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Exposure to inorganic arsenic increases the risk of basal cell carcinoma (BCC). Arsenic metabolism is a susceptibility factor for arsenic toxicity, and specific haplotypes in arsenic (+3 oxidation state) methyltransferase (AS3MT) have been associated with increased urinary fractions of the most toxic arsenic metabolite, methylarsonic acid (MMA). The aim of this study is to elucidate the association of AS3MT haplotypes with arsenic metabolism and the risk of BCC. Four AS3MT polymorphisms were genotyped in BCC cases (N = 529) and controls (N = 533) from Eastern Europe with low to moderate arsenic exposure (lifetime average drinking water concentration: 1.3 mg/L, range 0.01–167 mg/L). Urinary metabolites [inorganic arsenic (iAs), MMA, dimethylarsinic acid (DMA)] were analyzed by HPLC-ICPMS. Five AS3MT haplotypes (based on rs3740400 A/C, rs3740393 G/C, rs11191439 T/C and rs1046778 T/C) had frequencies >5%. Individuals with the CCTC haplotype had lower %iAs (P = 0.032) and %MMA (P = 0.020) in urine, and higher %DMA (P = 0.033); individuals with the CGCT haplotype had higher %MMA (P < 0.001) and lower %DMA (P < 0.001). All haplotypes showed increased risk of BCC with increasing arsenic exposure through drinking water (ORs 1.1–1.4, P values from <0.001 to 0.082), except for the CCTC haplotype (OR 1.0, CI 0.9–1.2, P value 0.85). The results suggest that carriage of AS3MT haplotypes associated with less-efficient arsenic methylation, or lack of AS3MT haplotypes associated with a more-efficient arsenic methylation, results in higher risk of arsenic-related BCC. The fact that AS3MT haplotype status modified arsenic metabolism, and in turn the arsenic-related BCC risk, supports a causal relationship between low-level arsenic exposure and BCC. Environ. Mol. Mutagen. 56:60–69, 2015.

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Key words: polymorphism; methylarsonic acid; dimethylarsinic acid; metabolism
INTRODUCTION

Inorganic arsenic occurs globally in drinking water, particularly ground water, and elevated exposure has been associated with increased risk of nonmelanoma skin cancers, including basal cell carcinoma (BCC) [Hughes et al., 2011; Karagas et al., 2001; IARC, 2012; Leonardi et al., 2012]. Humans, however, show marked variation in susceptibility to arsenic toxicity. The main known susceptibility factors are gender, nutrition, and the efficiency of metabolizing arsenic [Vahter, 2002]. Inorganic arsenic (iAs) is methylated in the body to methylarsonic acid (MMA) and dimethylarsinic acid (DMA). In humans, like most mammals, efficient methylation from iAs to DMA is associated with decreased reactivity and increased rate of urinary arsenic excretion [Vahter, 2002; Gardner et al., 2011]. Individuals and populations, however, show major differences in arsenic metabolism. Incomplete and less-efficient arsenic metabolism, with higher urinary fractions of iAs and MMA and lower fractions of DMA, seems to be a marker of sensitivity to certain arsenic-related diseases, including cancer [Lindberg et al., 2008b; Chung et al., 2009].

Genetic variants in the main arsenic-methylating enzyme, arsenic (+3 oxidation state) methyltransferase (AS3MT), have been shown to modify arsenic metabolism in several populations [Sumi and Himeno, 2012]. For eight AS3MT single nucleotide polymorphisms (SNPs) combined into haplotypes, four AS3MT haplotypes had similar effects on the arsenic metabolite pattern in two different populations, one from Argentina and one from Bangladesh [Engstrom et al., 2011]. Associations between AS3MT polymorphisms and toxic effects, such as skin lesions, have also been reported [Valenzuela et al., 2009; Pierce et al., 2012]. Our hypothesis is that AS3MT haplotypes that are associated with the arsenic metabolite pattern also are associated with the arsenic-related risk of BCC.

MATERIALS AND METHODS

Study Population

Within a European case–control study that previously identified arsenic exposure in drinking water to be associated with risk of BCC [Leonardi et al., 2012] we evaluated the impact of AS3MT haplotypes on: (1) the urinary pattern of arsenic metabolites (controls only) and (2) the association between lifetime arsenic concentration and BCC risk (cases and controls). BCC cases (N = 529) and controls (N = 533) were recruited as part of a large case–control study designed to evaluate the risk of various cancers in relation to low environmental arsenic exposure in Hungary, Romania, and Slovakia between 2002 and 2004 [Leonardi et al., 2012]. Study areas were defined as certain counties (total population 2.8 million) in Hungary (Bač, Békés, Csongrád, and Jász-Nagykun-Szolnok), Romania (Bihor and Arad) and Slovakia (Banská Bystrica and Nitra) [Lindberg et al., 2006]. These areas were selected because of past, and in some places ongoing, arsenic exposure through drinking water, including public water supplies. Exposure assessment was based on measurements of water supplies in the study, and data from routine monitoring of water in the past [Hough et al., 2010]. Overall, 25% of the population had a lifetime average water concentration above 10 μg/L and 8% above 50 μg/L [Hough et al., 2010]. The large majority of the study participants were of European ancestry.

Table I shows descriptive data of the study subjects. Incident cases of BCC (ICD-10 code C44) were identified and invited on the basis of histopathological examinations by pathologists. These were compared with hospital-based controls (general surgery, orthopedic and trauma patients with, for example, appendicitis, abdominal hernias, duodenal ulcers, cholelithiasis, and fractures), which were selected and included in the study based on a published set of criteria [Leonardi et al., 2012]. They were frequency-matched with the cases for age, gender, and county of residence. Patients with malignant tumors, diabetes or cardiovascular diseases were excluded as study controls. Cases and controls were included if they had resided in the study area for at least 1 year during their lifetime. Clinicians and pathologists were blinded to the exposure status of cases and controls.

Cases and controls were interviewed by trained personnel and they completed a general questionnaire, which included information on detailed residential history focused on identification of drinking water sources at home, individual cumulative sun exposure in summer, skin complexion, effects of sun exposure on skin and age(s) at diagnosis of BCC. Skin sensitivity to burns (effects of sun exposure) is an index based on the reported intensity of the cutaneous reaction to 1 hr midday sun exposure of the upper trunk, and skin complexion as the self-reported complexion of the skin as light, medium or dark. Clinicians collected venous blood and urine samples from cases and controls after the study participants had signed consent forms. The blood samples were kept frozen at −80°C until analysis. Further details about the study populations and sampling have been described previously [Lindberg et al., 2006; Thirumaran et al., 2006; Leonardi et al., 2012; Surdu et al., 2013].

Ethical approval was obtained from the Ethical Committee of the National Health Research Council and the Regional Ethical Committee of the Szent-Györgyi Albert University of Szeged (Hungary), from local hospitals and Public Health Departments (Romania), and from Ethical Committees established in hospitals and State Health Institutes (Slovakia).

Arsenic Exposure Assessment

All water samples were analyzed by one laboratory, using hydride generation-atom absorption spectrometry (HG-AAS), with a limit of detection (LOD) of 0.2 μg/L [Hough et al., 2010]. The exposure assessment details are described in [Hough et al., 2010]. In short, for each individual, the concentrations of arsenic in drinking water at addresses over their lifetime were derived from measurements at the time of the study and historical data provided by national water authorities. If historical data were not available, concentrations of arsenic in village, and private wells were assumed to have been the same as the currently measured concentrations. The lifetime average arsenic concentration (μg/L) constitutes the time-weighted average concentration of arsenic in residential drinking water over all the years with nonmissing arsenic concentrations over a participant’s lifetime. This variable was used as a measure of lifetime arsenic exposure in this study. Ten individuals (four controls and six cases) had missing data for lifetime average arsenic concentration and were thus excluded from the BCC analyses.

Arsenic Metabolism

Urine samples were collected for determination of individual arsenic metabolite patterns. Only urine samples from the controls were included in this study, as disease status may affect arsenic metabolism. Each participant was asked to provide an early morning urine sample. The samples were frozen at −20°C within 15 min of collection. Urinary
metabolites iAs, MMA, and DMA were measured with HPLC online with inductively coupled plasma-mass spectrometry (ICPMS) [Lindberg et al., 2006]. Total urinary arsenic metabolites (U-As), computed as the sum of iAs, MMA, and DMA, were used as a measure of the current exposure to iAs. The fractions (%) of the different metabolites in urine were used to assess the efficiency of arsenic metabolism. Concentrations of arsenic in urine were adjusted to the average specific gravity of the urine (1.018 g/mL) in the study population. There was a cluster of values close to 100% DMA at the lowest water arsenic concentrations, which likely was due to contribution of DMA from certain foods [Lindberg et al., 2006]. Additionally, there were uncertainties in the speciation data of urine samples with very low concentrations, as DMA is often the only species well above the limit of detection. Because such uncertainties would invalidate the urinary arsenic metabolite pattern as marker of arsenic methylation efficiency, we decided to include only individuals with total urinary arsenic metabolite concentrations >5 g/L (N = 285) in the genetic association analyses. Another eight individuals were excluded because they had high urinary arsenic concentrations (above 20 µg/L) and low water arsenic concentrations (below 1 µg/L) in combination with a high percentage of DMA in the urine (above 80%), suggesting food as a major source of DMA. The sample size after these exclusions was 277 individuals. For %iAs and %MMA, the numbers were somewhat lower (N = 269 and 245, respectively) since individuals with fractions of 0% were excluded when natural log (ln) transforming the data.

**Genotyping**

Several SNPs in AS3MT have been associated with arsenic metabolite pattern [review in Sumi and Himeno, 2012]. To evaluate if genetic variation in AS3MT influences the risk of arsenic-related BCC risk, we genotyped four SNPs, all of which were present in haplotypes that we previously have shown are associated with arsenic metabolism in populations from Argentina and Bangladesh [Engström et al., 2011]. The selection of SNPs to be genotyped was based on the following observations. Rs11191439 is a nonsynonymous SNP and carriers of the C-allele (coding for Thr) have higher %iAs, higher %MMA and lower %DMA compared to TT carriers [Drobna et al., 2004; Lindberg et al., 2007; Hernandez et al., 2008; Engstrom et al., 2011], i.e., less efficient arsenic metabolism. Carriers of the intronic rs3740400 A- and rs3740393 G-alleles, as well as rs1046778 (located in the 3'-untranslated region)
Haplotypes were constructed on the combination of these four SNPs (rs11191439, rs3740400, rs3740393, and rs1046778) were geno-
typed by allelic discrimination with Taqman assays (Applied Biosystems). DNA was isolated from peripheral blood using Qiagen mini-
preparation kits (Qiagen GmbH, Hilden, Germany). The four AS3MT SNPs (rs11191439, rs3740400, rs3740393, and rs1046778) were geno-
typed by allelic discrimination with Taqman assays (Applied Biosystems Foster City, CA) on an ABI 7900 instrument (Applied Biosystems).

The genomic region where AS3MT polymorphisms, Arsenic Metabolism, BCC Risk

### Statistical Analyses

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were combined regardless of direction (e.g., efficient or nonefficient methylator).

RESULTS

Genetic Background

All SNPs were in Hardy–Weinberg equilibrium. The LD values ($R^2$) were similar in all countries (Fig. 1) and to those of the Hapmap European reference population (CEPH, derived from US residents with northern and western European ancestry (CEU) [Thorisson et al., 2005]). Rs1046778 was in moderate LD with both rs3740400 and rs3740393 ($R^2$ = between 0.36 and 0.46, the C-alleles of both SNPs were associated to each other), whereas rs11191439 was not in LD with any SNP.

Fig. 1. LD values ($R^2$) for the AS3MT SNPs in the different study populations as well as the reference CEU population from Hapmap. SNPs are shown in the 5’ to 3’ direction. CEU denotes the reference Hapmap population (CEPH, derived from US residents with northern and western European ancestry, http://www.hapmap.org, no data available for rs3740400) [Thorisson et al., 2005].
Seven haplotypes were found in this population, five of which had frequencies of >5% and were included in the subsequent statistical analyses (Table II). The AGTT (denoted haplotype 1) was the most common (SNPs are in 5′–3′ order based on their position on the gene: rs3740400, rs3740393, rs11191439, and rs1046778).

**AS3MT Haplotypes and the Metabolism of Arsenic**

The control individuals included in the assessment of associations between AS3MT haplotype and arsenic metabolizing phenotype had a geometric mean U-As concentration of 15.0 μg/L (N = 277, range 5.0–140 μg/L), and geometric mean fraction of iAs of 8.5% (N = 269, range 1.1–88%), MMA of 15% (N = 245, range 1.5–63%) and DMA of 72% (arithmetic mean) (range 9.8–100%).

Two haplotypes, 2 and 4, were significantly associated with the arsenic metabolite pattern (Table III). Individuals with one or two copies of haplotype 2 had lower %iAs (P = 0.032), lower %MMA (P = 0.020) and higher %DMA (P = 0.033). Individuals with one or two copies of haplotype 4 had particularly high %MMA (P < 0.001) and low %DMA (<0.001). No statistically significant effects were seen for the most frequently occurring haplotype 1.

When combining the haplotypes that influenced arsenic metabolite pattern (haplotypes 2 and 4) (Table III) we found that individuals with one or two copies of haplotype 2 and zero copies of haplotype 4 had the lowest %iAs (P = 0.032) and %MMA (P < 0.001), and the highest %DMA (P < 0.001). Conversely, individuals with zero copies of haplotype 2 and one or two copies of haplotype 4 had the highest %iAs and %MMA, and the lowest %DMA. Individuals with zero copies of both haplotypes 2 and 4 had somewhat lower %MMA and higher %DMA than individuals with one or two copies of both haplotypes 2 and 4. However, the group with one or two copies of both haplotypes 2 and 4 comprised only 11 or 12 individuals so the results for this group have low power.

**AS3MT Haplotypes and Arsenic-Related BCC Risk**

Based on cases and controls, the risk of BCC increased with increasing lifetime concentration of arsenic in drinking water (OR 1.2 per 10 μg/L increase, 95% CI 1.1–1.3), as previously presented [Leonardi et al., 2012]. Results of the logistic regression analysis of haplotype-specific associations between lifetime water arsenic concentrations and BCC are shown in Table IV. All haplotypes showed increased risk of BCC with increasing arsenic exposure. The OR:s were above 1.0 for all analyses, but some analyses were not statistically significant. ORs ranged from 1.1 to 1.4, P values ranged from <0.001 to 0.082, except for among individuals with one or two copies of haplotype 2 (OR 1.0, CI 0.9–1.2, P value 0.85).

Individuals with zero copies of haplotype 1 had a higher BCC risk with increasing lifetime concentration of arsenic, compared to those with those with one or two copies [zero copies: OR 1.4, CI 1.1–1.8, P value 0.001; one copy: OR 1.2, CI 0.9–1.2, P value 0.36 (N = 492); and two copies OR 1.2, CI 1.0–1.4 (N = 290), P value 0.054]. There was a significant effect modification (P = 0.036). Similarly, individuals with zero copies of haplotype 2 had a significantly increased BCC risk (OR 1.2, CI 1.1–1.4, P value <0.001; P value for interaction 0.098). Haplotypes 3, 4, and 5 all showed higher risk with one or two copies than with zero copies.

**AS3MT Haplotype Combinations and Arsenic-Related BCC Risk**

We found a significant effect modification on BCC risk for the combination of haplotypes 1 and 2 (P = 0.010) and haplotypes 1 and 4 (P = 0.020; Table IV). The highest risk among all the haplotype combinations was seen for individuals with zero copies of both haplotype 1 and 2, who had 2.7 times higher risk in relation to arsenic exposure (CI 1.5–4.8). Individuals with one or two copies of haplotype 2 did not show any increased risk of BCC with increasing arsenic exposure, irrespective of the number of copies of haplotype 1. Also, for the combination of haplotype 1 and 4, individuals with zero copies of both haplotypes had the highest BCC risk (OR 1.6, CI 1.1–2.2). Individuals with zero copies of haplotype 1 in combination with one or two copies of haplotype 4, which was associated with very high urinary fraction of MMA (Table III), showed almost as high risk (OR 1.5). Individuals with one or two copies of both haplotypes had a lower risk of BCC (OR of 1.1, CI 0.7–1.6, P value 0.79), mainly due to the protective effects of haplotype 1. When combining haplotypes 2 and 4, the interaction term was not statistically significant and the analysis was hampered by having too few individuals with one or two copies of both haplotypes.

**DISCUSSION**

This study, carried out in a European population with low to moderate arsenic exposure, confirms earlier findings of an association between genetic variation in AS3MT and arsenic metabolism in Asian or American populations exposed to high levels of arsenic. The study also showed that this gene, which encodes a key arsenic-metabolizing enzyme, modifies the previously reported association between arsenic exposure and BCC [Leonardi et al., 2012]. Importantly, the observed gene–environment interaction supports the hypothesis that the association
between low-level arsenic exposure and BCC risk reflects a causal relationship.

The directions of the results suggest that polymorphisms in \textit{AS3MT} that associated with less-efficient arsenic methylation (especially lack of copies of haplotypes 1 and 2, particularly in combination, and one or two copies of haplotype 4) are associated with a higher risk of arsenic-related BCC. However, only haplotype 1 demonstrated a statistically significant effect modification, where individuals with one or two copies of haplotype 1 had lower BCC risk with increasing exposure compared to individuals with no copies of haplotype 1, but this haplotype did not seem to affect arsenic metabolism. Thus, \textit{AS3MT} may influence the susceptibility to BCC via other mechanisms not related to arsenic metabolism. Other functions of \textit{AS3MT} in the human body apart from arsenic metabolism have not yet been elucidated. An interesting finding was that individuals lacking both haplotype 1 and 2 had an almost three times increase in BCC risk per 10 \( \mu \text{g/L} \) increase in lifetime drinking water arsenic. Similarly, those lacking both haplotypes 1 and 4 had a 60\% increase in BCC risk for the same degree of arsenic exposure. It should be noted that the speciation of arsenic metabolites in urine was only performed in the controls and we cannot be sure if the effects found are the same in the individuals that developed cancer. This fact stresses that prospective studies are needed to follow individual arsenic metabolism and relate that to the development of cancer.

The \textit{AS3MT} haplotypes found in this population were the same as those previously found in Argentina and Bangladesh [Engstrom et al., 2011]. We also found three haplotypes that were not found in the studies in Argentina or Bangladesh: one haplotype with a frequency above 5\%
(haplotype 5), and two minor haplotypes. The haplotype frequencies were similar in all European countries studied here and included all four major haplotypes that have been shown to affect arsenic metabolism [Engstrom et al., 2011]. However, in the current study, the associations between AS3MT haplotypes and metabolism differed somewhat from the studied Asian and South American populations in that haplotypes 1 and 3 did not modify arsenic metabolism. This suggests that the SNPs analyzed are not functional themselves, but are in LD with functional variants that have a different LD pattern in these European populations compared with the Argentinean and Bangladeshi populations.

This study has several strengths. The study population is fairly large and several of the variant alleles had relatively high frequency. The LD analysis indicated a similar genetic structure for the AS3MT gene in the three different subpopulations and those could thus be pooled to increase the power of the analyses. Differences in unmeasured confounders between countries were partly addressed by including county, accounting for regional differences in arsenic exposure and BCC risk, as a variable in the models. All arsenic measurements and genotyping analyses were made by the same method and in the same laboratory, using adequate quality control. However, the low levels of arsenic exposure and the uncertainties in lifetime arsenic exposure assessments have likely hampered the evaluation of associations between AS3MT haplotypes and arsenic metabolism. The urinary arsenic concentrations were substantially lower in this study (median 6 μg/L for all individuals) compared with previous studies in Argentina and Bangladesh (medians 200 and 100 μg/L, respectively [Engstrom et al., 2011]). At such low levels of exposure, food may also be an important source of arsenic exposure, and food may contain both iAs, DMA, and arsenosugars, which are metabolized to DMA in the body [Lindberg et al., 2006]. Also, other urinary metabolites than DMA were often below the limit of detection, distorting the true metabolite distribution. Although we excluded individuals with arsenic in urine below 5 μg/L in these analyses, the arsenic metabolites in urine may, for some individuals, not represent exposure to exclusively inorganic arsenic. However, a limitation of this study is that for some of the analyses (e.g., the combination of haplotypes 2 and 4) there are a small number of individuals with certain haplotype/haplotype combinations. We attempted to find out if gender-specific effects of AS3MT could explain gender differences in arsenic metabolism [Lindberg et al., 2008a]. However, this appears not to be the case, as we observed similar directions of the associations for both men and women, and there were in general no big differences in genetic effect modification between the sexes (data not shown).

The association between genetic variation in AS3MT and cancer is less studied than the association between genetic variation in AS3MT and arsenic metabolism. A recent genome-wide association study [Pierce et al., 2012] confirmed our previous results of strong associations between AS3MT genotype and arsenic metabolism [Schlawicke Engstrom et al., 2007; Engstrom et al., 2011] and two SNPs in the chromosome 10 region near AS3MT (rs9527 and rs11191659) were also associated with increased risk for arsenic-related skin lesions [Pierce et al., 2012]. We have previously found that AS3MT haplotypes associated with levels of premalignant DNA damage in women exposed to arsenic [Hossain et al., 2012, Li et al., 2012]. To our knowledge, no studies have evaluated rs3740400 and arsenic-related cancer risk. Rs11191439, rs1046778 and rs3740393 have recently been evaluated in relation to arsenic-related bladder cancer. Individuals with one or more copies of the C allele (coding for Thr) in rs11191439 (here in haplotype 4) had an elevated risk of bladder cancer [Beebe-Dimmer et al., 2012]. No significant effects were found for rs1046778 and rs3740393 [Lesueur et al., 2012], but many of the analyses lacked statistical power. One study found that the Thr-allele of the rs11191439 polymorphism associated with a marginally increased risk of arsenic-related skin lesions [Valenzuela et al., 2009], whereas no effect was found in another study [De Chaudhuri et al., 2008] or in the present study. There was no association between rs3740393 and cancer risk (all types of cancer combined) in a population from Taiwan [Chung et al., 2009]. Also, two polymorphisms not evaluated in this study, rs11191453 and rs7085104, did not affect the risk of skin lesions in a Mexican population [Valenzuela et al., 2009]. Polymorphisms in genes other than AS3MT may also influence BCC risk, such as SNPs in DNA repair genes NBS1 and XRCC3, which showed effect modulation for BCC [Thirumaran et al., 2006]. Other candidates include SNPs in fibrous sheath interacting protein 1 (FSIP1) and solute carrier family 39, member 2 (SLC39A2) that influence the risk of arsenic-related bladder cancer, as shown in a study scanning 10,000 SNPs [Karagas et al., 2012] or SNPs in p21 or glutathione S-transferase omega 1 (GSTO1) and GSTO2 that influence the risk of arsenic-related urothelial carcinoma [Chung et al., 2008, 2011].

In conclusion, AS3MT haplotypes associated with the arsenic metabolite profile, and possibly also other AS3MT haplotypes, modified the risk of BCC among individuals with low exposure to arsenic. These results may explain part of the observed variation in susceptibility to arsenic-related cancer.

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**AUTHOR CONTRIBUTIONS**

Dr. Broberg and Prof. Vahter designed the study. Dr. Engström analyzed the data and drafted the manuscript. Drs. Fletcher, Leonard, Goessler, Gurzau, Koppova, Rudnai, and Kumar have been involved in sample collection, clinical data acquisition and exposure assessment. All authors have revised the manuscript and approved the final manuscript.

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