Main Title: Genomic and transcriptomic somatic alterations of hepatocellular carcinoma in non-cirrhotic livers

Short Title: Mutational landscape of non-cirrhotic HCC

Authors:
Zachary L Skidmore*, Jason Kunisaki*, Yiing Lin*, Kelsy C Cotto†, Erica K Barnell‡, Jasreet Hundal, Kilannin Krysiak, Vincent Magrini, Lee Trani, Jason R Walker, Robert Fulton, Elizabeth M Brunt, Christopher A Miller, Richard K Wilson, Elaine R Mardis, Malachi Griffith, William Chapman, Obi L Griffith

1. McDonnell Genome Institute, Washington University School of Medicine, St. Louis, MO
2. Department of Medicine, Washington University School of Medicine, St. Louis, MO
3. Department of Surgery, Washington University School of Medicine, St. Louis, MO
4. Department of Genetics, Washington University School of Medicine, St. Louis, MO
5. Emeritus Professor, Department of Pathology and Immunology, Washington University School of Medicine, St. Louis, MO
6. Siteman Cancer Center, Washington University School of Medicine, St Louis, MO

* These authors contributed equally
† Corresponding authors

Grant Support:
MG is funded by the National Human Genome Research Institute (NHGRI R00HG007940). OLG is funded by the National Cancer Institute (NIH NCI K22CA188163 and NIH NCI U01CA209936). This work, originally proposed by ERM as a center-initiated project, was also supported by a grant to RKW from the NHGRI (NIH NHGRI U54HG003079).

Correspondence:
Malachi Griffith
4444 Forest Park Ave,
St. Louis, MO, 63108,
Campus Box 8501
Email: mgriffit@wustl.edu

William Chapman
660 S. Euclid Ave,
St. Louis, MO, 63110,
Campus Box 8109
Email: chapmanwi@wudosis.wustl.edu

Obi L Griffith
4444 Forest Park Ave,
St. Louis, MO, 63108,
Campus Box 8501
Email: obigriffith@wustl.edu

Disclosures:
Erica K. Barnell is an owner, employee, and member of Geneoscopy Inc. Erica K. Barnell is an inventor of the intellectual property owned by Geneoscopy Inc. W. Chapman serves on the advisory board for Novartis Pharmaceutical and reports intellectual property with Pathfinder Therapeutics. The remaining authors state they have no conflicts of interest and nothing to disclose.

Author contributions:
ZLS, JK, YL, KCC, EKB, JH, KK, EMB, MG, WC, and OLG helped draft the manuscript. ZLS, JK, JH, MG, and OLG performed statistical analysis and interpreted data. YL, RKW, ERM, MG, WC, and OLG designed the study, acquired samples, and supervised the study. YL, VM, LT, JRW, RF, EMB, and CAM provided administrative, technical, and material support.

dbGAP accession: phs001106
Abstract

Background: Liver cancer is the second leading cause of cancer-related deaths worldwide. Hepatocellular carcinoma (HCC) risk factors include chronic hepatitis, cirrhosis, and alcohol abuse, whereby tumorigenesis is induced through inflammation and subsequent fibrotic response. However, a subset of HCC arises in non-cirrhotic livers. We characterized the genomic and transcriptomic landscape of non-cirrhotic HCC to identify features underlying the disease’s development and progression.

Methods: Whole genome and transcriptome sequencing was performed on 30 surgically resectable tumors comprised of primarily of non-cirrhotic HCC and adjacent normal tissue. Using somatic variants, capture reagents were created and employed on an additional 87 cases of mixed cirrhotic/non-cirrhotic HCC. Cases were analyzed to identify viral integrations, single nucleotide variants (SNVs), insertions and deletions (INDELS), copy number variants, loss of heterozygosity, gene fusions, structural variants, and differential gene expression.

Results: We detected 3,750 SNVs/INDELS and extensive CNVs and expression changes. Recurrent TERT promoter mutations occurred in >52% of non-cirrhotic discovery samples. Frequently mutated genes included TP53, CTNNB1, and APOB. Cytochrome P450 mediated metabolism was significantly downregulated. Structural variants were observed at MACROD2, WDPCP and NCKAP5 in >20% of samples. Furthermore, NR1H4 fusions involving gene partners EWSR1, GNPTAB, and FNIP1 were detected and validated in 2 non-cirrhotic samples.

Conclusion: Genomic analysis can elucidate mechanisms that may contribute to non-cirrhotic HCC tumorigenesis. The comparable mutational landscape between cirrhotic and non-cirrhotic HCC supports previous work suggesting a convergence at the genomic level during disease progression. It is therefore possible genomic-based treatments can be applied to both HCC subtypes with progressed disease.

Keywords: non-cirrhotic HCC; NR1H4; APOB

Highlights

- Non-cirrhotic HCC genomically resembles cirrhotic HCC
- Comprehensive genome- and transcriptome-wide profiling allows detection of novel structural variants, fusions, and undiagnosed viral infections
- NR1H4 fusions may represent a novel mechanism for tumorigenesis in HCC
- Non-cirrhotic HCC is characterized by genotoxic mutational signatures and dysregulated liver metabolism
- Clinical history and comprehensive omic profiling incompletely explain underlying etiologies for non-cirrhotic HCC highlighting the need for further research

Short Description

This study characterizes the genomic landscape of hepatocellular carcinomas (HCCs) in non-cirrhotic livers. Using 117 HCCs tumor/normal pairs, we identified 3,750 SNVs/INDELS with high variant frequency in TERT, TP53, CTNNB1, and APOB. CYP450 was significantly downregulated and many structural variants were observed. This characterization could assist in elucidating non-cirrhotic HCC tumorigenesis.

Introduction

Worldwide, there are approximately 750,000 new cases of hepatocellular carcinoma (HCC) each year [1]. Although HCC has the 5th highest incidence rate in men and 9th highest incidence rate in women, it has the second highest mortality rate of all cancer types [1]. HCC is traditionally associated with inflammation-inducing risk factors, which promote liver cirrhosis including: chronic hepatitis infections, such as hepatitis b virus (HBV) and hepatitis c virus (HCV), alcohol abuse, and non-alcoholic fatty liver disease [2]. However, approximately 20% of...
patients present with non-cirrhotic HCC in the absence of these risk factors [3]. If diagnosed early, patients with non-cirrhotic HCC maintain adequate liver function, allowing for effective tumor resection with exceptional prognosis when compared to patients with cirrhotic HCC [4]. However, late-stage diagnosis of non-cirrhotic HCC typically presents with larger and more aggressive tumors that are prone to metastasis [5]. Even with extensive tumor resection, approximately 50% of patients relapse within three years post-treatment [6].

Using high-throughput sequencing, researchers have previously characterized the genomic landscape of cirrhotic HCC [7–13]. These studies included whole genome, whole exome, and/or transcriptome sequencing with a focus on analyzing HCC induced by HBV, HCV, and/or cirrhosis. Prior studies, which have evaluated the genomics of cirrhotic and non-cirrhotic HCC, report that among the most significant and recurrent alterations are TERT mutations which typically occur at the promoter region [8,9,14]. Mutations within this region have been observed in a variety of cancer types beyond cirrhotic HCC, suggesting a common role of activating TERT promoter variants in oncogenesis and metastasis [15–17]. TERT expression in terminally differentiated cells promotes telomere maintenance and elongation [18]. Telomere maintenance is required for late-stage cancer propagation with TERT misregulation being harnessed by human cancers to evade mitotic catastrophe and apoptosis [19]. Previous studies have recognized that increases in TERT expression could serve as a proxy for telomere maintenance; however, late-stage tumors exhibit shortened telomeres in comparison to their normal counterparts, due to high turnover rates [20,21]. Among studies specific to cirrhotic HCC, the putative mechanisms of TERT activation can be divided into three categories: 1) HBV integration events in the TERT promoter [8,22], 2) point mutations (C228T and C250T) in the promoter region mutually exclusive of HBV integration [9,23], and 3) structural variations of the TERT promoter region [8,14].

This study characterizes biomarkers and elucidates recurrent anomalies in non-cirrhotic HCC. We identified somatic variants in 117 tumor samples whereby 52 samples were cirrhotic, 63 samples were non-cirrhotic, and 2 samples had an unspecified cirrhotic status. Using this cohort, we analyzed single nucleotide variants (SNVs), insertions and deletions (INDELs), structural variation (SV), copy number variation (CNV), loss of heterozygosity (LOH), differential expression, and viral integration events. This comprehensive approach uncovered the genomic features implicated in non-cirrhotic HCC to improve its diagnosis, prognosis, and treatment.

Methods
Refer to supplementary methods for more details.

Sample Procurement
The discovery cohort consisted of 30 primary tumor and adjacent matched non-tumor liver samples obtained through surgical resection from adult patients diagnosed with HCC between 2000 to 2011 at the Washington University School of Medicine. Within this cohort, 13 were male and 17 were female. Additionally, 2 were African American and 28 were Caucasian. None of these samples exhibited evidence of hepatocellular adenoma (HCA) and the non-cirrhotic samples did not show signs of advanced fibrosis. 1 sample was HBV positive and 4 samples were HCV positive according to clinical data. All other samples within the discovery cohort had an unknown clinical etiology. The extension-alpha and extension-beta cohorts had 16 HCC tumors with matched non-tumor liver and 71 tumor-only HCC samples, respectively. Discovery and extension-alpha cohort samples were flash-frozen prior to banking and extension-beta samples were derived from formalin fixed paraffin embedded (FFPE) blocks. Across both extension cohorts, 27 were female and 58 were male. Furthermore, 2 were Asian, 13 were African American, and 70 were Caucasian. Within the extension-alpha cohort, two samples
were HCV positive, one had chronic cholestasis, and the others had no known clinical etiology. Clinical data for the extension-beta cohort was as follows: 5 had known alcohol use, 8 were HBV positive, 29 were HCV positive, 2 were diagnosed with primary sclerosing cholangitis (PSC), and 6 samples were diagnosed with non-alcoholic steatohepatitis (NASH). From the extension-alpha cohort, 2 patients did not provide information on race and gender (Table S1). All patient samples were acquired after informed consent to an approved study by the Washington University School of Medicine Institutional Review Board (IRB 201106388).

Sample Preparation and Sequencing
DNA and RNA from samples in the discovery cohort were extracted using the QIAamp DNA Mini kit and Qiagen RNeasy Mini kit, respectively. Whole genome sequencing libraries were constructed using Kapa HYPER kits for use on the Illumina HiSeq 2000 platform. The Ovation RNA-seq System V2 (NuGen Inc) kit was used to generate RNAseq libraries. Resulting barcoded libraries were pooled prior to Illumina sequencing. To validate variants identified from WGS, a hybrid capture panel (CAP1) was designed and executed on the Illumina platform to capture fragments from the WGS libraries. The QIAamp DNA Mini kit was used to extract DNA from extension-alpha samples, which was subsequently sequenced using the CAP1 strategy. Finally, CAP1 sequencing was used to identify variants from the DNA extracted from extension-beta samples with the QIAamp DNA FFPE Tissue kit. A second hybrid capture panel (CAP2) utilized Nimblegen and spiked-in IDT probes that hybridized to the TERT promoter locus and HBV genome (designed against a consensus sequence for 10 common HBV strains, see supplementary methods). CAP2 sequencing was employed on all 117 samples. TERT promoter variants were also detected in the discovery and extension-alpha cohorts with Sanger sequencing. cDNA capture was performed on pooled samples from the extension cohorts.

Sequencing Alignment
WGS and CAP1 data were aligned to GRCh37 via the Genome Modeling System (GMS) using BWA [24,25]. Reads from the CAP2 data were competitively aligned using BWA [25] against the human reference genome (GRCh37) along with ten HBV genotypes for which complete genomes were available. RNAseq data were aligned with bowtie/tophat and expression was evaluated with cufflinks [26,27]. All raw RNAseq reads from the discovery cohort were also aligned against the HBV genomes for evidence of HBV expression at the RNA level. The predominant HBV strain was determined using relative coverage for competitive alignments. The precise location of the HBV integration site was identified from discordant read pairs from realigning HBV CAP2 reads to GRCh37 and the predominant HBV strain’s genome. A similar procedure was performed for HCV whereby both WGS and RNAseq reads were aligned against six HCV genotypes. The predominant HCV strain was determined using the total read support. To detect AAV1 and AAV2 integration, RNAseq reads were competitively aligned using kallisto [28] against AAV1 and AAV2 sequences.

Telomere Length Determination
Telomeric tumor:normal read ratios were determined from WGS data using the GMS and visualized in R. A Wilcoxon-Mann-Whitney test measured the significance of differences between telomere length in tumor and normal samples.

Variant Calling
Somatic variant analysis for single nucleotide variants (SNV) and insertions/deletions (INDEL) were performed on all three cohorts while germline variant analysis for these variants was performed on the discovery and extension-alpha cohort. Several computational tools within and outside of the GMS [29] were employed to facilitate variant calling and subsequent filtering based on variables including variant allele frequency, read count, and predicted pathogenicity.
Structural Variant, Copy Number Variant, and Loss of Heterozygosity Analysis
WGS data from samples within the discovery cohort were analyzed for structural variants (SV), copy number variation (CNV), and loss of heterozygosity (LOH). Manta [30] was used to identify SV events. Manta-reported breakpoints, along with a 10kb flank were annotated with biomaRt and ensembl (GRCh37.p13). Regions of CNV were identified with the GMS and LOH were identified using VarScan2 [29,31]. The DNAcopy circular binary segmentation algorithm generated segments of LOH and CNV, which served as input for GISTIC [32] to conduct a recurrence analysis.

NanoString nCounter Elements™ Tagsets: NR1H4 Fusion Validation
Fusion detection algorithms identified samples in the discovery cohort harboring gene fusions from RNAseq data. Fusion predictions involving NR1H4 were validated across all 117 samples using a NanoString nCounter® Elements™ TagSets assay. Sequences for predicted transcripts of the fusion calls that met certain read support criteria (≥10 spanning + encompassing reads and ≥1 spanning read) were sent to NanoString for probe design.

Survival and Clinical Analysis
The R “survival” package [33] was used to associate SV-affected genes and CNV/LOH-affected genomic regions with overall survival and recurrence free survival. Only mutated genes and genomic regions occurring in ≥ 4 discovery cohort samples were included in this analysis. A survival analysis was also applied to SNV/INDELs observed in all non-cirrhotic samples from the three cohorts. All Kaplan-Meier survival plots were created in R. Fisher’s exact test was used to test for clinical associations with variables: lymphovascular space invasion (LVSI), tumor differentiation status, cirrhosis, and liver disease. Samples without relevant clinical data were excluded. Significance was measured with a multiple test correction using the FDR methodology (q-value < 0.05).

Differential Expression and Pathway Analysis
Read counts for genes mutated in non-cirrhotic tumors and matched normal samples of the discovery cohort were used by the DEseq2 Bioconductor package [34] to perform differential gene expression analysis using a negative binomial distribution with samples as a blocking factor. Significance was measured with a Wald test and Benjamini & Hochberg multiple test correction (q-value < 0.5). Pathway analysis was performed using log2 differential expression data.

Results
Discovery Cohort
There were 30 patients included in the discovery cohort with tumors which were surgically resectable. These surgically resectable tumors were untreated, providing the opportunity to study HCC in the absence of chemotherapeutic intervention, which is normally incorporated in the treatment of cirrhotic HCC. Three of the patients within this cohort developed HCC in the setting of cirrhosis, all of which had been previously diagnosed with HCV. The remaining 27 individuals developed non-cirrhotic HCC, two of these individuals were diagnosed with HBV and another two individuals were diagnosed with HCV. To elucidate the genomic landscape of resected, primarily non-cirrhotic HCC, we performed whole genome sequencing (WGS), hybrid capture sequencing (CAP1), and transcriptome sequencing (RNAseq) on these 30 samples (Table 1). WGS failed for one tumor sample in the discovery cohort, therefore the final data for this cohort included WGS and CAP1 data for 29 samples (26 non-cirrhotic, 3 cirrhotic), and RNAseq data for 30 samples (27 non-cirrhotic, 3 cirrhotic). The sequencing analysis revealed a single previously unknown and undiagnosed HBV case with viral integration occurring at the
**TERT** promoter ([Figure S1, Table S1](#)). Median haploid coverage for WGS data was 35.6x (range: 28.5-39.3) and 58.4x (range: 46.8-94.4) for normal and tumor samples, respectively.

**Somatic mutations in the Discovery Cohort**

After filtering, we observed a median mutation burden of 1.31 mutations/Mb (range: 0.033-3.28), comprised of 2,633 SNVs and INDELs across all samples (range: 2-200, median: 77.5, mean=87.8) ([Figure 1, Table S1](#)). These variants were discovered across 2,245 genes with 258 of these genes mutated in more than one sample. Using WGS data from the 26 non-cirrhotic samples, we identified 6 genes that were significantly mutated above background mutation rates according to MuSiC: **ALB, APOB, CTNNB1, TP53, RB1, and RPS6KA3** ([Figure 1, Table S1](#)). With regards to all methods of sequencing (WGS, RNAseq, CAP1, and CAP2), the most frequently encountered variant was a SNV in the telomerase reverse transcriptase (**TERT**) promoter (C228T; G1295228A), which was identified in 17/30 samples and resulted in overexpression of **TERT** ([Figure S2, Table S2](#)). Within the exome, **TP53** was the most recurrently mutated gene and was observed in 8/29 of samples ([Table S1](#)). Beta catenin 1 (**CTNNB1**) was also significantly mutated within this cohort (6/29), whereby the majority of variants occurred at amino acids S37 and S45, both of which reside in a putative GSK3B phosphorylation site in exon 3 (ENST00000349496) ([Figure S3](#))[35]. Frameshift mutations in **APOB** were observed in 4/29 of samples ([Table S1](#)). Mutation signatures using the COSMIC database for the discovery cohort were investigated. Signatures 5 (unknown etiology), 4 (smoking damage association), 16 (unknown etiology), and 12 (liver damage association) were most prevalent and contributed to the overall cohort signature at 23%, 14%, 8%, and 7%, respectively ([Figure S4](#)).

**Transcriptome Analysis of the Discovery Cohort**

Differential gene expression analysis performed on the non-cirrhotic samples revealed that 11% of genes, including **TERT**, were upregulated (4,468/39,392) and 10% of genes, including **CTNNB1** and **WISP2**, were downregulated (4,114/39,392) compared to adjacent non-tumor liver tissue (q-value < 0.1) ([Table S1](#)). Comparison of gene log2 fold changes derived from the differential expression analysis revealed the cell cycle pathway as upregulated in the KEGG signaling and metabolism database (q-value ≤ 0.05). Similarly, we observed 16 pathways as down-regulated (q-value ≤ 0.05), most of which are related to metabolic liver processes. Genes such as **ADH5** and **EHHADH** were observed with reduced expression levels and participate in 38% (6/16) of these pathways. Using the Gene Ontology biological process database, we observed 107 pathways as significantly upregulated (q-value ≤ 0.05). The majority of the upregulated pathways were related to cellular division and DNA repair. In addition, 28 pathways were identified as significantly downregulated (q-value ≤ 0.05), many of which were related to liver metabolism ([Table S2](#)).

**Telomere lengths in the Discovery Cohort**

When evaluating the samples within the discovery cohort for telomere length at the DNA level, we observed that the majority of tumor samples exhibited shortened telomeres compared to their paired normal sample (p-value = 0.00011) ([Figure S2](#)). One exception was seen in sample HCC16_D, which was distinguished by abnormally high expression of **TERT** (FPKM=36) ([Figure S5](#)).

**Copy Number Variants and Loss of Heterozygosity in the Discovery Cohort**

We observed recurrent large scale amplification of the q-arm of chromosome 1 in ≥ 50% of the discovery cohort. Similarly, large scale deletions of the p-arms of chromosomes 8 and 17 were found in ≥ 40% of the cohort ([Figure 2](#)). In total, analysis with GISTIC and subsequent manual review revealed 75 unique regions across 17 chromosomes as recurrently amplified and 45
unique regions across 17 chromosomes as significantly deleted ($q < 0.05$) (Table S1). No significant associations with tumor differentiation status were made ($a=0.05$). Each CNV and LOH event was tested for their association with overall survival and recurrence free survival but no significant association could be made following multiple test correction. A total of 33 genes identified as recurrently deleted by GISTIC showed concordant decreased expression in tumor samples (Table S1). These include genes previously characterized as relevant to HCC development and progression: $HEYL$ [36] ($q$-value = 0.032), $UQCRH$ [37] ($q$-value = 0.032), and $MUTYH$ [38] ($q$-value = 0.048). A subset of these genes have also been implicated in tumorigenesis, metastasis, and progression of other cancer types and may prove to be relevant for HCC development and progression: $RPL11$ [39] ($q$-value = 0.048), $UBE2D3$ [40] ($q$-value = 0.032), $ARRB1$ [41] ($q$-value = 0.032), $ENG$ [42] ($q$-value = 0.049), and $ABLIM2$ [43] ($q$-value = 0.032).

A GISTIC analysis coupled with manual review identified 8 unique regions exhibiting recurrent LOH affecting the chromosomal arms of 6q, 8p, 13q, and 17p (Figure 2, Table S1). The 8p and 17p chromosomal arms were most susceptible to LOH, each occurring in $\geq 30\%$ of samples. The 8 genomic regions identified as recurrently affected by LOH contain the coding regions for: $TP53$, $RB1$, $DLC1$, $PFN1$, $ARID1B$, $LAMA2$, and $CLU$ (Table S1).

**Structural Variation in the Discovery and Extension Cohorts**

We identified 4,745 SV events affecting 3,801 genes across the discovery cohort, of which 737 were deletions, 1,650 were duplications, 450 were inversions, and 1,908 were translocations. Three genes were affected by SVs in $\geq 20\%$ of samples, $NCKAP5$ (N=11), $WDPCP$ (N=6), and $MACROD2$ (N=6) (Table S1). Translocations near the $TERT$ promoter region occurred in 2/29 samples, both of which were non-cirrhotic (Figure S5). Additionally, we detected a recurrent fusion involving $NR1H4$ with a diverse set of gene partners (173 fusion predictions) (Table S1). NanoString nCounter Elements™ Technology was used to assess 21 of these fusions. NanoString validated three $NR1H4$ fusions that were partnered with $EWSR1$, $GNPTAB$, and $FNIP1$ in 2 samples (Figure 3). One fusion ($CDK17-NR1H4$) was called by both Integrate and ChimeraScan but was not validated using NanoString. BLAT alignments revealed that these fusions result in sequence frameshifts and thereby likely inactivate the function of $NR1H4$.

**Variant Detection in Extension Cohorts**

To further study recurrently mutated genes and discover novel events intrinsic to non-cirrhotic HCC, we employed CAP1 on extension-alpha and extension-beta samples (N=87). In addition to variants identified in the discovery cohort, this extension study elucidated 69 and 1,022 variants in the extension-alpha and extension-beta cohorts, respectively (Figure S6). Using the variants identified with CAP1, we classified a total of 17 genes as significantly mutated ($q$-value $\leq 0.05$) (Table S1). Of the significantly mutated genes in the discovery cohort, all were confirmed as significantly mutated in the extension cohorts with the exception of $RB1$ and $RPS6KA3$. We tested for differences between mutated genes based on cirrhosis status within all cohorts using CAP1, but we were unable to identify any significant differences ($q$-value $\leq 0.05$).

**Germline mutations in Discovery and Extension-Alpha Cohorts**

Within the discovery cohort and extension-alpha cohorts, there were 4 genes recurrently mutated (24 tumors) in germline DNA after filtering, including: $AL356585.1$, $MUC19$, $SVIL$, and $DNAH5$. In addition, when evaluating deleterious calls predicted by four different methods (SIFT [44], Polyphen [45], ClinVar [46], CADD [47]), there were 11 variants that were identified as pathogenic. Of these 11 variants, 5 were in autosomal genes ($LAMA2$, $CYP4V2$, $SLC22A5$, $BMPR2$, $SLC26A4$) and 6 were in mitochondrial genes ($MT-CO1$, $MT-CO1$, $MT-ND3$, $MT-ND3$, $MT-CYB$, $MT-ND1$) (Table S1).
Viral Integration and TERT Promoter Mutation in Discovery and Extension Cohorts

Viral detection of HBV and HCV in samples from the discovery and extension cohorts was conducted using CAP2. This analysis validated the clinical diagnosis of HBV in 7 of 9 HBV+ HCC samples. Additionally, a clinically undiagnosed sample in the discovery cohort (HCC18_D) was shown to possess HBV infection both at the RNA and DNA level (Table S1). Manual review and BLAT analysis confirmed that HBV integrated at the TERT promoter locus in 1 out of the 10 HBV positive samples (Figure S1). Similarly HCV, a RNA virus, was detected in 5 samples in the discovery cohort, 3 of which confirmed a clinical HCV diagnosis. Within the extension cohort, 29 samples were clinically diagnosed with HCV. Competitive alignments using kallisto [28] were performed to detect integration of AAV1 and AAV2 in samples from the discovery cohort. AAV1 and AAV2 were both detected in one sample. AAV2 was detected in two other samples (Table S1).

Mutations in the TERT promoter region were detected in samples in the discovery and extension cohorts with WGS, Sanger sequencing, and the CAP2 panel using Nimblegen and spiked-in IDT probes. Point mutations in this region (C228T or C250T) were observed in 52.4% of the samples. Samples infected with HCV, determined by clinical assay, were found to be significantly enriched for TERT promoter mutations (p-value = 0.0051, Table S1); however this was no longer significant following a multiple test correction (q>0.05).

Clinical Associations and Survival Analysis

Among non-cirrhotic HCC samples in the discovery cohort, TERT promoter alterations (point mutations, HBV integration, and translocations) were not significantly associated with overall survival or recurrence free survival (Figure 4). No significant associations were observed between any other variants (SNV/INDEL, CNV, LOH, and SV) and clinical variables including lymphovascular space invasion, tumor differentiation, and tumor predisposition (e.g. HBV/HCV infection, alcohol abuse, cirrhosis, etc.).

Discussion

Genomic Landscape of Non-cirrhotic HCC Largely Resembles that of Cirrhotic HCC

The overarching purpose of this study was to test if genomic differences exist between non-cirrhotic HCC and HCC developed in the background of cirrhosis. Within our non-cirrhotic discovery cohort of 26 patients sequenced by WGS, the median mutation burden and recurrent somatic/germline mutations closely resembled those previously reported in cirrhotic HCC [48]. Recurrently mutated genes in the discovery cohort included TERT (55%), TP53 (28%), CTNNB1 (21%), and APOB (13%), all of which have been previously observed in cirrhotic HCC at similar frequencies [9,48]. We did not find any significant difference when comparing recurrently mutated genes between 52 cirrhotic and 63 non-cirrhotic samples for which cirrhotic status was available using CAP1. The LOH events [10,49–52] and CNVs [9–11,52,53] observed within our cohort are similar to those previously observed in cirrhotic HCC using alternative, investigative approaches including SNP array analysis. Within these regions, several genes such as LAMA2, ATK3, EFF1A1, and PFN1 [14,54–56] have been previously investigated in the context of HCC development and progression.

Structural variants identified in non-cirrhotic patients’ samples were also largely similar to SVs reported in cirrhotic HCC. However, of the three genes that were mutated in over 20% of samples in the discovery cohort, only MACROD2 had been previously described. With its ADP-ribosyl-hydrolase activity, MACROD2 has a well characterized role in reversing ADP-ribosylation of proteins involved in a variety of cellular processes, including the DNA repair pathway [57]. A study in Japan identified 32/268 HCC samples (cirrhosis status not specified)
with SVs involving MACROD2, which was associated with a better overall prognosis [8]. Though our data did not support this same clinical association, our findings suggest a similar mutation frequency of MACROD2 in non-cirrhotic HCC. The observed TERT structural variants, which occurred in 2 non-cirrhotic samples, have also been detected in cirrhotic HCC. This study corroborates previous findings of TERT activation via SVs, HBV integration, and point mutations in HCC patients [8,9].

Several studies have associated accumulation of β-catenin with cirrhotic HCC tumorigenesis [58]. However, in cases of HCC with reduced inflammation and fibrosis, researchers have observed an absence of β-catenin accumulation [59]. It is therefore intriguing that despite observing activating mutations within exon 3 of CTNNB1, we observe CTNNB1 downregulation in non-cirrhotic samples of the discovery cohort. Accumulation of β-catenin is possible in the context of CTNNB1 down-regulation since exon 3 mutations serve to prevent the degradation of the β-catenin protein [58]. The majority of other studies observe accumulation of β-catenin in the context of a primarily asian cohort whereas our cohort represents a western population. Future investigations are required to understand the role of CTNNB1 expression and β-catenin accumulation in non-cirrhotic HCC.

**Biological Pathways in Liver are Dysregulated**
Investigations into significantly enriched pathways in non-cirrhotic samples from the discovery cohort revealed broad downregulation of pathways related to liver metabolism (GO [60,61], KEGG [62–64]) and upregulation of pathways involved in cellular division and replication (GO) (Table S2). This is expected given that tumor cells are dividing more frequently and losing normal liver function. We observed downregulation of pathways linked to cytochrome P450 (CYP450) mediated xenobiotic metabolism. Previous investigations of CYP activity and expression in cirrhotic and HBV infected HCC demonstrate that CYP activity is dysregulated in HCC tumor cells [65]. Given the potential for CYPs to facilitate individualized treatment options for HCC patients, it is possible treatment strategies for non-cirrhotic HCC may also involve CYPs.

**Survival Analysis Does Not Identify Prognostic Potential for Observed Mutations**
A survival analysis was performed on non-cirrhotic samples of the discovery cohort to identify large scale mutational events (SVs, CNVs, and LOH) that may serve as prognostic biomarkers; however, no significant association could be made. The survival analysis was also extended to SNVs and INDELs. TERT promoter alterations have been associated with poor prognosis in cancers such as glioblastomas [15,66] and melanomas [16]. In terms of HCC however, previous reports that investigated TERT promoter status with respect to survival yield conflicting results [23,67]. We found no association between TERT promoter mutations and prognosis. Taken as a whole, it appears activating TERT mutations serve as a common mechanism for tumorigenesis but additional investigations are required to definitively determine whether or not these mutations can serve a prognostic role in patients afflicted with HCC.

**Low-frequency, Novel Mutations in APOB and NR1H4 and Novel Structural Variants Identified**
In this study, we identified predicted loss of function/damaging somatic and germline APOB mutations in 6/26 non-cirrhotic samples in the discovery cohort (Table S1). Among the somatic APOB mutations, all were observed at a relatively high VAF (>0.25), suggesting these mutations were present early in tumorigenesis and may be present in pre-malignant sites within these tissues. It has been shown that abnormal glycosylation resulting from the formation of bisecting-GlcNAc disrupts APOB function and leads to a fatty liver disease phenotype [68]. We
observed APOB mutations outside of a fatty liver disease phenotype suggesting a more prominent role in tumorigenesis.

We also identified a novel recurrent somatic, loss of function, gene fusion event involving NR1H4 in 2 non-cirrhotic samples within the discovery cohort (Table S1). Low sample quality (particularly FFPE samples) and use of RNA-based validation methods may have prevented sensitive detection of these fusion events resulting in underestimation of fusion frequency in additional cases. Biologically, NR1H4 has a well-defined role to prevent the accumulation of bile acid (BA) within the liver, which could otherwise lead to HCC development. In vivo studies have demonstrated that NR1H4 loss predisposes mice to spontaneous hepatocarcinogenesis [69,70] and obstructs hepatocyte regeneration following partial hepatectomy [71]. NR1H4 fusions may therefore represent a novel mechanism of HCC development.

Other novel SV events that were observed in the discovery cohort involved NCKAP5 and WDPCP, both of which play a role in cilia function. NCKAP5L, which is a paralog to NCKAP5, functions to stabilize and strengthen microtubule structure [72]. Additionally, reduced expression of WDPCP has been shown to inhibit proper ciliogenesis in the presence of proinflammatory cytokines [73]. Given the previous association of ciliopathies with cancer, these recurrent structural variants in non-cirrhotic HCC might play a role in tumorigenesis and metastasis [74].

Genotoxic and virologic etiologies partially explain Non-Cirrhotic HCC

Recent work by Zucman-Rossi et al. has outlined etiologies for HCC development [75]. Within our discovery cohort, 23/29 samples exhibited a mutational signature of T->C mutations at an ApTpN context (weight > 0.1) (Table 2, Figure S4). These signatures (signature 5,16) have been associated with genotoxic injury and were previously observed in HCC patients with high alcohol and tobacco consumption. We do not have data on tobacco consumption; however, most of these cases did not have reported alcohol consumption (17/23), which suggests that another unknown factor may be contributing to this observation. Among these 23 samples, a non-cirrhotic case exhibited a signature consistent with aflatoxin exposure (signature 24) and harbored an R249S mutation in TP53. Interestingly this case was clinically diagnosed with HBV, which has been suggested to have a synergistic effect with aflatoxin exposure, facilitating HCC development [75,76]. In addition, we observed a single HCV positive case which also possessed signatures 5 and 16. It is curious that while HCV is a predisposing factor for liver cirrhosis, and this patient exhibited cirrhosis, etiologies associated with signatures 5 and 16 are not typically associated with cirrhosis [75]. With the exception of a final HBV case, we could not identify an etiology for the remaining 5 samples. The dysregulation of liver metabolism identified in the pathway analysis may represent a cause or symptom of the etiologies leading to tumorigenesis. This could apply to the unknown etiologies in the remaining samples, suggesting liver dysfunction is necessary for disease progression. Emerging evidence suggests that HCC can develop in the context of nonalcoholic fatty liver disease (NAFLD) or NASH, even in the absence of advanced fibrosis or cirrhosis [77–79]. Our understanding and recognition of the clinical features associated with NASH and NAFLD has evolved. Therefore, it is likely that patients were affected by these conditions, but not clinically diagnosed at the time, within our discovery cohort. Further research is needed to elucidate the relationship between etiologies of HCC and their association with liver metabolism.

Conclusion

It has been observed that the underlying etiologies contributing to tumorigenesis of non-cirrhotic and cirrhotic HCC are unique [75]. Despite distinct evolutions of these tumor subtypes, our findings describe a convergence of both subtypes onto a similar genomic landscape during disease progression. This genomic similarity suggests in vitro and in vivo models for
investigating HCC biology may be relevant to both HCC subtypes for advanced disease. Clinically, genomic-based diagnostic, prognostic, and treatment strategies that were previously established in patients with cirrhotic disease may also be extended to patients with progressed non-cirrhotic HCC.
BIBLIOGRAPHY AND REFERENCES CITED

[1] Ferlay J, Soerjomataram I, Dikshit R, Eser S, Mathers C, Rebelo M, et al. Cancer incidence and mortality worldwide: sources, methods and major patterns in GLOBOCAN 2012. Int J Cancer 2015;136:E359–86.

[2] Fattovich G, Stroffolini T, Zagni I, Donato F. Hepatocellular carcinoma in cirrhosis: incidence and risk factors. Gastroenterology 2004;127:S35–50.

[3] Alkofer B, Lepennec V, Chiche L. Hepatocellular cancer in the non-cirrhotic liver. J Visc Surg 2011;148:3–11.

[4] Maeda T, Shimada M, Harimoto N, Tsujita E, Aishima S-I, Tanaka S, et al. Prognosis of early hepatocellular carcinoma after hepatic resection. Hepatogastroenterology 2008;55:1428–32.

[5] Llovet JM, Brú C, Bruix J. Prognosis of hepatocellular carcinoma: the BCLC staging classification. Semin Liver Dis 1999;19:329–38.

[6] Shah SA, Cleary SP, Wei AC, Yang I, Taylor BR, Hemming AW, et al. Recurrence after liver resection for hepatocellular carcinoma: risk factors, treatment, and outcomes. Surgery 2007;141:330–9.

[7] Schulze K, Imbeaud S, Letouzé É, Alexandrov LB, Calderaro J, Rebouissou S, et al. Exome sequencing of hepatocellular carcinomas identifies new mutational signatures and potential therapeutic targets. Nat Genet 2015;47:505–11.

[8] Fujimoto A, Furuta M, Totoki Y, Tsunoda T, Kato M, Shiraishi Y, et al. Whole-genome mutational landscape and characterization of noncoding and structural mutations in liver cancer. Nat Genet 2016;48:500–9.

[9] Cancer Genome Atlas Research Network. Electronic address: wheeler@bcm.edu, Cancer Genome Atlas Research Network. Comprehensive and Integrative Genomic Characterization of Hepatocellular Carcinoma. Cell 2017;169:1327–41.e23.

[10] Laurent-Puig P, Zucman-Rossi J. Genetics of hepatocellular tumors. Oncogene 2006;25:3778–86.

[11] Guichard C, Amaddeo G, Imbeaud S, Ladeiro Y, Pelletier L, Maad IB, et al. Integrated analysis of somatic mutations and focal copy-number changes identifies key genes and pathways in hepatocellular carcinoma. Nat Genet 2012;44:694–8.

[12] Kan Z, Zheng H, Liu X, Li S, Barber TD, Gong Z, et al. Whole-genome sequencing identifies recurrent mutations in hepatocellular carcinoma. Genome Res 2013;23:1422–33.

[13] Fujimoto A, Totoki Y, Abe T, Boroevich KA, Hosoda F, Nguyen HH, et al. Whole-genome sequencing of liver cancers identifies etiological influences on mutation patterns and recurrent mutations in chromatin regulators. Nat Genet 2012;44:760–4.

[14] Jhunjhunwala S, Jiang Z, Stawiski EW, Gnad F, Liu J, Mayba O, et al. Diverse modes of genomic alteration in hepatocellular carcinoma. Genome Biol 2014;15:436.

[15] Mosrati MA, Malmström A, Lysiak M, Krysztofiak A, Hallbeck M, Milos P, et al. TERT promoter mutations and polymorphisms as prognostic factors in primary glioblastoma. Oncotarget 2015;6:16663–73.

[16] Huang FW, Hodis E, Xu MJ, Kryukov GV, Chin L, Garraway LA. Highly recurrent TERT promoter mutations in human melanoma. Science 2013;339:957–9.

[17] Hosen I, Rachakonda PS, Heidenreich B, de Verdier PJ, Ryk C, Steineck G, et al. Mutations in TERT promoter and FGFR3 and telomere length in bladder cancer. Int J Cancer 2015;137:1621–9.

[18] Jafri MA, Ansari SA, Alqahtani MH, Shay JW. Roles of telomeres and telomerase in cancer, and advances in telomerase-targeted therapies. Genome Med 2016;8:69.

[19] Blasco MA. Telomeres and human disease: ageing, cancer and beyond. Nat Rev Genet 2005;6:611–22.

[20] Saini N, Srinivasan R, Chawla Y, Sharma S, Chakraborti A, Rajwanshi A. Telomerase...
activity, telomere length and human telomerase reverse transcriptase expression in hepatocellular carcinoma is independent of hepatitis virus status. Liver Int 2009;29:1162–70.

[21] Fredriksson NJ, Ny L, Nilsson JA, Larsson E. Systematic analysis of noncoding somatic mutations and gene expression alterations across 14 tumor types. Nat Genet 2014;46:1258–63.

[22] Jiang Z, Jhunjhunwala S, Liu J, Haverty PM, Kennemer MI, Guan Y, et al. The effects of hepatitis B virus integration into the genomes of hepatocellular carcinoma patients. Genome Res 2012;22:593–601.

[23] Kawai-Kitahata F, Asahina Y, Tanaka S, Kakinuma S, Murakawa M, Nitta S, et al. Comprehensive analyses of mutations and hepatitis B virus integration in hepatocellular carcinoma with clinicopathological features. J Gastroenterol 2016;51:473–86.

[24] Griffith M, Griffith OL, Smith SM, Ramu A, Callaway MB, Brummett AM, et al. Genome Modeling System: A Knowledge Management Platform for Genomics. PLoS Comput Biol 2015;11:e1004274.

[25] Li H, Durbin R. Fast and accurate short read alignment with Burrows-Wheeler transform. Bioinformatics 2009;25:1754–60.

[26] Trapnell C, Williams BA, Pertea G, Mortazavi A, Kwan G, van Baren MJ, et al. Transcript assembly and quantification by RNA-Seq reveals unannotated transcripts and isoform switching during cell differentiation. Nat Biotechnol 2010;28:511–5.

[27] Trapnell C, Pachter L, Salzberg SL. TopHat: discovering splice junctions with RNA-Seq. Bioinformatics 2009;25:1105–11.

[28] Bray NL, Pimentel H, Melsted P, Pachter L. Near-optimal probabilistic RNA-seq quantification. Nat Biotechnol 2016;34:525–7.

[29] Griffith M, Griffith OL, Smith SM, Ramu A, Callaway MB, Brummett AM, et al. Genome Modeling System: A Knowledge Management Platform for Genomics. PLoS Comput Biol 2015;11:e1004274.

[30] Chen X, Schulz-Trieglaff O, Shaw R, Barnes B, Schlesinger F, Källberg M, et al. Manta: rapid detection of structural variants and indels for germline and cancer sequencing applications. Bioinformatics 2016;32:1220–2.

[31] Koboldt DC, Zhang Q, Larson DE, Shen D, McLellan MD, Lin L, et al. VarScan 2: somatic mutation and copy number alteration discovery in cancer by exome sequencing. Genome Res 2012;22:568–76.

[32] Mermel CH, Schumacher SE, Hill B, Meyerson ML, Beroukhim R, Getz G. GISTIC2.0 facilitates sensitive and confident localization of the targets of focal somatic copy-number alteration in human cancers. Genome Biol 2011;12:R41.

[33] Therneau TM, Grambsch PM. Modeling Survival Data: Extending the Cox Model. Springer Science & Business Media; 2013.

[34] Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. Genome Biol 2014;15. https://doi.org/10.1186/s13059-014-0550-8.

[35] Miyoshi Y, Iwao K, Nagasawa Y, Aihara T, Sasaki Y, Imaoka S, et al. Activation of the beta-catenin gene in primary hepatocellular carcinomas by somatic alterations involving exon 3. Cancer Res 1998;58:2524–7.

[36] Kuo K-K, Jian S-F, Li Y-J, Wan S-W, Weng C-C, Fang K, et al. Epigenetic inactivation of transforming growth factor-β1 target gene HEYL, a novel tumor suppressor, is involved in the P53-induced apoptotic pathway in hepatocellular carcinoma. Hepatol Res 2014;45:782–93.

[37] Park E-R, Kim S-B, Lee J-S, Kim Y-H, Lee D-H, Cho E-H, et al. The mitochondrial hinge protein, UQCRH, is a novel prognostic factor for hepatocellular carcinoma. Cancer Med 2017;6:749–60.
Krupa R, Czarny P, Wigner P, Wozny J, Jablkowski M, Kordek R, et al. The Relationship Between Single-Nucleotide Polymorphisms, the Expression of DNA Damage Response Genes, and Hepatocellular Carcinoma in a Polish Population. DNA Cell Biol 2017. https://doi.org/10.1089/dna.2017.3664.

Takafuji T, Kayama K, Sugimoto N, Fujita M. GRWD1, a new player among oncogenesis-related ribosomal/nucleolar proteins. Cell Cycle 2017:1–7.

Guan GG, Wang WB, Lei BX, Wang QL, Wu L, Fu ZM, et al. UBE2D3 is a positive prognostic factor and is negatively correlated with hTERT expression in esophageal cancer. Oncol Lett 2015;9:1567–74.

Miele E, Po A, Begalli F, Antonucci L, Mastronuzzi A, Marras CE, et al. β-arrestin1-mediated acetylation of Gli1 regulates Hedgehog/Gli signaling and modulates self-renewal of SHH medulloblastoma cancer stem cells. BMC Cancer 2017;17:488.

Kokaji E, Shimomura A, Minamisaka T, Nakajima T, Miwa S, Hatta H, et al. Endoglin (CD105) and SMAD4 regulate spheroid formation and the suppression of the invasive ability of human pancreatic cancer cells. Int J Oncol 2018;52:892–900.

Hwang SJ, Lee HW, Kim HR, Song HJ, Lee DH, Lee H, et al. Overexpression of microRNA-95-3p suppresses brain metastasis of lung adenocarcinoma through downregulation of cyclin D1. Oncotarget 2015;6:20434–48.

Kumar P, Henikoff S, Ng PC. Predicting the effects of coding non-synonymous variants on protein function using the SIFT algorithm. Nat Protoc 2009;4:1073–81.

Adzhubei I, Jordan DM, Sunyaev SR. Predicting functional effect of human missense mutations using PolyPhen-2. Curr Protoc Hum Genet 2013;Chapter 7:Unit7.20.

Landrum MJ, Lee JM, Benson M, Brown GR, Chao C, Chitipiralla S, et al. ClinVar: improving access to variant interpretations and supporting evidence. Nucleic Acids Res 2017;46:D1062–7.

Rentzsch P, Witten D, Cooper GM, Shendure J, Kircher M. CADD: predicting the deleteriousness of variants throughout the human genome. Nucleic Acids Res 2019;47:D886–94.

Lee J-S. The mutational landscape of hepatocellular carcinoma. Clin Mol Hepatol 2015;21:220–9.

Zhao Z, Chen G-Y, Long J, Li H, Huang J. Genomic losses at 5q13.2 and 8p23.1 in dysplastic hepatocytes are common events in hepatitis B virus-related hepatocellular carcinoma. Oncol Lett 2015;9:2839–46.

Zhu Q, Gong L, Liu X, Wang J, Ren P, Zhang W, et al. Loss of heterozygosity at D8S262: an early genetic event of hepatocarcinogenesis. Diagn Pathol 2015;10:70.

Herath NI, Leggett BA, MacDonald GA. Review of genetic and epigenetic alterations in hepatocarcinogenesis. J Gastroenterol Hepatol 2006;21:15–21.

Hashimoto K, Mori N, Tamesa T, Okada T, Kawauchi S, Oga A, et al. Analysis of DNA copy number aberrations in hepatitis C virus-associated hepatocellular carcinomas by conventional CGH and array CGH. Mod Pathol 2004;17:617–22.

Chen S-L, Lu S-X, Liu L-L, Wang C-H, Yang X, Zhang Z-Y, et al. eEF1A1 Overexpression Enhances Tumor Progression and Indicates Poor Prognosis in Hepatocellular Carcinoma.
Palazzo L, James DI, Waddell ID, Ahel I. Studying Catabolism of Protein ADP-Ribosylation. Methods Mol Biol 2017;1608:415–30.

Okabe H, Kinoshita H, Imai K, Nakagawa S, Higashi T, Arima K, et al. Diverse Basis of β-Catenin Activation in Human Hepatocellular Carcinoma: Implications in Biology and Prognosis. PLoS One 2016;11:e0152695.

Lee JM, Yang J, Newell P, Singh S, Parwani A, Friedman SL, et al. β-Catenin signaling in hepatocellular cancer: Implications in inflammation, fibrosis, and proliferation. Cancer Lett 2014;343:90–7.

Ashburner M, Ball CA, Blake JA, Botstein D, Butler H, Cherry JM, et al. Gene ontology: tool for the unification of biology. The Gene Ontology Consortium. Nat Genet 2000;25:25–9.

Gene Ontology Consortium. The Gene Ontology resource: enriching a GOLd mine. Nucleic Acids Res 2021;49:D325–34.

Kanehisa M, Goto S. KEGG: kyoto encyclopedia of genes and genomes. Nucleic Acids Res 2000;28:27–30.

Kanehisa M. Toward understanding the origin and evolution of cellular organisms. Protein Sci 2019;28:1947–51.

Kanehisa M, Furumichi M, Sato Y, Ishiguro-Watanabe M, Tanabe M. KEGG: integrating viruses and cellular organisms. Nucleic Acids Res 2021;49:D545–51.

Yan T, Lu L, Xie C, Chen J, Peng X, Zhu L, et al. Severely Impaired and Dysregulated Cytochrome P450 Expression and Activities in Hepatocellular Carcinoma: Implications for Personalized Treatment in Patients. Mol Cancer Ther 2015;14:2874–86.

Simon M, Hosen I, Gousias K, Rachakonda S, Heidenreich B, Gessi M, et al. TERT promoter mutations: a novel independent prognostic factor in primary glioblastomas. Neuro Oncol 2015;17:45–52.

Lee HW, Park TI, Jang SY, Park SY, Park W-J, Jung S-J, et al. Clinicopathological characteristics of TERT promoter mutation and telomere length in hepatocellular carcinoma. Medicine 2017;96:e5766.

Ihara Y, Yoshimura M, Miyoshi E, Nishikawa A, Sultan AS, Toyosawa S, et al. Ectopic expression of N-acetylglucosaminyltransferase III in transgenic hepatocytes disrupts apolipoprotein B secretion and induces aberrant cellular morphology with lipid storage. Proc Natl Acad Sci U S A 1998;95:2526–30.

Kim I, Morimura K, Shah Y, Yang Q, Ward JM, Gonzalez FJ. Spontaneous hepatocarcinogenesis in farnesoid X receptor-null mice. Carcinogenesis 2007;28:940–6.

Degirolamo C, Modica S, Vaccia M, Di Tullio G, Morgano A, D’Orazio A, et al. Prevention of spontaneous hepatocarcinogenesis in farnesoid X receptor-null mice by intestinal-specific farnesoid X receptor reactivation. Hepatology 2015;61:161–70.

Wang Y-D, Chen W-D, Moore DD, Huang W. FXR: a metabolic regulator and cell protector. Cell Res 2008;18:1087–95.

Mori Y, Inoue Y, Tanaka S, Doda S, Yamanaka S, Fukuchi H, et al. Cep169, a Novel Microtubule Plus-End-Tracking Centrosomal Protein, Binds to CDK5RAP2 and Regulates Microtubule Stability. PLoS One 2015;10:e0140968.

Ma Y, Sun Y, Jiang L, Zuo K, Chen H, Guo J, et al. WDPCP regulates the ciliogenesis of human sinonasal epithelial cells in chronic rhinosinusitis. Cytoskeleton 2017;74:82–90.

Hassounah NB, Nunez M, Fordyce CA, Roe DJ, Nagle RB, Bunch TA, et al. Inhibition of Ciliogenesis Promotes Hedgehog Signaling, Tumorigenesis, and Metastasis in Breast Cancer. Mol Cancer Res 2017. https://doi.org/10.1158/1541-7786.MCR-17-0034.

Zucman-Rossi J, Villanueva A, Nault J-C, Llovet JM. Genetic Landscape and Biomarkers of Hepatocellular Carcinoma. Gastroenterology 2015;149:1226–39.e4.

C. M, Michael C. Synergistic Interaction Between Aflatoxin and Hepatitis B Virus in Hepatocarcinogenesis. Aflatoxins - Recent Advances and Future Prospects, 2013.
[77] Huang DQ, El-Serag HB, Loomba R. Global epidemiology of NAFLD-related HCC: trends, predictions, risk factors and prevention. Nat Rev Gastroenterol Hepatol 2021;18:223–38.
[78] Calzadilla Bertot L, Adams LA. The Natural Course of Non-Alcoholic Fatty Liver Disease. Int J Mol Sci 2016;17. https://doi.org/10.3390/ijms17050774.
[79] Kucukoglu O, Sowa J-P, Mazzolini GD, Syn W-K, Canbay A. Hepatokines and adipokines in NASH-related hepatocellular carcinoma. J Hepatol 2021;74:442–57.
[80] Skidmore ZL, Wagner AH, Lesurf R, Campbell KM, Kunisaki J, Griffith OL, et al. GenVisR: Genomic Visualizations in R. Bioinformatics 2016;32:3012–4.
[81] Skidmore ZL, Campbell KM, Cotto KC, Griffith M, Griffith OL. Exploring the Genomic Landscape of Cancer Patient Cohorts with GenVisR. Curr Protoc 2021;1:e252.
[82] Zhou X, Edmonson MN, Wilkinson MR, Patel A, Wu G, Liu Y, et al. Exploring genomic alteration in pediatric cancer using ProteinPaint. Nat Genet 2016;48:4–6.
Figure 1: Genomic landscape of the non-cirrhotic discovery cohort exhibits similarity with cirrhotic HCC.
GenVisR [80,81] depiction of coding somatic mutations, structural variants, \textit{TERT} promoter mutations (G1295228A), and validated fusions are shown for samples in the discovery cohort which were recurrently (>25%) or significantly mutated. Where there are multiple mutations for the same gene/sample, the most severe mutation is displayed (severity follows the order listed in the legend). The percentage of samples for which a gene is mutated is shown on the left. Mutation Frequency represents the total number of mutations within individual samples. DUP=duplication; DEL=deletion; TRA=translocation.
Figure 2: Genome wide CNV and LOH in the discovery cohort
Recurrent regions of LOH (A) and CNV amplification/deletion (B) are shown for samples within the discovery cohort using the Bioconductor package, GenVisR [80,81]. The proportion of samples with LOH (dark blue), copy number gain (red), and copy number loss (blue) within each chromosomal region are depicted at the top of the panel.
Figure 3: NanoString validation of NR1H4 fusions observed in the discovery cohort
A. NanoString count values for fusions FNIP1-NR1H4, NR1H4-EWSR1, and NR1H4-GNPTAB across samples in the discovery cohort. Fusion diagrams are shown for: (B) NR1H4-GNPTAB, (C) NR1H4-EWSR1, and (D) FNIP1-NR1H4. E. The CDK17-NR1H4 fusion event was detected by Integrate and ChimeraScan but not validated with NanoString.
A

Overall Survival Probability

Time (Weeks)

p-value = 0.825

B

Recurrence Free Survival Probability

Time (Weeks)

p-value = 0.588

Mutation Status
- Mutant
- Wildtype
Figure 4: Overall and recurrence free survival analysis for TERT promoter mutations
Kaplan Meier curves for TERT mutation status (C228T/C250T promoter mutation, HBV promoter integration or TERT structural variant) for all non-cirrhotic samples. The probability of overall survival (OS) (A) and recurrence free survival (RFS) (B) across all non-cirrhotic samples is shown.
Figure S1: Evidence for a case of undiagnosed HBV integration at the TERT promoter

A. WGS reads originating from a competitive alignment between GRCh37 and HBV are shown for HCC18_D using the Integrative Genomics Viewer (IGV). B. The soft clipped sequence from panel A (red box) was inputted into BLAT to show that it aligns upstream of the TERT transcription start site (TSS). C. Reads which are soft clipped in panel A are shown aligning to the intronic region of human chromosome 5 in the HCC18_D tumor sample upstream of the TERT TSS.
Figure S2: HCC tumors exhibit shortened telomeres
A. *TERT* expression (log10(FPKM + 1)) within each matched tumor and normal sample from the discovery cohort, excluding HCC20 (top panel). B. Tumor:normal read ratios of telomeric regions are plotted as vertical bars in the bottom panel. The black horizontal line represents a read ratio of 1.
Armadillo/beta-catenin-like repeats

Atypical Arm repeats

Protein length

CTNB1

COSMIC

5418 mutations

31 tissue types

Protein Del

Protein FRAMESHIFT

Protein NONSENSE

Protein INTRON

Protein SPLICE_REGION

Protein SPLICE

Protein INTRON

Protein Somatic

Protein SILENT

Protein MISSENSE

Protein PROTEINDEL

Protein PROTEININS

22 mutations

Close

S33N

A239V

S47T

D32E

S45L

S33_S33del

A43P

A652_E12splice

A652_E13splice

T653R

R661Q

R661P

F683Y

W690C

D695Y

R710H

Y716C

G725G

P744T

D751N

M763I

L762P

D778N
Figure S3: GSK3B phosphorylation site mutational hotspot in CTNNB1 is observed in non-cirrhotic HCC
ProteinPaint [78] diagram illustrating amino acid (AA) changes within the coding sequence of CTNNB1. Legends describing the functional domains and variant types of the AA changes along CTNNB1 from our dataset and that of COSMIC's database are indicated at the bottom of the figure. The light-blue highlighted region indicates exon 3 of the CTNNB1 gene.
Figure S4: Discovery cohort samples are primarily composed of 4 mutational signatures
The weights from 0 to 1 of relative COSMIC mutational signature distribution are shown for each sample in the discovery cohort. Samples are hierarchically clustered using Euclidean distances based on assigned weights, signatures are ordered from left to right by decreasing weight.
Figure S5: Evidence for translocations involving TERT
Reads depicting translocation events involving TERT in tumor samples for (A) HCC23_D and (B) HCC29_D are shown. In both figures, the left panels contain dark blue reads that align to regions of chromosome 2 (HCC23_D) and chromosome 1 (HCC29_D). Their paired read aligns to a region ~2-3kbp upstream of the TERT promoter on chromosome 5, shown on the right.
Figure S6: Genomic landscape of the extension cohort
Manually reviewed coding, somatic mutations from the extension alpha and beta cohorts (N=87) derived from CAP1. In situations where there are multiple mutations for the same gene/sample, the most deleterious mutation is shown following the order of the mutation type legend. Clinical variables for each sample are included at the bottom. Tumor mutational frequency for each sample is plotted at the top.