SCIENTIFIC REPORTS

Received: 7 March 2018 Accepted: 3 July 2018 Published online: 13 July 2018

OPEN ACGH detects distinct genomic alterations of primary intrahepatic cholangiocarcinomas and matched lymph node metastases and identifies a poor prognosis subclass

Ruben Jansen¹, Birte Moehlendick^{1,2}, Christoph Bartenhagen³, Csaba Tóth⁴, Nadja Lehwald¹, Nikolas H. Stoecklein¹, Wolfram T. Knoefel¹ & Anja Lachenmayer^{1,5}

Lymph node metastases (LNM) are an important prognostic factor for patients with intrahepatic cholangiocarcinoma, but underlying genetic alterations are poorly understood. Whole genome array comparative genomic hybridization (aCGH) was performed in 37 tumors and 14 matched LNM. Genomic analyses of tumors confirmed known and identified new (gains in 19g) copy number alterations (CNA). Tumors with LNM (N1) had more alterations and exclusive gains (3p, 4q, 5p, 13q) and losses (17p and 20p). LNM shared most alterations with their matched tumors (86%), but 79% acquired new isolated gains [12q14 (36%); 1p13, 2p23, 7p22, 7q11, 11q12, 13q13 and 14q12 (>20%)]. Unsupervised clustering revealed a poor prognosis subclass with increased alterations significantly associated to tumor differentiation and survival. TP53 and KRAS mutations occurred in 19% of tumors and 6% of metastases. Pathway analyses revealed association to cancer-associated pathways. Advanced tumor stage, microvascular/perineural invasion, and microscopic positive resection margin (R1) were significantly correlated to metastases, while N1-status, R1-resection, and poor tumor differentiation were significantly correlated to survival. ACGH identified clear differences between N0 (no LNM) and N1 tumors, while N1 tumors and matched LNM displayed high clonality with exclusive gains in the metastases. A novel subclass with increased CNAs and poor tumor differentiation was significantly correlated to survival.

Cholangiocarcinoma is the second most common liver malignancy worldwide and accounts for nearly 3% of all gastrointestinal cancers^{1,2}. Although intrahepatic cholangiocarcinoma (iCCA) accounts only for 10% of all primary liver malignancies, its incidence and mortality rate is increasing worldwide^{3,4}. The 5-year survival rate for patients with iCCA varies between <5% and 20-40% depending on the tumor stage at diagnosis. While surgical resection remains to be the only curative therapy so far⁵, the occurrence of LNM is an independent prognostic factor for iCCA with a significant correlation to poor survival⁶. A recent consensus statement recommended to consider lymph node dissection although strong clinical data supporting this approach is still missing⁷. While some centers perform a routine lymphadenectomy in order to improve staging and to increase the amount of prospective information for the patient, others even recommend adjuvant chemotherapy for patients with nodal positive tumors. Nevertheless all clinical trials and studies conducted so far were not able to show a clear benefit for the adjuvant treatment of this disease and data of ongoing clinical trials is still urgently awaited. Although several studies recently performed comprehensive genetic analyses of primary iCCA, a clear understanding of how LNM develop and how advanced tumor stages could be treated is still not available. ACGH has been successfully

¹Department of Surgery, Heinrich-Heine University and University Hospital Duesseldorf, Duesseldorf, Germany. ²Institute of Pharmacogenetics, University Hospital Essen, Essen, Germany. ³Department of Experimental Pediatric Oncology, University of Cologne, Cologne, Germany. ⁴Institute of Pathology, University Hospital Heidelberg, Heidelberg, Germany. ⁵Department of Visceral Surgery and Medicine, Inselspital, Bern University Hospital, University of Bern, Bern, Switzerland. Correspondence and requests for materials should be addressed to A.L. (email: anja. lachenmayer@insel.ch)

Features	Total (n = 60)	N0 $(n=31)$	N1 (n = 23)	<i>p</i> -Value
Tumor stage				
T1+2	32 (53%)	22 (71%)	9 (39%)	
T3+4	22 (37%)	7 (23%)	14 (61%)	0.007
Median tumor size				
Pathology (in cm)	6.6 (1.5–16)	6.3 (0.7-14)	7.1 (2.3–13)	n.s.
Node stage	54	31 (52%)	23 (38%)	
Median number of resected lymph nodes	9 (0-39)	7 (1-13)	15 (1-39)	< 0.001
Distant metastasis				
M0	48 (80%)	29 (94%)	17 (74%)	
M1	7 (12%)	1 (3%)	4 (17%)	0.063
Differentiation				
Well	1 (2%)	1 (3%)	0	n.s.
Moderate	29 (48%)	17 (55%)	11 (48%)	n.s.
Poor	25 (42%)	12 (39%)	12 (52%)	n.s.
Tumor surgery				
R0 (tumor-free margin)	48 (80%)	28 (90%)	15 (65%)	
R1	12 (20%)	3 (10%)	8 (35%)	0.024
Vascular invasion				
V0	41 (68%)	29 (94%)	11 (48%)	
V1 (microvascular)	13 (22%)	1 (3%)	11 (48%)	< 0.001
Perineural invasion				
Pn0	33 (55%)	25 (81%)	6 (26%)	
Pn1	13 (22%)	4 (13%)	9 (39%)	0.001

Table 1. Histopathological data.

used in colorectal cancer to identify molecular alterations in primary tumors and their matched metastases^{8,9}. As an accepted standard for the detection of copy number alterations (CNA), aCGH yields a high throughput by low costs, providing benefits when working with a large number of samples compared to expensive next generation sequencing technologies¹⁰⁻¹². Analyses of grouped pairs of primary tumors and matched metastases have recently been used in pancreatic cancer and melanoma to assess the polyclonality and genetic heterogeneity, but are still not reported for intrahepatic cholangiocarcinomas¹³⁻¹⁵. In addition, a CNA based algorithm using CNAs of the cancer founding clone was recently developed to identify cancer types in circulating tumor cells and cell-free DNAs¹⁶.

Therefore the aim of our study was (1) to analyze copy number alterations in intrahepatic cholangiocarcinomas and their corresponding lymph node metastases and (2) to analyze and correlate clinical and histopathological factors.

Results

Clinical and histopathological data correlated with LNM and outcome in iCCA. Twenty-three (38%) patients of our cohort had nodal positive iCCA (Table 1). Patients with LNM had significantly more lymph nodes resected than patients without (15 vs. 7, p < 0.001, Table 1). Advanced tumor stages (T3/4) (p = 0.007), smoking (p = 0.008), R1 resection (p = 0.024), microvascular (p < 0.001) and perineural (p < 0.001) invasion were significantly correlated with the occurrence of LNM (Table 1 and Supplementary Table 1). Cases without clinical data were eliminated from the analysis. Resection modus had no influence and no differences were seen between different surgical approaches (Supplementary Table 2).

Kaplan-Meier analysis showed significantly worse survival of patients with LNM, (p = 0.049). The median survival rate for patients with N1 tumors was 18 months compared to 26 months for patients with N0 tumors. While R1 (p < 0.001) and poor tumor differentiation (p < 0.001) were significantly correlated with poor outcome (Fig. 1), the influence of tumor size was not statistically significant (p = 0.061, Supplementary Fig. 1). Nevertheless, none of the above factors reached statistical significance in the multivariate analysis. Gender, vascular invasion, perineural invasion and distant metastases did not affect patient survival rates. Patients with LNM developed significantly more (11 vs. 5, p = 0.012, Supplementary Table 3) and earlier distant metastases than those with N0 tumors (8 vs. 17 months). Median survival and medium follow-up were 21 months. In our cohort the 1/3/5-year survival rate was 60%, 16% and 12%, respectively.

Genetic Alterations in iCCAs and their matched LNM. All tumor samples (primary tumors and LNM) were analyzed by aCGH and several chromosomal gains and losses were detected. In our cohort the overall amount of detected losses was higher compared to the amount of gains. The penetrance plot in Fig. 2 illustrates all detected CNAs. N1 primary tumors harbored more CNAs than tumors without, in N1 primary tumors, a mean number of 45.13 CNAs was detected compared to 37.43 CNAs in N0 tumors (Supplementary Fig. 2). In addition, the maximum amount of CNAs was higher in N1 (3–161 CNAs) compared to N0 tumors (4–137 CNAs) (Supplementary Fig. 2).







Figure 2. aCGH results. In this penetrance plot all iCCA samples (N1, N0 primary tumors and LNM) are considered together (n = 51). Chromosomal gains are displayed in red and chromosomal losses in green. The chart sums up the gains and losses that are present in at least 30% of the samples. Gains on chromosome 19q have not been described for iCCA so far. The horizontal line indicates the threshold of 30%, the y axis represent the % of cases with aberration.



Figure 3. Copy number alterations of N0 and N1 primary tumors. This penetrance plot shows the differences between N0 and N1 primary tumors. Chromosomal gains are displayed in red and losses in green. Additional gains in N1 tumors are indicated (arrows). The horizontal line indicates the threshold of 30%, the y axis represent the % of cases with aberration.



Figure 4. Congruent altered cytobands of matched primary tumors and LNM. This figure shows the number of primary tumor and LNM pairs (in percentage) found with concurrent altered cytobands (resolution = 320 bands per haploid set).

Fig. 2). LNM also harbored more CNAs (44.22), compared to N0 tumors (Supplementary Table 4), however, this result was not statistically significant (One-Way ANOVA by Kruskal-Wallis test with Dunn's multiple comparisons test, p = 0.9896).

Nodal positive primary tumors showed significant additional gains and losses compared to N0 tumors. Gains were found on chromosome arms 3p (p = 0.006), 4q (p = 0.016), 5p (p = 0.002) and 13q (p = 0.039, Fig. 3), losses could be detected on chromosome 17p (p < 0.001) and 20p (p = 0.006) (Fig. 3). Gains and losses that were detected in N0 tumors were also found in N1 tumors, and harbored relevant cancer-related genes (Supplementary Table 5).

When primary tumors and their matched LNM were analyzed, few but specific differences were detected. Only 14% of the pairs had incongruent CNAs, while 86% of the primary tumors and their matched LNM shared the same alterations. Seventy-nine % of these latter pairs shared up to 20 CNAs, while 7% had congruent CNAs in 21–40 cytobands (resolution = 320 bands per haploid set) (Fig. 4). Despite the great similarity between tumors and their corresponding LNM, 79% of the LNM acquired additional specific isolated gains compared to their matched primary tumors: 36% showed isolated gains on chromosome 12q14 and more than 20% displayed isolated gains on chromosome bands 1p13, 2p23, 7p22, 7q11, 11q12, 13q13 and 14q12 (Fig. 5). Supplementary



Figure 5. Isolated gains at 12q14 and 14q12. Example of the moving average (colored lines) and alterations (colored areas) from a matched primary tumor and LNM pair. The aCGH profile of the LNM is displayed in blue and the profile of the primary tumor in red. The overlap of these profiles clearly shows additional isolated gains in LNM on 12q14 and 14q12 (arrows) compared to its primary tumor.



Figure 6. Cluster analysis of primary tumors. This cluster analysis shows two groups of primary tumors. Group B harbored significantly more poorly differentiated tumors and was associated with poor survival rates. Chromosomal losses are displayed in green and gains in red. Blue and red columns on the left side indicate the different primary tumors (N1 = blue; N0 = red).

Fig. 3 displays the specific CGH profile of each patient to support these findings. These gains harbor relevant cancer-related genes (Supplementary Table 6). New isolated losses in LNMs occurred infrequently in less than 20% of the samples.

Unsupervised clustering of CNAs of all primary tumors identified two distinct groups (group A and group B) (Fig. 6). Intriguingly, group B tumors were more poorly differentiated compared to group A (G3/4 vs. G1/2, p = 0.034) and had a significant shorter survival (13 vs. 31 months, p = 0.035) than group A (Supplementary Fig. 4). Subgroup B tumors also harbored significantly more CNAs (60.5 vs. 21.5, p = 0.0246, Mann Whitney U test) than subgroup A. In addition, the maximum amount of CNAs was higher in subgroup B (161 vs. 71) (Supplementary Fig. 5). Subgroup B also showed significant additional gains and losses compared to subgroup A (Supplementary Table 7) that contained relevant cancer-related genes (Supplementary Table 8). Other analyzed clinical factors (gender, smoking, tumor size, regional lymph node metastasis, distant metastasis, liver cirrhosis, perineural and vascular invasion, tumor surgery and tumor markers) showed no significant correlation to one of the subgroups. When the unsupervised clustering was performed for N1 tumors and their corresponding LNMs, LNM and primary tumors of most of the pairs clustered next or in great proximity to each other due to the high genetic similarity between the pairs (Fig. 7).

We performed mutational analyses for the common and cancer-relevant mutations in the genes *KRAS* and *TP53*. Mutational analysis (18 N1 primary tumors and 18 matched lymph node metastases) revealed mutations of *TP53* in six samples (19%) and mutations of *KRAS* in two samples (6%). Of the six *TP53* mutations, four were detected for Exon 7 (11%) and two for Exon 5 (six %) (4 mutations were found in primary tumors, 2 in LNM). The *KRAS* mutations were detected for Exon 2 (both were found in primary tumors). Mutations in the primary tumor were also detected in the corresponding lymph node metastasis, except for one sample. Since the mutation analyses were only performed in N1 tumors, no information was available for the penetrance in N0 iCCAs. While no CNAs could be detected for the regions harboring the *TP53* mutation, we found one gain on 12p12 (location



Figure 7. Cluster analysis of primary tumors and LNM. Due to the high genetic similarity LNM and their matched primary tumors clustered next or in great proximity to each other (boxes on the right side). Chromosomal losses are displayed in blue and gains in yellow.

of the *KRAS* mutation). Interestingly, the amount of CNAs did not differ between tumors with TP53 inactivating mutations and those without.

Pathway analysis and Therapeutic Target Database analysis. KEGG pathway analysis of the genes located within the additional gains and losses that distinguish N1 from N0 primary tumors, and analysis of the genes located within the isolated gains of the lymph node metastases showed a significant activation of cancer-associated pathways. The genes altered in N1 primary tumors are involved in Wnt-, MAPK-, JAK-STAT-, PPAR-, mTOR-, TGF- β , p53- and VEGF-signaling pathways. An *Ecm-receptor* interaction could also be shown (Supplementary Table 7 and Supplementary Fig. 6). Genes located on the isolated LNM gains are mainly involved in *chemokine-signaling* pathways which also influence cellular growth, cell survival, cell migration and apoptosis (Supplementary Table 9 and Supplementary Fig. 7). For genes located within the additional gains and losses of subclass B, we detected a correlation to the JAK-STAT-, SCLC-, Thyroid cancer-, p53-, DNA replication-, MAPK-, Wnt-, TGF-beta-, Hedgehog- and Chemokin signaling pathways (Supplementary Table 10).

The additional gains and losses in the N1 primary tumors hold 1721 known genes, including 43 genes that are known to play important roles in cancer-associated pathways. Twenty genes located within the N1 CNAs matched as targets (Therapeutic Targets Database) for 41 potential substances, of which 26 are in phase two or a later stage of a clinical trial (Supplementary Table 11). Thirty-five genes located within the LNM CNAs were matched to 206 potential substances, of which 132 are in phase two or a later stage of clinical trial (Supplementary Table 12). The poor prognosis subgroup B also showed additional gains and losses, holding 180 genes known to play important roles in cancer-associated pathways. Fourteen genes had a match as a therapeutic target, revealing 53 potential substances targeting these genes (Supplementary Table 13). For genes located on deleted regions we looked for activating drugs, while drugs for genes on gained regions should work as inhibitors, antagonists, or antibodies. Interestingly, none of the identified drugs have been tested in metastatic iCCA patients so far.

Discussion

Intrahepatic cholangiocarcinoma is the second most common primary liver tumor with an increasing incidence worldwide and a very poor outcome due to its aggressive tumor biology and limited curative treatment options¹⁷. LNM have been identified as an important prognostic factor therefore, we aimed to analyze copy number alterations in primary tumors along with their matched LNM and together with important clinical factors.

ACGH of all primary tumors confirmed known chromosomal aberrations for iCCA¹⁸, but also detected frequent novel gains in 19q not previously described. 19q holds the growth factor TGFB1 and several genes involved in cancer associated pathways¹⁹. A recently published analysis suggested that enhanced circulating TGF- β 1 is associated with an increased risk of hepatocellular carcinoma, but data for iCCA has not been reported yet²⁰. However, careful interpretation is warranted and results need to be validated with other methods in the future, as chromosome 19 is often subject to labeling artifacts in aCGH studies^{21,22}. Interestingly, the overall number of CNAs was much higher in N1 tumors compared to N0 tumors with distinctly different CNAs (additional gains on chromosome arms 3p, 4q, 5p and 13q and further losses in 17p and 20p in N1). This might reflect the progression from N0 to N1 with genomic alterations also responsible for metastasis. Accordingly, the in silico pathway analyses revealed that chromosomal gains of N1 tumors harbor genes involved in pathways important for tumor progression and metastasis, such as the MAPK-, Wnt-, JAK-STAT-, PPAR-, TGF-β-, p53-, VEGF- and mTOR-signaling pathways^{18,19,23,24}. These gains include genes on 5p known to be involved in the JAK-STAT signaling pathway and the development of metastases²⁵, Growth Hormone Receptor (GHR), which was overexpressed in metastatic breast cancer²⁶ and Leukemia Inhibitory Factor Receptor Alpha (LIFR), which was involved in chemoresistance of cholangiocarcinoma cells²⁷. In addition, gains on 13q have been reported to play an important role for esophagus and bladder cancer, gliomas and glioblastomas^{23,28-31}, but not yet for iCCA. This chromosomal area also contains Transcription Factor Dp-1 (TFDP1), which plays a crucial role in cell cycle progression³² and the oncogene Fms Related Tyrosine Kinase 1 (*FLT1*), known to induce metastasis in breast cancer models³³. Overexpression of the latter was also significantly associated with poor survival in cholangiocarcinoma³⁴.

Significant losses were detected in chromosomal regions harboring important tumor suppressor genes. Chromosome region 17p contains *TP53*, which is lost or mutated in half of all human cancers³⁵. It could recently be shown, that the metastatic potential of cholangiocarcinoma cells can be increased by the deletion of *TP53*, underscoring its role for the development of metastasis^{35,36}. We also detected mutations for *TP53* in four N1 tumors and two matched LNM. In addition, the in our samples altered regions 20p12 and 20p13 hold the known tumor suppressor genes Phospholipase C Beta 1 and 4 (*PLCB1*, *PLCB4*) and Casein Kinase 2 Alpha 1 (*CSNK2A1*), all previously described to be associated with advanced tumor stages and bad outcome in different cancers but not yet in iCCA³⁷⁻³⁹. Taken together aCGH is able to detect clear differences between N1 and N0 iCCAs with several significant gains and losses in N1 tumors harboring important cancer-related genes that might play an important role for the development of LNM.

Pairwise comparison of primary tumors and corresponding LNM revealed a great similarity regarding chromosomal gains and losses and the amount of CNAs, but also detected additional, unique isolated gains on the chromosome band 12q14 (36%) and on 1p13, 2p23, 7p22, 7q11, 11q12, 13q13 and 14q12 (>20%). Matched genomic analyses of primary tumors and their corresponding metastases are rarely reported in the literature, although they have been proposed to be an ideal tool to identify "driver" molecular changes responsible for the development of metastases⁴⁰. The specific isolated gain on chromosome band 12q14 that occurred in 36% of our LNM holds several interesting genes like NF-κB activating kinase TANK Binding Kinase 1 (*TBK1*), transcription factor High Mobility Group AT-Hook 2 (*HMGA2*), MDM2 Proto-Oncogene (*MDM2*), Glucosamine (N-Acetyl)-6-Sulfatase (*GNS*) and Cullin Associated And Neddylation Dissociated 1 (*CAND1*). The latter one is known to be important for chromosomal stability and has recently been reported to play a key role in prostate cancer^{41,42}. *TBK1* was successfully targeted in oral squamous cell carcinoma and osteosarcoma leading to a decrease in tumor activity^{43,44} and *MDM2* amplifications are known to be involved in tumorigenesis of liposarcoma⁴⁵⁻⁴⁷. Furthermore, several studies have shown the importance of HMGA2 expression for metastasis and tumor progression in different malignancies but not yet in iCCA⁴⁸⁻⁵³. To the best of our knowledge this is the first matched primary tumor-LNM analysis in iCCA describing a high clonality between primary tumors and LNM.

Unsupervised clustering of all CNAs revealed one group of patients with a significant correlation to poor survival and poor tumor differentiation independent of the lymph node status. This poor prognosis group harbored significantly more CNAs including chromosomal alterations at 7p, 9p and 17q that have previously been described to be related to poor tumor differentiation and worse survival in iCCA. Genes located in these regions include C-X-C Motif Chemokine Receptor 4 (*CXCR4*), which inhibited tumor cell proliferation and neural invasion in cell lines of hilar cholangiocarcinoma when silenced⁵⁴, Toll Like Receptor 2 (*TLR2*), known to promote cell migration and invasion by modulating NF-κB pathway in iCCA⁵⁵, and Interleukin-8 (*IL8*), associated to advanced TNM stages and recurrence of hilar cholangiocarcinoma⁵⁶. Other altered regions characterizing this poor prognosis subclass include C-X-C Motif Chemokine Ligand 5 (*CXCL5*), known to promote growth and metastasis in iCCA⁵⁷, transcription factor FOS Like 1 (*FOSL1*) and Erb-B2 Receptor Tyrosine Kinase 2 (ErbB2), both important for tumorigenesis of cholangiocarcinoma^{58,59}.

Molecular classes are increasingly reported for several different cancers including iCCA. They are often based on broad genetic analyses including whole genome RNA expression, high-density single-nucleotide polymorphism arrays and mutational analyses with correlations to important clinical and prognostic factors⁶⁰. While their prognostic relevance and the biological significance needs to be thoroughly validated before a clinical application in iCCA might be feasible, the general advantages of using cancer hallmarks as cancer biomarkers in form of gene signatures and algorithms based on large patient cohorts and a training/validation set approach has now repeatedly been shown for several different cancers in the literature^{61–63}. Finally, Therapeutic Targets Database (TTD) analyses revealed relevant drugs targeting genes located in the CNAs of our N1 primary tumors, LNMs and the samples of the poor prognosis subclass, that could be tested in preclinical models of iCCA. In addition, altered regions that did not match with known anti-tumoral compounds, should be explored as potential targets for new anti-tumoral therapies in the future. In respect to the suspected high chromosomal instability of the samples of our poor prognosis class, PARP inhibitors might be considered to overcome poor responses to classical chemotherapy⁶⁴.

While our main study results are centered around the genomic differences and similarities of iCCA and their corresponding LNMs, we were also able to confirm several recently reported clinical and histopathological factors associated to LNM and outcome of intrahepatic cholangiocarcinoma. In addition we identified R1-resection and smoking to be significantly associated to LNM in iCCA in the univariate analysis.

ACGH of iCCA primary tumors and LNM revealed known and novel CNAs which might be responsible for tumor progression and the development of LNM. For the first time, a matched analysis of primary iCCA tumors and their corresponding LNM was performed showing highly similar CNAs with very specific novel alterations in the LNM. Multiple genes located in the altered regions of our cohort are known to be involved in cancer-associated pathways and have been reported as potential targets for new anti-tumoral treatment strategies. Lastly and most importantly, we identified a novel molecular subclass with a high amount of CNAs and a significant association to poor tumor differentiation and poor prognosis.

Materials and Methods

Statistical analysis. Statistical analysis was performed using SPSS software (version 22.0). Demographic data and baseline characteristics of the patients were analyzed by calculating frequencies and percentages for discrete variables and means, maximum and minimum for continuous variables. The Kolmogorov-Smirnov-Test was used to assess normality of tumor markers, nodal stage and tumor size. T-test was then used to test differences between nodal stage, tumor marker levels and tumor size (for normal distribution). Differences in

etiological data, tumor characteristics and chromosomal alterations were tested with the Chi-squared test. Overall survival was assessed with the Kaplan-Meier estimator and Cox regression analysis for the multivariate analysis. A *p*-value of 0.05 or less was considered significant.

Human Tissue Samples. A total of 78 formalin-fixed, paraffin-embedded (FFPE) samples were collected from 60 patients diagnosed with intrahepatic cholangiocarcinoma and resected between 2003 and 2013 at the Department of General-, Visceral- and Pediatric Surgery of the University Hospital Duesseldorf, Germany. During surgery, most patients received a complete lymph node dissection. Pathological diagnosis of iCCA was always confirmed by an expert pathologist and immunostaining was performed in most of the cases (44/60). Clinical factors and tumor characteristics of the patients who underwent liver surgery are described in Supplementary Tables 1 and 2. Follow-up data was available for 59 patients (one patient was lost to follow-up). The study was approved by the institutional ethics committee of the University Duesseldorf, Germany (#3821). We hereby confirm that all research was performed in accordance with relevant guidelines and regulations and informed consent was obtained from all participants and/or their legal guardians.

Immunohistochemistry. Immunohistochemistry was performed on one tumor tissue block, representative for the whole lesion. At first 4µm thick paraffin sections were cut and dried overnight. For immunohistochemical analysis the following antibodies were used: CK7 (Dako, Denmark, monoclonal antibody, clone OV-TL 12/30, dilution 1/6000) and a CK20 (Dako, Denmark, monoclonal antibody, clone Ks20.8, dilution 1/400), CK8 (Becton, Dickinson and Company, monoclonal antibody, clone CAM5.2, dilution 1/400), CK19 (Dako, Denmark, monoclonal antibody, dilution 1/200) and CEA (Dako, Denmark, polyclonal, dilution 1/40.000). Tumor tissue was considered as iCCA, if CK 7 and CK 19 showed positive staining and cytokeratin 20 showed a negative staining⁶⁵. The stainings were carried out using an Avidin–Biotin method with antigen retrieval. The staining was performed on an automated Ventana IHC instrument with a Ventana basic DAB detection kit. Our immunohistochemical analysis was validated through positive and negative controls (by omitting the primary antibody).

DNA extraction. Invasive tumorous tissue was distinguished from non-neoplastic tissue with two-µm-thick H&E sections. Tumor tissue and non-neoplastic tissue then were macrodissected and DNA extraction and sample preparation was performed from ten nine-µm-thick FFPE sections according to the manufacturers' protocol for FFPE samples (QIAamp DNA FFPE Tissue Kit QIAGEN, Hilden, Germany). DNA concentration was estimated with the Qubit[®] 2.0 Fluorometer (Life Technologies, Carlsbad, California, USA) and considered as sufficient, if the DNA concentration was 100 ng/µl or higher. DNA quality was evaluated by measuring fragmentation of the DNA using Agarose gel electrophoresis.

Array comparative genomic hybridization. ACGH analyses on oligonucleotide arrays were carried out according to the manufacturers' protocol (Agilent Oligonucleotide Array-Based CGH for Genomic DNA Analysis, Version 6.4, August 2011, G4410-90010). Basically, for each sample non-neoplastic tissue DNA from the same patient was used as a reference. Tumor cell and reference gDNA were both random-primed labeled with Cyanine 5-/Cyanine 3-dUTP and hybridized on the 8×60 k-DNA-microarray chips. The array slides were washed and subsequently scanned using the Microarray Scanner G2565CA by Agilent Technologies with 3 µm resolution and 16 bit color depth.

Analysis of aCGH Data. The generated data from the scanned images were extracted with the Feature Extraction software (Agilent Technologies, Version 10.7.3.1, Protocol CGH_107_Sep09) and output files were imported into the Genomic Workbench 5.0.14 software. We used the aberration detection method-2 (ADM-2) algorithm with a threshold of 6.0 and the diploid peak centralization for evaluating copy number alterations (CNA). The ADM-2 algorithm, has been shown to provide reliable results in CNA detection of cancer cells⁶⁶. This algorithm searches for intervals in which a statistical score based on the average quality weighted log ratio of the sample and reference channels exceeds a specific threshold. For our analyses we used a semi-stringent threshold of 6.0 to reduce false-positive results and enhance the detection of true-positive copy number alterations. Since cancer cells often show major aberrations and a dominant ploidy deviant to 2, the diploid peak centralization algorithm is the method of choice to normalize the aCGH fluorescence ratios, since it centers the data so that the log ratios of the probes, that are in copy number 2 regions are centered around zero, instead of making the most-common-ploidy the new zero point. An aberration filter with a minimum log2ratio of ± 0.25 and a minimum number of 3 affected oligonucleotides was defined. Further analyses of group and pairwise comparisons were performed using data matrix charts with a resolution of 320 mphs. Data matrix charts were generated with the Genomic Workbench 5.0.14 software, displaying CNA as binary code (1 = gain, -1 = loss and 0 = no alteration). Significance of differences in the chromosomal alterations was tested with the Chi-squared test as described above (in der Statistical Anaylsis section). Penetrance plots of the CNA were generated using the internet based genomic copy number data curator progenetix⁶⁷. A threshold of 30% was chosen to clearly distinguish true CNAs from background noise. Unsupervised clustering of microarray data was performed using Eisen cluster algorithm with settings to euclidean distance and pairwise maximum linkage to compare the similarity of the samples⁶⁸. These options led to empirically good performance in analyses in comparison of DNA microarray expression and copy number data.

Mutational analysis. Mutational analysis was performed for the oncogene *Kirsten Rat Sarcoma Viral Oncogene Homolog (KRAS)* (Exon 2) and the *tumor suppressor p53 (TP53)* (Exon 5 and 7). Therefore 18 N1 primary tumors and 18 matched lymph node metastases were analyzed. Polymerase chain reaction (PCR) was performed under the following conditions: $12.5 \,\mu$ l DreamTaq Green PCR Master Mix (Thermo Fisher Scientific, Schwerte, Germany), $0.5 \,\mu$ l Primer-Mix ($10 \times$, final concentration $0.2 \,\mu$ M), $11.5 \,\mu$ l H2O and $0.5 \,\mu$ l Template DNA

were mixed and the following program was executed in a thermal cycler: 95 °C for 3 min, 95 °C for 30 sec, melting temperature (Tm) for 30 sec, 72 °C for 20 sec, repeat 30 times and finally 72 °C for 5 min. The primer sequences and Tm can be found in Table 14. The PCR products were purified according to the manufacturers' protocol (MinElute PCR Purification Kit, QIAGEN, Hilden, Germany). The analysis was validated through positive and negative controls. Sanger sequencing was performed at GATC Biotech (Konstanz, Germany).

Therapeutic Targets and Pathway Analysis. Regions with isolated amplifications in lymph node metastases compared to their corresponding primary tumors were further analyzed for known genes and matched against the Therapeutic Targets Database $(TTD)^{69}$. We used gene symbols as possible druggable targets and searched for approved drugs or promising substances in clinical trials that would target these genes. In addition genes located in the regions with chromosomal gains and losses, distinguishing N1 from N0 primary tumors, were analyzed for their involvement in cancer-associated pathways and matched against the TTD. Pathway analyses were performed using KEGG pathway analysis, with hypergeometric testing and the R function p.adjust for this purpose⁷⁰. Significance level was adjusted to always identify the ten pathways with the most significant p-values and a minimum number of two genes were required for a pathway to be considered.

References

- Blechacz, B. R. & Gores, G. J. Cholangiocarcinoma. *Clinics in liver disease* 12, 131–150, ix, https://doi.org/10.1016/j.cld.2007.11.003 (2008).
- Vauthey, J. N. & Blumgart, L. H. Recent advances in the management of cholangiocarcinomas. Seminars in liver disease 14, 109–114, https://doi.org/10.1055/s-2007-1007302 (1994).
- 3. DeOliveira, M. L. *et al.* Cholangiocarcinoma: thirty-one-year experience with 564 patients at a single institution. *Annals of surgery* 245, 755–762, https://doi.org/10.1097/01.sla.0000251366.62632.d3 (2007).
- Yang, J. D. et al. Biliary tract cancers in Olmsted County, Minnesota, 1976–2008. Am J Gastroenterol 107, 1256–1262, https://doi. org/10.1038/ajg.2012.173 (2012).
- 5. Shaib, Y. & El-Serag, H. B. The epidemiology of cholangiocarcinoma. *Seminars in liver disease* 24, 115–125, https://doi. org/10.1055/s-2004-828889 (2004).
- de Jong, M. C. et al. Intrahepatic cholangiocarcinoma: an international multi-institutional analysis of prognostic factors and lymph node assessment. Journal of clinical oncology: official journal of the American Society of Clinical Oncology 29, 3140–3145, https://doi. org/10.1200/JCO.2011.35.6519 (2011).
- 7. Weber, S. M. et al. Intrahepatic cholangiocarcinoma: expert consensus statement. HPB (Oxford) 17, 669–680, https://doi.org/10.1111/hpb.12441 (2015).
- Diep, C. B. et al. The order of genetic events associated with colorectal cancer progression inferred from meta-analysis of copy number changes. Genes, chromosomes & cancer 45, 31–41, https://doi.org/10.1002/gcc.20261 (2006).
- Cardoso, J., Boer, J., Morreau, H. & Fodde, R. Expression and genomic profiling of colorectal cancer. *Biochimica et biophysica acta* 1775, 103–137, https://doi.org/10.1016/j.bbcan.2006.08.004 (2007).
- Costa, J. L., Meijer, G., Ylstra, B. & Caldas, C. Array comparative genomic hybridization copy number profiling: a new tool for translational research in solid malignancies. *Semin Radiat Oncol* 18, 98–104, https://doi.org/10.1016/j.semradonc.2007.10.005 (2008).
- Pinkel, D. & Albertson, D. G. Array comparative genomic hybridization and its applications in cancer. Nat Genet 37(Suppl), S11–17, https://doi.org/10.1038/ng1569 (2005).
- Idris, S. F., Ahmad, S. S., Scott, M. A., Vassiliou, G. S. & Hadfield, J. The role of high-throughput technologies in clinical cancer genomics. *Expert Rev Mol Diagn* 13, 167–181, https://doi.org/10.1586/erm.13.1 (2013).
- Lee, S. Y. et al. Comparative genomic analysis of primary and synchronous metastatic colorectal cancers. Plos one 9, e90459, https:// doi.org/10.1371/journal.pone.0090459 (2014).
- Campbell, P. J. et al. The patterns and dynamics of genomic instability in metastatic pancreatic cancer. Nature 467, 1109–1113, https://doi.org/10.1038/nature09460 (2010).
- Turajlic, S. et al. Whole genome sequencing of matched primary and metastatic acral melanomas. Genome Res 22, 196–207, https:// doi.org/10.1101/gr.125591.111 (2012).
- Zou, J. & Wang, E. eTumorType, An Algorithm of Discriminating Cancer Types for Circulating Tumor Cells or Cell-free DNAs in Blood. Genomics, proteomics & bioinformatics 15, 130–140, https://doi.org/10.1016/j.gpb.2017.01.004 (2017).
- Bartella, I. & Dufour, J. F. C. Diagnosis and Staging of Intrahepatic Cholangiocarcinoma. J Gastrointestin Liver Dis 24, 481–489, https://doi.org/10.15403/jgld.2014.1121.244.chl (2015).
- Sia, D., Tovar, V., Moeini, A. & Llovet, J. M. Intrahepatic cholangiocarcinoma: pathogenesis and rationale for molecular therapies. Oncogene 32, 4861–4870, https://doi.org/10.1038/onc.2012.617 (2013).
- Zhang, B., Kirov, S. & Snoddy, J. WebGestalt: an integrated system for exploring gene sets in various biological contexts. *Nucleic Acids Res* 33, W741–748, https://doi.org/10.1093/nar/gki475 (2005).
- Lu, W. Q., Qiu, J. L., Huang, Z. L. & Liu, H. Y. Enhanced circulating transforming growth factor beta 1 is causally associated with an increased risk of hepatocellular carcinoma: a mendelian randomization meta-analysis. *Oncotarget*, https://doi.org/10.18632/ oncotarget.13218 (2016).
- Lestou, V. S., Lomax, B. L., Barrett, I. J. & Kalousek, D. K. Screening of human placentas for chromosomal mosaicism using comparative genomic hybridization. *Teratology* 59, 325–330, https://doi.org/10.1002/(SICI)1096-9926(199905)59:5<325::AID-TERA3>3.0.CO;2-N (1999).
- 22. Kallioniemi, O. P. et al. Optimizing comparative genomic hybridization for analysis of DNA sequence copy number changes in solid tumors. Genes, chromosomes & cancer 10, 231–243 (1994).
- 23. Koo, S. H. *et al.* Genetic alterations in hepatocellular carcinoma and intrahepatic cholangiocarcinoma. *Cancer Genet Cytogenet* **130**, 22–28 (2001).
- Shiraishi, K. *et al.* Genetic aberrations detected by comparative genomic hybridization in biliary tract cancers. *Oncology* 57, 42–49, doi:11999 (1999).
- Teng, Y., Ross, J. L. & Cowell, J. K. The involvement of JAK-STAT3 in cell motility, invasion, and metastasis. Jakstat 3, e28086, https:// doi.org/10.4161/jkst.28086 (2014).
- 26. Stajduhar, E. *et al.* Expression of growth hormone receptor, plakoglobin and NEDD9 protein in association with tumour progression and metastasis in human breast cancer. *Tumour Biol* **35**, 6425–6434, https://doi.org/10.1007/s13277-014-1827-y (2014).
- Morton, S. D. *et al.* Leukemia inhibitory factor protects cholangiocarcinoma cells from drug-induced apoptosis via a PI3K/AKTdependent Mcl-1 activation. *Oncotarget* 6, 26052–26064, https://doi.org/10.18632/oncotarget.4482 (2015).
- Kallioniemi, A. et al. Identification of gains and losses of DNA sequences in primary bladder cancer by comparative genomic hybridization. Genes, chromosomes & cancer 12, 213–219 (1995).

- 29. Shinomiya, T. *et al.* Comparative genomic hybridization of squamous cell carcinoma of the esophagus: the possible involvement of the DPI gene in the 13q34 amplicon. *Genes, chromosomes & cancer* 24, 337–344 (1999).
- Weber, R. G. et al. Characterization of genomic alterations associated with glioma progression by comparative genomic hybridization. Oncogene 13, 983–994 (1996).
- Rijken, A. M. et al. Genomic alterations in distal bile duct carcinoma by comparative genomic hybridization and karyotype analysis. Genes, chromosomes & cancer 26, 185–191 (1999).
- 32. Jooss, K. et al. Proto-oncogenic properties of the DP family of proteins. Oncogene 10, 1529-1536 (1995).
- Qian, B. Z. et al. FLT1 signaling in metastasis-associated macrophages activates an inflammatory signature that promotes breast cancer metastasis. J Exp Med 212, 1433–1448, https://doi.org/10.1084/jem.20141555 (2015).
- Hoffmann, A. C. *et al.* EGFR, FLT1 and heparanase as markers identifying patients at risk of short survival in cholangiocarcinoma. *Plos one* 8, e64186, https://doi.org/10.1371/journal.pone.0064186 (2013).
- Powell, E., Piwnica-Worms, D. & Piwnica-Worms, H. Contribution of p53 to metastasis. *Cancer Discov* 4, 405–414, https://doi. org/10.1158/2159-8290.CD-13-0136 (2014).
- El Khatib, M. et al. Activation of Notch signaling is required for cholangiocarcinoma progression and is enhanced by inactivation of p53 in vivo. Plos one 8, e77433, https://doi.org/10.1371/journal.pone.0077433 (2013).
- Rabjerg, M. et al. Molecular characterization of clear cell renal cell carcinoma identifies CSNK2A1, SPP1 and DEFB1 as promising novel prognostic markers. APMIS 124, 372–383, https://doi.org/10.1111/apm.12519 (2016).
- Orchel, J. et al. Expression patterns of kinin-dependent genes in endometrial cancer. Int J Gynecol Cancer 22, 937–944, https://doi. org/10.1097/IGC.0b013e318259d8da (2012).
- Guerrero-Preston, R. et al. Key tumor suppressor genes inactivated by "greater promoter" methylation and somatic mutations in head and neck cancer. Epigenetics 9, 1031–1046, https://doi.org/10.4161/epi.29025 (2014).
- Tang, M. H. *et al.* Remarkable similarities of chromosomal rearrangements between primary human breast cancers and matched distant metastases as revealed by whole-genome sequencing. *Oncotarget* 6, 37169–37184, https://doi.org/10.18632/oncotarget.5951 (2015).
- Korzeniewski, N., Hohenfellner, M. & Duensing, S. CAND1 promotes PLK4-mediated centriole overduplication and is frequently disrupted in prostate cancer. *Neoplasia* 14, 799–806 (2012).
- Murata, T. et al. miR-148a is an androgen-responsive microRNA that promotes LNCaP prostate cell growth by repressing its target CAND1 expression. Prostate Cancer Prostatic Dis 13, 356–361, https://doi.org/10.1038/pcan.2010.32 (2010).
- Liu, S. & Feng, P. MiR-203 Determines Poor Outcome and Suppresses Tumor Growth by Targeting TBK1 in Osteosarcoma. Cell Physiol Biochem 37, 1956–1966, https://doi.org/10.1159/000438556 (2015).
- 44. Bai, L. Y. et al. BX795, a TBK1 inhibitor, exhibits antitumor activity in human oral squamous cell carcinoma through apoptosis induction and mitotic phase arrest. Eur J Pharmacol, https://doi.org/10.1016/j.ejphar.2015.11.032 (2015).
- Coindre, J. M., Pedeutour, F. & Aurias, A. Well-differentiated and dedifferentiated liposarcomas. Virchows Arch 456, 167–179, https://doi.org/10.1007/s00428-009-0815-x (2010).
- Hameed, M. Pathology and genetics of adipocytic tumors. Cytogenet Genome Res 118, 138–147, https://doi.org/10.1159/000108294 (2007).
- Toledo, F. & Wahl, G. M. MDM2 and MDM4: p53 regulators as targets in anticancer therapy. Int J Biochem Cell Biol 39, 1476–1482, https://doi.org/10.1016/j.biocel.2007.03.022 (2007).
- Liu, Y., Liang, H. & Jiang, X. miR-1297 promotes apoptosis and inhibits the proliferation and invasion of hepatocellular carcinoma cells by targeting HMGA2. Int J Mol Med 36, 1345–1352, https://doi.org/10.3892/ijmm.2015.2341 (2015).
- Jun, K. H., Jung, J. H., Choi, H. J., Shin, E. Y. & Chin, H. M. HMGA1/HMGA2 protein expression and prognostic implications in gastric cancer. *Int J Surg* 24, 39–44, https://doi.org/10.1016/j.ijsu.2015.10.031 (2015).
- Sun, M. et al. RKIP and HMGA2 regulate breast tumor survival and metastasis through lysyl oxidase and syndecan-2. Oncogene 33, 3528–3537, https://doi.org/10.1038/onc.2013.328 (2014).
- Zhao, X. P. et al. Overexpression of HMGA2 promotes tongue cancer metastasis through EMT pathway. J Transl Med 14, 26, https:// doi.org/10.1186/s12967-016-0777-0 (2016).
- Cai, J., Shen, G., Liu, S. & Meng, Q. Downregulation of HMGA2 inhibits cellular proliferation and invasion, improves cellular apoptosis in prostate cancer. *Tumour Biol*, https://doi.org/10.1007/s13277-015-3853-9 (2015).
- Shi, Z. et al. Silencing of HMGA2 suppresses cellular proliferation, migration, invasion, and epithelial-mesenchymal transition in bladder cancer. *Tumour Biol*, https://doi.org/10.1007/s13277-015-4625-2 (2015).
- Tan, X. Y., Chang, S., Liu, W. & Tang, H. H. Silencing of CXCR4 inhibits tumor cell proliferation and neural invasion in human hilar cholangiocarcinoma. *Gut Liver* 8, 196–204, https://doi.org/10.5009/gnl.2014.8.2.196 (2014).
- 55. Liu, B. et al. TLR2 promotes human intrahepatic cholangiocarcinoma cell migration and invasion by modulating NF-kappaB pathway-mediated inflammatory responses. The FEBS journal 283, 3839–3850, https://doi.org/10.1111/febs.13894 (2016).
- Sun, Q., Li, F., Sun, F. & Niu, J. Interleukin-8 is a prognostic indicator in human hilar cholangiocarcinoma. Int J Clin Exp Pathol 8, 8376–8384 (2015).
- Zhou, S. L. et al. CXCL5 contributes to tumor metastasis and recurrence of intrahepatic cholangiocarcinoma by recruiting infiltrative intratumoral neutrophils. Carcinogenesis 35, 597–605, https://doi.org/10.1093/carcin/bgt397 (2014).
- Yang, L., Feng, S. & Yang, Y. Identification of transcription factors (TFs) and targets involved in the cholangiocarcinoma (CCA) by integrated analysis. *Cancer Gene Ther* 23, 439–445, https://doi.org/10.1038/cgt.2016.64 (2016).
- Sirica, A. E. Role of ErbB family receptor tyrosine kinases in intrahepatic cholangiocarcinoma. World J Gastroenterol 14, 7033–7058 (2008).
- Sia, D. *et al.* Integrative molecular analysis of intrahepatic cholangiocarcinoma reveals 2 classes that have different outcomes. *Gastroenterology* 144, 829–840, https://doi.org/10.1053/j.gastro.2013.01.001 (2013).
- Wang, E. et al. Predictive genomics: a cancer hallmark network framework for predicting tumor clinical phenotypes using genome sequencing data. Seminars in cancer biology 30, 4–12, https://doi.org/10.1016/j.semcancer.2014.04.002 (2015).
- Li, J. et al. Identification of high-quality cancer prognostic markers and metastasis network modules. Nature communications 1, 34, https://doi.org/10.1038/ncomms1033 (2010).
- Gao, S. *et al.* Identification and Construction of Combinatory Cancer Hallmark-Based Gene Signature Sets to Predict Recurrence and Chemotherapy Benefit in Stage II Colorectal Cancer. *JAMA oncology* 2, 37–45, https://doi.org/10.1001/jamaoncol.2015.3413 (2016).
- 64. Bai, H. *et al.* Genetic and epigenetic heterogeneity of epithelial ovarian cancer and the clinical implications for molecular targeted therapy. *J Cell Mol Med* **20**, 581–593, https://doi.org/10.1111/jcmm.12771 (2016).
- 65. Khan, S. A. et al. Guidelines cholangiocarcinoma 2013. pdf. Gut. 61(12), 1657-1669 (2012).
- Lipson, D., Aumann, Y., Ben-Dor, A., Linial, N. & Yakhini, Z. Efficient calculation of interval scores for DNA copy number data analysis. J Comput Biol 13, 215–228, https://doi.org/10.1089/cmb.2006.13.215 (2006).
- Baudis, M. & Cleary, M. L. Progenetix.net: an online repository for molecular cytogenetic aberration data. *Bioinformatics* 17, 1228–1229 (2001).
- Eisen, M. B., Spellman, P. T., Brown, P. O. & Botstein, D. Cluster analysis and display of genome-wide expression patterns. Proc Natl Acad Sci USA 95, 14863–14868 (1998).

- 69. Yang, H. *et al.* Therapeutic target database update 2016: enriched resource for bench to clinical drug target and targeted pathway information. *Nucleic Acids Res* 44, D1069–1074, https://doi.org/10.1093/nar/gkv1230 (2016).
- Kanehisa, M., Furumichi, M., Tanabe, M., Sato, Y. & Morishima, K. KEGG: new perspectives on genomes, pathways, diseases and drugs. Nucleic Acids Res 45, D353–D361, https://doi.org/10.1093/nar/gkw1092 (2017).

Acknowledgements

This work was supported by the Duesseldorf School of Oncology (funded by the Comprehensive Cancer Center Duesseldorf/Deutsche Krebshilfe and the Medical Faculty HHU Duesseldorf). The views and opinions expressed are those of the authors and do not necessarily reflect those of the Duesseldorf School of Oncology. Birte Moehlendick was financed by a grant from the German Research Foundation (DFG ST0464/2-2) to Nikolas H. Stoecklein.

Author Contributions

Ruben Jansen conducted all the experiments, collected and interpreted data, wrote the manuscript. Birte Moehlendick supervised and assisted the experiments and data analyses, reviewed the paper and performed literature research. Christoph Bartenhagen performed the aCGH data analyses. Csaba Tóth performed the histopathological analyses and interpretation of the slides. Nadja Lehwald performed literature research and reviewed the manuscript. Nikolas Stoecklein designed the study, supervised the experimental work and analysis, and assisted in the data interpretation. Wolfram T. Knoefel provided the tumor samples and assisted in the study design as well as in the data analysis. Anja Lachenmayer designed the study, supervised the data collection and analysis, and co-wrote the manuscript.

Additional Information

Supplementary information accompanies this paper at https://doi.org/10.1038/s41598-018-28941-6.

Competing Interests: The authors declare no competing interests.

Publisher's note: Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

Open Access This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made. The images or other third party material in this article are included in the article's Creative Commons license, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons license and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this license, visit http://creativecommons.org/licenses/by/4.0/.

© The Author(s) 2018