Rossman et al. 2004). Arsenic levels commonly found in contaminated drinking water in the U.S. elicit changes in gene expression profiles in a number of critical gene networks (Andrew et al. 2007), alter the vertebrate innate immune response (Nayak et al. 2007), and interfere with DNA repair processes (Andrew et al. 2006). Genomic profiling studies have reported that moderate changes in low-level exposure conditions elicited different expression profiles, suggesting arsenic affects biological systems at low levels, and these effects are very complex (Andrew et al. 2003).

In a human study in Argentina, Concha et al. (1998) found fetal arsenic levels similar to levels detected in mothers exposed via drinking water, indicating that arsenic crosses the placenta. Arsenic can affect placental vasculogenesis and increase the rate of spontaneous abortions (Andrew et al. 2006; He et al. 2007); cause epigenetic modifications (Xie et al. 2007); and induce neural tube defects, cause axial skeletal abnormalities, and reduce mean fetal weight in transplacentally exposed mice without evidence of maternal toxicity (Hill et al. 2008). Gene expression studies in exposed mouse fetal liver cells recapitulate those associated with transplacental arsenic-induced mouse liver tumors (Liu et al. 2008).

Early environmental challenges can affect disease susceptibility later in life in the absence of apparent alterations in normal developmental programs (Hales and Barker 2003, 2004; Newbold et al. 2006; Vickers et al. 2007). Diseases postulated to have developmental windows of vulnerability include cancers and neurologic, reproductive, and metabolic disorders (Birnbaum and Fenton 2003; Davey et al. 2007; Ho et al. 2006; Lahiri et al. 2007; Weidman et al. 2007). In utero exposure in mice was associated with dose-related adult-onset liver cancer, adrenal cortical adenoma in male offspring, and ovarian tumors and lung cancer in female offspring (Shen et al. 2007; Waalkes et al. 2007). The molecular basis of these correlations is largely unknown.

Emerging evidence from evolutionarily divergent species suggests that the immune response is compromised by low levels of arsenic and likely reflects functional disruption of critical genes and networks. An overrepresentation of genes involved in immunosuppression was correlated with arsenic exposure (Andrew et al. 2008), and arsenic down-regulated genes involved in immune modulation in mouse embryonic cells (Yu et al. 2008). Low levels of arsenic also reduced immune and defense responses in birds and fish (Aggarwal et al. 2008; Andrew et al. 2008; Yu et al. 2008). Additional studies in a tractable model are needed to better understand the extent to which arsenic perturbs regulatory networks involved in immune response during development.

The zebrafish (Danio rerio) is an ideal model for studying developmental toxicology and for understanding the connections between environmental exposures and human diseases (Pichler et al. 2003). Zebrafish are highly fecund with short generation times of 3–5 months. Eggs are fertilized externally and are relatively large (0.6 mm), transparent, and readily manipulated (Mattingly et al. 2001; Pichler et al. 2003). Rapid development from a zygote to the hatching period (~48 hr) provides advantages over mammalian models for observing organogenesis. Zebrafish are significantly more economical to maintain than are mammalian models, enabling more experimental opportunities (Pichler et al. 2003).

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Zebrafish resources are abundant, including a sequenced genome, microarray resources, and cDNA libraries. Despite the evolutionary distance separating zebrafish and humans (~400 million years), it is an important model of human diseases such as craniofacial, hematopoietic, cardiovascular, and neurodegenerative disorders and cancer (Goessling et al. 2007; Lieschke and Currie 2007). Large-scale genetic screens have produced zebrafish mutants with phenotypes similar to human disorders (Goessling et al. 2007; Lieschke and Currie 2007). The combination of these features makes the zebrafish uniquely suited for investigating the effects of environmental exposures and the consequences on vertebrate development.

In this study we aimed to identify genes and networks targeted by low levels of arsenic during vertebrate embryonic development. We exposed zebrafish embryos to arsenic; evaluated effects on transcription by whole genome microarray analysis, gene enrichment, and pathway analyses; and confirmed results by quantitative reverse transcriptase (RT) polymerase chain reaction (QPCR). Here we describe the effects of arsenic on the expression levels of a cohort of genes predicted to affect multiple vertebrate immune response processes.

Materials and Methods

Zebrafish. Wild-type AB zebrafish were maintained on a 14/10-hr light/dark cycle at the zebrafish facility of the University of Maine–Orono in a recirculating system with water at 28.5°C and a flow rate of 150 L/min. Animal husbandry was in accordance with the University’s Institutional Animal Care and Use Committee standards. Animals were treated humanely and with regard for alleviation of suffering.

Arsenic exposures. Zebrafish adults were spawned; embryos were then collected, scored for viability, and sorted by developmental stage. We used embryos between 2-cell and 16-cell stage [0.25–1.5 hr postfertilization (hpf)] exclusively for the studies described here.

For microarray analysis, we exposed pooled samples of 50 zebrafish to 0, 10, or 100 ppb sodium meta-arsenite (NaAsO₂; As³⁺; Fluka, St. Louis, MO) up to 48 hpf. These concentrations are environmentally relevant and do not cause toxicity. All exposures were in 0.5 x E2 media (7.5 mM NaCl, 0.25 mM KCl, 0.5 mM MgSO₄, 0.075 mM KH₂PO₄, 0.0025 mM Na₂HPO₄, 0.5 mM CaCl₂, and 0.35 mM NaHCO₃). Embryos were kept at a density of 3–6/µL with one medium change per 24 hr. All exposures and controls were performed in triplicate in plastic Petri dishes at 28.5°C.

For QPCR analysis, we exposed pooled samples as described above, except that embryos were exposed at 24 hr. Four or five biological replicates were performed in plastic Petri dishes at 28.5°C.

RNA extraction and reverse transcription. Total RNA was recovered from whole embryos using TRIzol (Invitrogen, Carlsbad, CA) according to the manufacturer’s protocol. We assessed RNA used for microarray and QPCR experiments by microchip analysis on an Agilent 2100 Bioanalyzer (Agilent Technologies, Foster City, CA). First-strand cDNA was synthesized from 2 µg of total RNA using Moloney murine leukemia virus RT from a RETROscript kit (Ambion, Austin, TX) according to the manufacturer’s protocol. Total RNA without RT were used as negative controls.

Microarray analysis. We performed microarray analysis to evaluate the effects of arsenic exposure on global transcription during zebrafish development. Total RNA (~10 µg per biological replicate) was sent to the Affymetrix Core Facility of the Oregon Health and Sciences University (Portland OR), where microarray analyses were performed in triplicate on Affymetrix GeneChip Zebrafish Genome arrays consisting of 14,900 transcripts, for a total of nine independent arrays: three controls, three low-concentration As³⁺ (10 ppb) exposures, and three high-concentration As³⁺ (100 ppb) exposures.

Sample labeling. We converted RNA to double-stranded cDNA using Superscript Reverse Transcriptase (Invitrogen) and an oligo-dT primer linked to a 17 RNA polymerase binding site sequence. Amplified and labeled cRNA (‘target’) was produced by in vitro transcription using T7 RNA polymerase, biotin-UTP, and biotin-CTP (Enzo Diagnostics, Inc., Farmingdale, NY). Target yield was measured by ultraviolet absorbance (A260).

Array hybridization and processing. We fragmented labeled target at 95°C in the presence of high [Mg²⁺] and combined with bio-tinylated hybridization control oligomer and biotinylated control cRNAs for BioB, BioC, BioD, and Cnx (Affymetrix) in hybridization buffer. We hybridized 10 µg of target with the arrays overnight, followed by washing, staining with streptavidin-phycocerythrin (Molecular Probes, Carlsbad, CA), signal amplification

Table 1. Genes and primers analyzed by QPCR.

| Gene symbol (name)³ | Fold change² | Gene accession ID | Primer sequence | Amplicon (bp) |
|---------------------|-------------|------------------|----------------|--------------|
| akt2 (v-akt murine thymoma viral oncogene homolog 2) | −2.4 | NC | NM_198146 | 5′-GAGATCAAGCTGTCGTACAGA-3′ (f) |
| asst1 (argininosuccinate synthetase 1) | −2.0 | NC | NM_001004603 | 5′-GAGGATCAGATATCGTGGAG-3′ (f) |
| C3 (similar to complement C3) | NC | −2.0 | XM_001343352 | 5′-GCTGTCGACCTTCTCAAA-3′ (f) |
| fn1 (fibronectin 1) | −1.8 | −1.7 | NM_131520 | 5′-TGCCAGGACCTGTTGTT-3′ (f) |
| foxo5 (forkhead box O5) | −1.9 | NC | NM_131085 | 5′-GTCAACCATCGATTGTGAT-3′ (f) |
| gapdh (glyceraldehyde-3-phosphate dehydrogenase) | Norm | Norm | NM_00111514 | 5′-TGGGGGTGAAAGAAGA-3′ (f) |
| nkfb2 (nuclear factor of kappa light polypeptide gene enhancer in B-cells 2) | −2.3 | NC | NM_001001840 | 5′-CACGCCTCAAGAAGATGAT-3′ (f) |
| notch1a (notch homolog 1a) | −2.3 | −2.0 | NM_131441 | 5′-TGAATGGAGACCTGTTGTT-3′ (f) |
| notch1b (notch homolog 1b) | −2.0 | −2.0 | NM_131302 | 5′-GCTATGGACCTGTTGTT-3′ (f) |
| pik3r1 (similar to phosphoinositide-3-kinase, regulatory subunit, polypeptide 1) | −1.9 | −1.7 | XM_67872 | 5′-TCACTTCCTCAAGATGTT-3′ (f) |
| plg (plasminogen) | NC | NC | NM_201472 | 5′-AAACAAGAATTCGATGTT-3′ (f) |

Abbreviations: F, forward primer; NC, no change; Norm, normalizer; R, reverse primer.

²Fold change in gene expression between As³⁺ (10 or 100 ppb) and control (0 ppb) experiments as determined by QPCR analysis. FC = −(2 ΔCt) where ΔCt = ΔCtexposed − ΔCtcontrol. ΔCtexposed and ΔCtcontrol are the normalized threshold cycles for exposed and control samples, respectively. Official zebrafish gene symbols and gene accession numbers from NCBI (2009).
with biotinylated anti-streptavidin antibody (Vector Laboratories, Burlingame, CA), and a final staining step on the Fluidics Station 400 (Affymetrix). The distribution of fluorescent material on the processed array was determined using the Affymetrix GeneChip laser scanner; image inspection was performed manually. We created Affymetrix data files [cell intensity (CEL) files] from each array using GCOS (GeneChip Operating Software; Affymetrix).

**Statistical analysis.** We implemented a statistical process similar to the one reported by Gosse et al. (2008). Probe-level data from CEL files were normalized using robust multiarray analysis (Irizarry et al. 2003) as implemented in Bioconductor (http://www.bioconductor.org). Quality control was performed using log-intensity (CEL) files from each array using GCOS (GeneChip Operating Software; Affymetrix).

We created image files using the Affymetrix GeneChip laser scanner; material on the processed array was determined using the Affymetrix GeneChip laser scanner; final staining step on the Fluidics Station 400 (Vector Laboratories, Burlingame, CA), and a standard curve and primer efficiencies were determined for all genes analyzed by QPCR. Primer sequences are listed in Table 1.

**Results**

**Arsenic-mediated perturbation of gene transcription.** We implemented several strategies during the statistical analysis of the microarray data sets. First, an uncorrected t-test identified 766 genes differentially expressed between control and treated samples (p < 0.05). However, data sets composed of > 15,600 measurements are expected to yield > 770 measurements by chance with p < 0.05. After correcting for multiple hypothesis testing (Bonferroni correction), none of the genes rose to the level of significance. Other parametric tests performed similarly. Second, we implemented a nonparametric fold-change ranking approach as recommended by the MAQC Consortium (Guo et al. 2006; Shi et al. 2006) by which we selected genes with a) at least a ± 1.4-fold change between control and exposed data sets and b) substantial within-group consistency as evidenced by a two-tailed t-test p-value ≤ 0.1. This approach yielded 99 differentially expressed genes, of which 55 had an uncorrected p-value ≤ 0.05. This refined data set was hierarchically clustered to produce the heat map shown in Figure 1.

**Figure 1.** Hierarchical clustering of transcripts significantly modified by exposure to arsenic for 48 hr. Transcripts were selected by a fold-change ranking approach implemented in R (R Development Core Team 2009) using an absolute fold change of 1.4 and a threshold (p-value) of 0.1. Each column represents the expression level of a probe set in a pooled group of 50 animals (green and red indicate decreased and increased expression, respectively). Arrays are grouped by arsenic concentration (control (0 ppb), 10 ppb, and 100 ppb). Transcripts were clustered by hierarchical clustering using the complete linkage algorithm and Pearson correlation metric in R.
revealing a pattern suggesting concentration-dependent effects even though corrected p-values for this ranked set did not rise to the significance level of ≤ 0.05. Control (Figure 1, columns 1 and 2) and high-As\(^{3+}\)-exposed (Figure 1, columns 6–8) biological replicates exhibited consistent reproducibility among replicates, whereas low-As\(^{3+}\)-exposed replicates (Figure 1, columns 3–5) did not. For example, biological replicates 1 and 2 of the 10-ppb As\(^{3+}\) set (columns 3 and 4) were consistent with each other, whereas replicate 3 (column 5) exhibited a response in line with the biological replicates treated with 100-ppb As\(^{3+}\).

**Table 2.** Immune processes enriched with genes perturbed by arsenic.

| GO ID\(^a\) | GO process                      | Gene count | p-Value      | Human genes\(^b\)                        |
|-------------|---------------------------------|------------|--------------|-----------------------------------------|
| GO:0002253  | Activation of immune response   | 4          | 3.97 × 10\(^{-3}\) | C3, C4B, C9, CFHR1                     |
| GO:0006952  | Defense response                | 8          | 1.81 × 10\(^{-2}\) | C3, C4B, C9, CFHR1, FN1, IN5, P2RY1, TCF1 |
| GO:0009605  | Response to external stimulus   | 8          | 3.07 × 10\(^{-2}\) | A2M, C3, C4B, C9, CFHR1, FN1, INS, PLG  |
| GO:0006950  | Response to stress              | 10         | 6.45 × 10\(^{-2}\) | APOA4, C3, C4B, C9, CFHR1, CSNK1, EPAS1, FN1, INS, PLG |
| GO:0006955  | Immune response                 | 9          | 7.12 × 10\(^{-2}\) | C3, C9, C4B, CFHR1, CT5, G2A1, NFX2, NOTCH1, TCF12 |
| GO:0002520  | Immune system development       | 4          | 8.25 × 10\(^{-2}\) | FOX03, NFX2, NOTCH2, PIK3R1             |

\(^a\)Data from The Gene Ontology (2009). \(^b\)Data from NCBI (2009).

**Computational prediction of an arsenic-modulated molecular network affecting immune response.** We used IPA to identify molecular relationships among genes predicted by microarray analysis to be differentially expressed in response to arsenic. In addition, noninput molecules are inserted by IPA in order to merge small networks generated from user data. We identified orthologous human genes for 79 of the 99 differentially expressed zebrafish genes described above. Among these, 64 were in the IPA knowledgebase. IPA identified a highly significant network (p < 10\(^{-41}\)) containing 20 of the 64 (31%) input genes and 15 bridging genes (Figure 2). This network was significantly associated with immune response, cancer, and gastrointestinal disease (p ≤ 0.02).

Genes within this network are involved in specific immune functions such as complement activation (p < 2.8 × 10\(^{-46}\)), migration of immune response cells (e.g., monocytes, macrophages; p < 9.2 × 10\(^{-39}\)), and respiratory burst (p < 1.2 × 10\(^{-3}\)). The IPA-derived network also contained three genes (cugbp2, foxo5, and pik3r1) predictive of prenatal arsenic exposure based on a recent epidemiologic study (Fry et al. 2007), two of which (foxo5 and pik3r1) were observed in the 99 genes identified by the microarray analysis. These results support the hypothesis that the zebrafish is a valuable model organism for understanding the complex mechanisms of arsenic action during vertebrate development.

IPA results were corroborated by a gene ontology enrichment analysis conducted using the online Functional Annotation Clustering tool from DAVID (Dennis et al. 2003). Among the 79 differentially expressed genes with identifiable human orthologs, the immune response biological process [Gene Ontology (GO) ID no. GO:0006955 (The Gene Ontology 2009)] was significantly overrepresented (p < 0.05). Analysis using the same 79 genes from the microarray analysis plus the 15 bridging genes from IPA yielded a subset of GO terms, 53% of which were associated with the immune system or defense responses (Table 2).

**CTD analysis.** CTD contains a robust data set for arsenic that describes molecular interactions between 20 different arsenic compounds and 1,709 genes and proteins (Davis et al. 2008). Comparison of the 79 differentially expressed genes from the microarray analysis with curated arsenic-interacting genes in CTD yielded an overlapping set of 11 genes. Among these 11 genes, 5 were members of the predicted pathway (Figure 2, circled genes). CTD also corroborated 5 of the 15 bridging genes inserted by IPA (Figure 2, boxed genes). These genes were present on the microarrays but were not included in the top 99 genes identified by nonparametric rank-based analysis.
CTD also supported arsenic-associated GO terms and diseases identified by DAVID and IPA. The most common disease categories associated with arsenic in CTD include immune system diseases, neoplasms, nervous system diseases, skin diseases, digestive system diseases, and metabolic disorders (Davis et al. 2008). Results from IPA network analysis, microarray analysis, and data mining of CTD converged, reinforcing the potential effects of arsenic on this predicted network.

**Quantitative analysis of transcript levels of IFA network genes.** We analyzed a total of 10 genes within the network by QPCR: 6 derived from the microarray study (C3, fn1, foxo5, notch1a, notch1b, and plg; Figure 2); 2 derived from the microarray study and corroborated by CTD (ass1, pik3r1), and 2 bridging genes corroborated by CTD (aks2, nfkb2). We selected a 24-hr time point in order to analyze the effects of As3+ on genes potentially regulating the immune response during developmental stages earlier than the one analyzed by microarray. As shown in Figure 3 and Table 1, the genes selected for this analysis were down-regulated by As3+. Remarkably, this gene set exhibited a more robust and significant response to the lowest levels of As3+ (10 ppb) compared with responses to 10-fold higher levels of As3+ (9 of 11 genes vs. 6 of 11). Only 4 of the 11 genes tested showed similar responses to As3+ at both concentrations: fn1, notch1a, notch1b, and pik3r1. One gene, plg, had no expression change in control versus treated embryos, even though it was identified through microarray analysis to be affected by As3+.

We identified canonical pathways associated with these nine arsenic-responsive genes using IPA (Figure 4). Many of the pathways were involved in immune function, the most significant being acute-phase response signaling. Other pathways included lympho toxin receptor signaling, interleukin signaling, CD28 signaling, Trem1 (triggering receptor expressed on myeloid cells 1) signaling, and the complement system. Collectively, gene ontology and pathway enrichment analyses of microarray and QPCR results demonstrate that arsenic perturbs genes and networks that are involved in the immune response during vertebrate development.

**Discussion**

In this article we study the effect of As3+ exposure on the transcription profile of zebrafish embryos at 24 and 48 hpf by QPCR and microarray analysis, respectively. QPCR results demonstrate statistically significant down-regulation of multiple genes with critical functions in immune system development and, by inference, immunity at 24 hpf. In contrast, microarray analysis performed at 48 hpf reveals the opposite pattern: Genes down-regulated at 24 hpf are up-regulated at 48 hpf in the presence of As3+. The microarray analysis provided a robust target gene set that underwent changes in expression due to As3+ exposure, even though the set was below the threshold of significance as determined by statistical analysis. This observation underscores the need to take into account that low-dose exposures may not result in dramatic changes in gene expression that stand up to rigorous statistical tests but nevertheless contribute to significant biological effects that can be validated (Gosse et al. 2008).

The genes we evaluated by QPCR showed expression changes for at least one exposure concentration and in many cases for both. For example, fn1, notch1a, notch1b, and pik3r1 were down-regulated at least 1.7-fold by both 10 and 100 ppb As3+ compared with controls, whereas aks2, ass1, and nfkb2 responded significantly to 10 ppb As3+ (fold change ≥ 2) but not to 100 ppb As3+. Conversely, C3 was down-regulated 2-fold by 100 ppb As3+ but showed no effect to 10 ppb As3+ exposures. This variable As3+ concentration-dependent phenomenon has been observed before. In a study by Bodwell et al. (2004), the effects of As3+ on activation of a reporter gene driven by a glucocorticoid-regulated promoter were stronger at low concentrations. In fact, they observed 75% of the maximal induction at 6 ppb As3+, with a peak at 60 ppb As3+, dropping off rapidly at higher concentrations to below basal levels of expression.

In the present study, IPA analysis identified a gene network containing 20 of 79 genes
identified by microarray analysis (Figure 2). We found two genes, foxo5 (the zebrafish ortholog of human FOXO3A) and pik3r1, in a group of 170 "sentinel" genes reported as predictive of transplacental As3+ exposure in infants (Fry et al. 2007). Additionally, bridging genes inserted into the network by IPA included akt2, cugbp2, nfkbia, and pi3k. The genes akt2, nfkbia, and pi3k are network hubs linking most genes within the set of 20 genes, and akt2 and pi3k are members of the Akt/P3K signal transduction axis important for insulin function, whereas nfkbia is required for immune processes (Speirs et al. 2004). Fry et al. (2007) identified nfkbia as one of two hub genes connecting a subnetwork enriched with proinflammatory genes that were arsenic responsive. The concordance of arsenic-mediated effects in humans and zebrafish reinforces the immune system’s ability to elucidate the effects of As3+ exposure during embryogenesis.

Epide miolog ic studies have linked exposure to inorganic arsenic (As3+) to multiple human diseases (Hughes 2006). Several of these diseases, most notably cancer, might be rooted in abnormalities seen in infants (Fry et al. 2007). Additionally, bridging genes inserted into the network by IPA included akt2, cugbp2, nfkbia, and pi3k. The genes akt2, nfkbia, and pi3k are network hubs linking most genes within the set of 20 genes, and akt2 and pi3k are members of the Akt/P3K signal transduction axis important for insulin function, whereas nfkbia is required for immune processes (Speirs et al. 2004). Fry et al. (2007) identified nfkbia as one of two hub genes connecting a subnetwork enriched with proinflammatory genes that were arsenic responsive. The concordance of arsenic-mediated effects in humans and zebrafish reinforces the immune system’s ability to elucidate the effects of As3+ exposure during embryogenesis.

These results establish a link between arsenic and physiological inflammation disorders or activation of the adaptive immune system in humans and rodents chronically exposed to As3+ (Fry et al. 2007; Straub et al. 2007; Vahidinia et al. 2007; Wu et al. 2003). These results establish a link between As3+ exposure and immune disorders, yet with the exception of work by Nayak et al. (2007), they do not indicate whether the effects are restricted to disrupting the function of specific populations of differentiated cells or if exposure at early stages of development leads to abnormal immune responses. Nayak et al. (2007) measured a significant decrease in expression of several immune system modulators, including member of the interleukin family, juvenile fish previously exposed to low levels of As3+.

However, insight into arsenic-mediated effects on the embryonic transcriptome requires more attention in order to gain greater insight into the effects of arsenic on the immune system and associated disorders.

Conclusion

This study shows that the zebrafish is a valuable model organism to enhance understanding of the effects of arsenic on aspects of embryology, including networks affecting immune development. The networked genes uncovered by this study are highly evolutionarily conserved at various levels, including conservation of chromosomal synteny (data not shown). In addition, there is significant overlap between the results of this study and observations made by others. These data enhance understanding about developmental responses induced by arsenic and provide novel insight into the molecular actions of arsenic on the immune system.

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