Hepatocellular carcinoma (HCC) is a heterogeneous cancer with active Wnt signaling. Distinct dysregulation of Wnt-pathway constituents characterize two different Wnt-related classes: CTNNB1 class (138 cases [21.5%]) or Wnt-TGFβ class (177 cases [27.6%]). CTNNB1 class was characterized by dysregulation of liver-specific Wnt-targets and the absence of nuclear β-catenin. Sorafenib decreased Wnt signaling and β-catenin protein in HepG2 (CTNNB1 class), SNU387 (Wnt-TGFβ class), SNU398 (CTNNB1-mutation), and Huh7 (lithium-chloride-pathway activation) cell lines. In addition, sorafenib attenuated expression of liver-related Wnt-targets GLUL, LGR5, and TBX3. The suppressive effect on CTNNB1 class–specific Wnt-pathway activation was validated in vivo using HepG2 xenografts in nude mice, accompanied by decreased tumor volume and increased survival of treated animals.

Conclusions: Distinct dysregulation of Wnt-pathway constituents characterize two different Wnt-related molecular classes (CTNNB1 and Wnt-TGFβ), accounting for half of all HCC patients. Sorafenib modulates β-catenin/Wnt signaling in experimental models that harbor the CTNNB1 class signature.

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Introduction

Hepatocellular carcinoma (HCC) is the most common primary liver cancer and its incidence is increasing worldwide (1). As the third leading cause of cancer-related death and the most common cause of death among cirrhotic patients, it is emerging as a major global health problem (1). Hepatitis B and C viral infections and alcohol abuse are the main risk factors for HCC development. Curative treatments (e.g., resection, transplantation, or local ablation) are available only to patients with early-stage disease, but are
Translational Relevance

Hepatocellular carcinoma (HCC) is a major health problem globally with increasing incidence worldwide. Despite recent advancements in the understanding of its molecular pathogenesis and treatment, the knowledge of key molecular drivers and pathways remains ill-defined. Although Wnt signaling involvement in the pathogenesis of several malignancies is known, its specific role in HCC is unclear. Here, we dissect 2 major mechanisms of Wnt-activation in a large cohort of HCC samples reflecting different molecular subclasses of HCC. The Wnt-TGFβ subclass recapitulates classical Wnt signaling as described in other cancers and is linked to a more aggressive phenotype, whereas the CTNNB1 subclass is characterized by liver-specific Wnt-activation mediated by CTNNB1-mutations associated with a less aggressive phenotype. The fact that the multikinase inhibitor sorafenib was able to partially disrupt the activation of Wnt signaling in a xenograft model of liver cancer, opens the path to further explore specific mechanism of action of this molecule, and to explore its role in abrogating Wnt signaling in other cancers.

Limited by high recurrence rates, which impair patient outcomes. At advanced stages, the multikinase inhibitor sorafenib is the only effective treatment, although vigorous efforts are underway to better characterize the molecular pathogenesis of liver cancer to refine the efficacy of other molecular targeted therapies (2, 3).

Wnt-signaling is involved in multiple physiological processes, embryonic development, and cancer (4). The pathway is activated upon Wnt-ligand binding to frizzled receptors (Fzd) followed by cytosolic accumulation of β-catenin through prevention of glycogen synthases kinase three β (GSK3β) mediated phosphorylation of the β-catenin Ser/Thr domain. Cytosolic β-catenin can translocate to the nucleus to initiate transcription of target-genes through interaction with T-cell factor (TCF)/lymphoid enhancer factor (LEF) transcription factors (4). Aberrant activation of Wnt can also result from mutations in the β-catenin gene (CTNNB1), which is the second most frequent mutation observed in HCC after the p53 tumor suppressor (TP53; ref. 5).

Several genomic studies have identified molecular sub-classes of HCC using unsupervised clustering of mRNA or miRNA expression data. However, there is no universally accepted molecular classification for this disease. Nonetheless, Wnt signaling associated molecular classes have been reported by different investigators emphasizing the importance of Wnt-aberrations in liver cancer (5, 6). We recently reported a molecular classification of HCC, which includes a CTNNB1 class characterized by overexpression of liver-related Wnt-target-genes (which have been described in the liver, but are also found in other tissues, e.g., glutamine synthetase, GCLC, and leucine-rich repeat-containing G protein-coupled receptor 5, LGR5), enrichment in nuclear β-catenin staining and CTNNB1-mutations (6). This subclass strongly overlapped with Bouyault’s Wnt-related G5-6 classes, which were also characterized by mutations in CTNNB1 and significant overexpression of liver-related Wnt-target-genes (5, 6). The biologic similarity of HCC tumors within CTNNB1 class was further strengthened by a significant overlap with our recently published miRNA class A (7). In parallel, we characterized the molecular subclass S1 (Wnt-TGFβ) related to TGFβ-activated Wnt signaling through a meta-analysis of 603 HCC patients (8). With no association to CTNNB1-mutations or over-expression of liver-related Wnt-targets, this new molecular class was defined by activation of commonly reported Wnt-target-genes (e.g., CCND1 and MYC; ref. 8). Wnt signaling was also reportedly activated in a hepatic stem-cell-like HCC class showing a close correlation with EpCAM expression and induction of known Wnt-target-genes BAMBI and DKK1 (9, 10). Even though the dysregulation of Wnt signaling in hepatocellular cancer is described (3, 11), the simultaneous presence of multiple Wnt-related molecular profiles in cancer is a new concept that requires thorough characterization. It remains unclear to what extent different Wnt-targets, their transcriptional regulators and their downstream signals contribute to Wnt signaling diversity. In addition, efforts to develop antitumoral agents specifically targeting this cascade have not yet been developed.

Herein, we show that distinct differential expression of Wnt-pathway-members, upregulation of Wnt-target-genes, enrichment of CTNNB1-mutation-signature, and β-catenin and glutamine synthetase immunostaining characterize the 2 coexisting but distinct Wnt-related classes, CTNNB1 and Wnt-TGFβ, in HCC. We also provide evidence that sorafenib modulates Wnt/β-catenin signaling in experimental models of liver cancer recapitulating the human CTNNB1 class.

Materials and Methods

Human samples, mRNA, and miRNA expression arrays data

Gene expression profiling of a total of 642 human samples was analyzed for this study (Fig. 1). We used as training (91 HCV-related HCC; GSE8943 and GSE20594) and independent validation sets (144 mixed etiology HCC; GSE19977 without overlapping samples with the training set), samples previously profiled in the studies of our HCC Genomic Consortium (Supplementary Table S1). Normal liver tissue (n = 10) was obtained from patients with nonmalignant diseases. Written consents from patients, and institutional review board approvals were obtained from all institutions. We further analyzed gene expression of 407 HCC samples of patients with mixed etiology available from 5 public HCC data sets (Supplementary Table S2; ref. 8). Gene expression of 21 human liver cancer cell lines analyzed in this study is publicly available at http://www.broadinstitute.org/ccle.

Genomic profiling of Wnt-related mRNAs and miRNAs

Wnt-pathway mRNA and miRNA lists. To characterize Wnt-pathway alterations in the Wnt-related subclasses of HCC, we tested gene expression of 210 genes obtained from
publicly available Wnt-pathway lists (Supplementary Table S3: BIOCARTA_WNT_PATHWAY, KEGG_WNT_SIGNALING_PATHWAY, REACTOME_SIGNALING_BY_WNT, and WNT_SIGNALING), the Molecular Signatures Database and the Cancer Cell Map Pathway Map. In addition, all miRNAs reported to be involved in Wnt signaling in the literature were summarized in a separate list of 49 miRNAs associated with Wnt signaling (Supplementary Table S4).

Wnt-related subclasses of HCC. CTNNB1 class was assessed in the training set as previously reported (6) and determined by nearest template prediction (NTP) method (12) in 6 additional data sets (validation set and 5 publicly available independent gene expression data sets with sample size >50; ref. 8). Wnt-TGFβ class samples were identified as previously published (8) or NTP (validation set). Remaining samples not captured by either class were defined as non-Wnt.

Analysis of Wnt pathway alterations in Wnt-related classes. Differential mRNA and miRNA expression of Wnt-pathway members were analyzed by the GenePattern platform (www.broadinstitute.org/genepattern, Broad Institute, Boston, MA) using Comparative Marker Selection module (13), and false discovery rate (FDR) for multiple hypothesis testing correction (14).

We generated a gene signature to characterize each Wnt-related subclass by using the training set of samples. The CTNNB1-WntGenes-signature was obtained from genes of the Wnt-pathway-mRNAs significantly altered (FDR < 0.05) in CTNNB1 class samples compared with normal liver and non-Wnt HCC samples. The same approach was used to generate the Wnt-TGFβ-WntGenes-signature. Then these 2 signatures were validated by NTP using 6 additional data sets. CTNNB1- and Wnt-TGFβ class marker genes were excluded from the WntGenes-signatures before the validation analyses.

To characterize the miRNA profile differentially expressed in these 2 subclasses, we compared Wnt-pathway-miRNAs in each subclass versus non-Wnt samples in the training set. Thirty-three of 49 Wnt-pathway-miRNAs could be analyzed through the miRNA platform used (7) and we additionally excluded probes if >60% of the samples had low expression values (cut off <30, 2/33, 6.1%).

Association of CTNNB1 mutations to CTNNB1 class. The published strategy of a signature recapitulating TP53 mutations in breast cancer was used to generate a gene-signature of CTNNB1-mutation (15). The signature was generated with genes differentially expressed in HCC samples harboring or not CTNNB1-mutations (FDR < 0.05) in the training set. The ability of the signature to capture CTNNB1 mutated samples was validated in 1 HCC data set (5). NTP analysis was used to assess overlap between CTNNB1 class and the gene signature of CTNNB1 mutations in 6 additional data sets.

Immunohistochemistry

Immunohistochemistry (IHC) was carried out on 5 μm sections of formalin-fixed paraffin-embedded (FFPE) tissue by heat-induced epitope retrieval in 10 mmol/L sodium citrate (pH 6) before blocking with 5% BSA-PBS. Samples were incubated overnight at 4°C with anti-β-catenin and anti-glutamine synthetase (GS) antibodies at 1:1,000 dilutions (BD Biosciences). EnVision™+ System-HRP was applied as secondary antibody (Dako) and sections were counterstained with hematoxylin. Positive staining was defined according to percentage of stained cells (0 = no cells, 1 = 1%–4%, 2 = 5%–19%, 3 = 20%–39%, 4 = 40%–59%, 5 = 60%–79%, 6 = 80%–100% of positive cells) and intensity (0–3). Percentage scores were multiplied by intensity scores to yield an overall score. Nuclear and cytoplasmic β-catenin score of >3 and membranous β-catenin, and

Figure 1. Flowchart of study design. Flow chart displaying the different analyses carried out according to the data sets used. Boxes in gray represent information recently published by our group (6).
cytoplasmatic GS cytoplasmic score of >11 were considered as positive.

**Cell lines and drug treatments**

SNU398, SNU387, and HepG2 cells were purchased from ATCC (Manassa, VA). Huh7 (Riken Bioresource Center) and HepG2 cells were cultured in DMEM, SNU398, and SNU387 cells in RPMI medium supplemented with 10% FBS, 1% penicillin (100 U/mL) and 1% streptomycin (100 µg/mL). Cells were plated in 12- to 24-well plates or 10 cm² dishes and cultured overnight. Sorafenib (LC Laboratories) was added for 6 hours before cells were processed for chemiluminescence, Western blot, immunocytofluorescence, and TCF/LEF reporter assay. The Wnt-activator lithium-chloride (LiCl; ref. 16) was added to Huh7 cells 30 minutes after sorafenib treatment. Final DMSO concentration in all experiments was <0.05%. Gene expression of 21 cell lines was used to identify those that better characterize CTNNB1 and Wnt-TGFβ classes.

**TCF/LEF reporter assay**

Cells were transduced using Transduction-ready TCF/LEF reporter (luciferase) lentiviral particles (CLS-018L; SA Biosciences) following manufacturer’s instructions. Luminescence was assayed in cells stably expressing TCF/LEF luciferase reporter using a Luciferase Assay Kit (E1501; Promega) and a Veritas Microplate Luminometer (Promega). Luminescence/cell number (10,000–750,000 cells) line graphs were used to quantify luciferase activity with recombinant luciferase standard (E170A; Promega).

**Immunocytofluorescence and Western blot**

Standard methods were used as described in detail in Supplementary Methods.

**Real-time PCR**

Total mRNA was obtained from 24 hours treated cells and reversed transcribed using Sprint5™ Complete-Double Primed strips (ClonTech). TaqMan® Gene Expression Assays (Applied Biosystems) were used to measure mRNA levels of GLUL (assay ID: Hs00101305_g1), LGR5 (Hs00173664_m1), and TBX3 (Hs00255591_m1) according to manufacturer’s instructions. LightCycler480, and ΔΔCt method were used to quantify the data (normalized to glyceraldehyde-3-phosphate dehydrogenase (GAPDH Hs00266705_g1) after the methodology recently described (17).

**Animal experiments**

All animal studies were done upon Mount Sinai School of Medicine Institutional Animal Care and Use Committee approval. BALB/c female mice were injected with 5 × 10⁶ HepG2 cells subcutaneously and followed as described previously (18). Sorafenib (30 mg/kg/day) was administered by gavage using canola oil as vehicle. Animals were sacrificed and tumors collected when tumor volume reached 1 cm³ according to the following formula = length × width² × 0.4 (18).

**Statistical analyses**

Bars represent the mean ± standard error of the mean. Comparisons between groups were made using 2-tailed t- or U-test for continuous variables, and Fisher exact test for comparison of proportions. Correlations were calculated with the nonparametric Spearman’s coefficient. Gene set enrichment analysis was used to assess enrichment of which the CTNNB1- and Wnt-TGFβ class signature in the cell lines as previously described (8). All calculations were done with the SPSS package (SPSS 15.0).

**Results**

**Molecular portrait of two human Wnt HCC subclasses: CTNNB1 and Wnt-TGFβ**

**Differential expression of Wnt genes in CTNNB1 and Wnt-TGFβ classes.**

To identify Wnt-pathway differential alterations in CTNNB1- and Wnt-TGFβ classes, we analyzed the gene expression of 210 Wnt-pathway-related genes in comparison with normal liver and HCCs not related to Wnt classes in the training set. Tumors of the CTNNB1 class (24/91, 26.4%) and Wnt-TGFβ class (23/91, 25.3%) showed significant dysregulation of 36 and 48 Wnt-pathway-mRNAs, respectively (FDR < 0.05; Supplementary Figs. S1 and Table S5). Overall, both classes had only 4 significantly dysregulated genes in common (SALL1, PRKCD, PLAU, and MAP1B) but with expression in opposite directions. Therefore, each molecular class was characterized by its own specific Wnt-related mRNA expression pattern. Supplementary Table S6 provides a full list of genes, subcellular localization and effect on Wnt signaling (activation/inhibition; ref. 19).

A higher number of activating alterations are observed in Wnt-TGFβ class samples, among which 5 of them (FZD7, FZD6, TCF4, FRAT2, and ARBB2) have been related to Wnt activation in cancer (refs. 20–25; Fig. 2A). CTNNB1 class was characterized by CDH1 downregulation, and a significant number of potentially inhibiting alterations (Supplementary Fig. S1). Several WNTs and FZD were significantly downregulated in both classes inhibiting Wnt signaling (Supplementary Tables S5 and S6).

In CTNNB1 class, 7 of 9 reported liver-related Wnt-target genes (GLUL, REG3A, LGR5, MERTK, TBX3, EPBB2, and SPARCL1; ref. 26) were significantly upregulated (Fig. 2B, Supplementary Table S7) and showed a significant correlation with the expression of several transcription factors dysregulated in that class (SALL1, TLE1, CTNNB1P1, TCF7, TCF7L1, SENP2, FSTL1, SMAD3, RUVBL1, and CTBP2; Supplementary Tables S5 and S8). In contrast, the classical Wnt-targets MMP7, PLAU, RUNX2, CDC2, and CCND3 (Fig. 2B, Supplementary Table S3) were only significantly upregulated in Wnt-TGFβ class and showed a significant correlation to most altered transcription factors (HDAC1, CDX1, TCF4, SALL1, Sox1, TCF1, and TAK1; Supplementary Tables S5 and S8).

In summary, clear differential expression of Wnt-pathway genes contributes to the molecular diversity of our 2 previously reported Wnt-related molecular classes (Supplementary Fig. S1). Interestingly, CTNNB1- and Wnt-TGFβ
Figure 2. Expression alterations of Wnt pathway genes and miRNAs in 2 Wnt subclasses of HCC, CTNNB1 (blue) and Wnt-TGFβ (green), in the training set. A, distribution of CTNNB1 mutations (6), β-catenin IHC nuclear and cytosolic staining (6), and G6/GS-6 molecular profile (5). Heatmap shows Wnt pathway genes differentially expressed (arrows) between CTNNB1 (blue) or Wnt-TGFβ (green) subclasses. Only genes potentially related to Wnt pathway activation are shown (see Supplementary Fig. S1 for full list of genes and non-Wnt samples). Highlighted gene names correspond to those that have been related to induction of Wnt pathway activation in cancer (20–25). B, full display of mRNA-based molecular classifications showing CTNNB1 subclass (7) and Wnt-TGFβ subclass (9) in the training set. Heatmap shows liver-related Wnt target genes (5) and classical Wnt target genes (list obtained from public databases, see Materials and Methods) differentially expressed between either Wnt class compared with the rest of HCC samples (B) and are summarized in (C) as classical Wnt target genes.
class express liver-related and classical Wnt-target-genes, respectively, which are significantly correlated to the expression of class-specific transcription factors.

MiRNA array analysis also reflected key differences between both classes (Supplementary Fig. S2). Although 19 Wnt-related-miRNAs were significantly dysregulated in CTNNB1 class, only 4 miRNAs were differentially expressed in the Wnt-TGFβ class. Dysregulated miRNAs were either subclass exclusive or dysregulated in opposed directions in CTNNB1 class compared with Wnt-TGFβ class. (Supplementary Table S10).

Validation of CTNNB1 and Wnt-TGFβ class-specific expression profiles in six independent HCC data sets. To validate the significant differential expression of Wnt-pathway-miRNAs, we generated 2 gene sets, named CTNNB1- and Wnt-TGFβ-WntGenes-signature, comprising the Wnt-related genes dysregulated in each class (Supplementary Fig. S1B). Instead of validating the dysregulation for each gene, we evaluated if the global dysregulation defined by the 2 signatures overlapped with their corresponding molecular classes. We tested the CTNNB1- and Wnt-TGFβ-WntGenes-signatures in an independent cohort of 144 samples from our Consortium, and in 5 HCC data sets reported from other groups.

In our validation set, 76/144 (52.8%) samples had active Wnt signaling, with 35 (24.3%) belonging to the CTNNB1 class and 41 (28.5%) to the Wnt-TGFβ class. Their corresponding gene-signatures of dysregulated Wnt-pathway-miRNAs were associated with the CTNNB1- and Wnt-TGFβ classes (P < 0.001 and P = 0.005, respectively; Fig. 3). Next, the expression of liver-related Wnt-target-genes was analyzed by comparative marker selection, which remained significantly upregulated in 7 (GLUL, REG3A, LGR5, MERTK, TBX3, EPHB2, and SPARCL1) of 9 genes in CTNNB1 class (Supplementary Table S7). Since no mutation data were available for samples of the validation class, we generated a CTNNB1-mutation-signature with the mutated samples of the training set to predict the mutation status. Samples identified were significantly enriched in CTNNB1 class (P < 0.001; ref. Fig. 3).

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**Figure 3.** Validation of 2 distinct Wnt classes and their specific Wnt-related mRNA expression profiles in an independent set of HCCs. A, prediction of Chiang’s (6) and Hoshida’s (8) CTNNB1- and Wnt-TGFβ class in our validation set using NTP (FDR < 0.05). CTNNB1 class was significantly enriched for combined nuclear-cytoplasmic β-catenin (P < 0.001) and cytoplasmic glutamine-synthetase (P < 0.005) immunohistochemistry (IHC). Samples identified by CTNNB1-mutation-signature, and CTNNB1-WntGenes-signature significantly correlated with CTNNB1 class (P < 0.001), whereas samples identified by Wnt-TGFβ-WntGenes-signature significantly overlapped with Wnt-TGFβ class samples (P = 0.005). B-D, β-catenin and glutamine synthetase immunohistochemistry: percentage of positive samples in Chiang’s subclasses in the validation set and representative positive immunostaining for (A) nuclear and cytoplasmic β-catenin, (B) membrane β-catenin, and (C) glutamine synthetase.

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In an additional approach, we interrogated 5 publicly available HCC genomic data sets representing a total of 407 patients to assess the relevance of Wnt-activation in HCC and to further confirm our class-specific Wnt-genes expression profiles. Among these 407 HCC patients, 194 (47.7%) harbored activation of the Wnt-pathway, 81 (19.9%) in CTNNB1 class, and 113 (27.8%) for Wnt-TGFβ class. In 4/5 and 5/5 HCC data sets, samples identified by the CTNNB1- and the Wnt-TGFβ-GeneSet-signatures, respectively, significantly correlated with the molecular classes CTNNB1 and Wnt-TGFβ, respectively (Supplementary Fig. S3 and Table S11). In all data sets analyzed, CTNNB1 class was also significantly enriched in samples called by the CTNNB1-mutation-signature (Supplementary Fig. S3 and Table S11). Specifically, CTNNB1-mutation-signature was validated in 1 data set where CTNNB1 mutations were reported (5). Overall, the prediction accuracy of our signature was of 91% (specificity = 87%, sensitivity = 100%; Supplementary Fig. S4).

Overall 315/642 (49.1%) HCC samples analyzed in this study showed active Wnt signaling belonging either to the CTNNB1 class (21.5%) or the Wnt-TGFβ class (27.6%). Particularly, Wnt-TGFβ class was enriched in patients with non-HCV etiology (35%) compared with those with HCV-related HCC (17%). Thus, the existence of 2 distinct Wnt-pathway mRNA expression patterns for CTNNB1- and Wnt-TGFβ classes could be confirmed in a total of 5 and 6 independent HCC data sets, respectively.

Nuclear β-catenin and glutamine synthetase protein in the CTNNB1 class. To better characterize Wnt-activation at the protein level, immunostaining for β-catenin and glutamine synthetase as liver Wnt-target was assessed in our validation set (Fig. 3). Strong nuclear and cytoplasmic β-catenin staining was observed in 28/35 (80%) CTNNB1 class samples compared with 18/108 (16.6%) of non-Wnt HCC samples (P < 0.001), although membrane staining was significantly decreased in 9/35 (25.7%) in CTNNB1 class compared with 70/108 (64.8%) of the remaining HCCs (P < 0.001; Fig. 3). β-Catenin staining distribution was similar in the training and validation cohorts (Supplementary Table S9). Similarly, glutamine synthetase staining was enriched in 26/35 (74.3%) CTNNB1 class samples compared with 23/108 (21.3%) non-Wnt HCCs (P < 0.001). Although both Wnt-related molecular classes show clear alteration of the Wnt-pathway at the mRNA and miRNA level, only the CTNNB1 class displayed increased cytoplasmic and nuclear β-catenin and glutamine synthetase by immunostaining.

CTNNB1- and Wnt-TGFβ class signatures in liver cancer cell lines. We next sought to identify liver cancer cell lines with a gene expression profile similar to either CTNNB1- or Wnt-TGFβ class. For that purpose, we obtained the gene expression data of 21 liver cancer cell lines (Supplementary Table S12 and Fig. S3; refs. 6 and 8). Analyzed by Gene Set Enrichment Analysis, HepG2 cells were significantly enriched in the CTNNB1 class-signature, although SNU387 cells were significantly enriched in the Wnt-TGFβ class-signature (P < 0.005). Therefore, HepG2 cells were used as the cell line resembling the CTNNB1 class, SNU387 cells as a model for the Wnt-TGFβ class, and SNU398 cells as a model harboring CTNNB1-mutation. Huh7 showed no clear similarity to either molecular class and was used to experimentally activate Wnt signaling.

Sorafenib alters Wnt signaling in vitro and in vivo

Sorafenib decreases TCF/LEF luciferase reporter activity and β-catenin protein levels in human liver cancer cell lines. Because of clear dysregulation of Wnt signaling in human HCC, we sought to investigate whether sorafenib can block this cascade in experimental models of liver cancer. HepG2, SNU398, SNU387, and Huh7 cells were stably transduced with luciferase reporter of TCF/LEF, β-catenin/Wnt signaling, and according to baseline TCF/LEF luciferase reporter expression, cells could be ordered as HepG2 > SNU398 > Huh7 > SNU387 (Supplementary Fig. S6). Baseline expression of luciferase in HepG2 was >1,000 times greater than in Huh7. These levels are consistent with the mutation status of Wnt components in these cells. HepG2 express a truncated β-catenin lacking the Ser/Thr domain that regulates its degradation, whereas SNU398 have a point mutation in a key Ser in the same domain (S37C). Huh7 and SNU398 have no known mutations in β-catenin or other Wnt-related gens (27, 28). Sorafenib was effective in modulating transcriptional activity through TCF site, and in decreasing β-catenin protein levels in cells with truncated/mutated β-catenin (HepG2, SNU398), in cells with wild-type β-catenin (SNU387) and in cells with wild-type β-catenin and induction of Wnt signaling by GSK3β-inhibitor LiCl (Huh7; Fig. 4A and Supplementary Fig. S7; ref. 16).

Immunocytofluorescence further showed that treatment with sorafenib prevents β-catenin translocation to the nucleus in LiCl-treated Huh7 cells, and decreases total β-catenin levels in HepG2 cells (Fig. 4B and C). Taken together, these data suggest that sorafenib can decrease Wnt/β-catenin signaling by modulating transcriptional activity through the TCF/LEF site and reduces β-catenin protein levels in liver cancer cells.

Sorafenib reduces liver-related Wnt target genes in experimental models of CTNNB1 class. As previously stated, HepG2 cells recapitulate the CTNNB1 class-signature in culture. mRNA levels of 3 known liver-related Wnt-target-genes GLUL, LGR5, and TBX3 were significantly decreased after sorafenib treatment in vitro (Fig. 5A). Thus, we tested the effect of sorafenib in a subcutaneous xenograft mouse model using HepG2 cells. In this xenograft model, sorafenib treatment (30 mg/kg/day) reduced tumor volume (Fig. 5C) and extended median survival from 22 days in control animals (n = 7) to 34 days in treated mice (Fig. 5D, n = 7). After sorafenib treatment, GLUL and LGR5 were significantly downregulated in HCC murine xenografts, and TBX3 showed a trend of decreased expression (Fig. 5B). Therefore, sorafenib’s effect on modulating aberrant β-catenin/Wnt signaling is also observed in a xenograft model resembling the CTNNB1-molecular class.
Discussion

This study reports distinct Wnt-pathway alterations at the transcriptome, miRNA, and protein expression level that characterize 2 distinct Wnt-associated molecular classes, CTNNB1 and Wnt-TGFβ, in human HCC. Markers and known drivers of Wnt activation are observed in both...
subclasses (Fig. 6). In addition, we show that sorafenib can modulate β-catenin/Wnt signaling in liver cancer cells, and in CTNNB1 class-like HepG2 xenograft tumors in mice.

Although several molecular classes have been associated with Wnt signaling in HCC (5, 6, 8) and the concept of liver-related and classical Wnt signaling has been proposed (29), no detailed characterization of underlying mechanisms has been pursued yet. Our group recently reported 2 fundamentally different molecular classes associated to Wnt signaling in HCC based on expression of liver-related
class) Wnt-target-genes (6, 8). Both have been identified as distinct classes in several independent HCC data sets (8), underscoring their biologic divergence. Although the CTNNB1 class was correlated with CTNNB1 mutations, nuclear β-catenin staining, and tumor diameter >3 cm (6), Wnt-TGFB class was associated with TGFB activation, cytoplasmic β-catenin staining, vascular invasion, satellitosis, and greater risk of early recurrence after surgical resection (8).

The expression analysis of 210 Wnt-pathway-genes in 642 HCC samples strengthens the concept of 2 mutually exclusive Wnt-related molecular classes in HCC by revealing class-specific Wnt-related expression profiles in these classes. Several Wnt-pathway proteins show class-specific dysregulation at the membrane, cytoplasm, and nucleus, reinforcing the presence of 2 different, active Wnt-pathways in liver cancer. Besides CTNNB1 mutations, we identify CDH1 downregulation as the key contribution to Wnt signaling activation in CTNNB1 class samples. In contrast, several genes dysregulated in Wnt-TGFB class characterize Wnt signaling activation in cancer (refs. 20–25; Fig. 6). This is the case of both FZD7 (21) and TCF4 (22), which upregulation has been previously described to activate Wnt signaling in the absence of β-catenin mutations. In addition, the expression of Wnt-related transcription factors seems to be class-specific and significantly correlates to liver-related (CTNNB1 class) or classical (Wnt-TGFB class) Wnt-target-genes. Regarding TCF4 overexpression, it was previously related to classical Wnt-target-genes expression (e.g., CCNDs or MMRs), even though isoform-specific functions remain unclear and would require further analysis (30). Interestingly, common Wnt pathway target genes MYC and CCND1 were not found to be associated to Wnt activation in the liver, as described elsewhere (5, 31).

In accordance with our previously reported miRNA-based HCC classification that identified miRNA class A significantly enriched in the CTNNB1 class (7), we found significantly more Wnt-pathway-miRNAs dysregulated in the CTNNB1 class than in the Wnt-TGFB class. Although several studies have implicated these Wnt-related miRNAs in Wnt signaling, including the association of miR-375 expression to CTNNB1-mutation (32), no Wnt class-specific miRNA expression profiles or Wnt-related miRNA-signatures have been described so far. Whether aberrant expression of miRNAs characterizing CTNNB1 class-related Wnt signaling has functional relevance needs to be further elucidated. Because a TP53-mutation signature reportedly predicts mutation in breast cancer accurately (15), we aimed to estimate CTNNB1-mutations with a mutation-signature generated in our training set. Its prediction accuracy was 91% when tested in an independent data set (5) and it was enriched in CTNNB1 class in 6 independent HCC data sets, underscoring the idea that CTNNB1-Wnt is activated and driven by CTNNB1-mutations in HCC.

At the protein level we confirmed Wnt-pathway activation in CTNNB1 class by positive nuclear and cytoplasmic β-catenin immunostaining, although in Wnt-TGFB class β-catenin staining remained membranous. Glutamine synthetase staining as a surrogate of active β-catenin signaling was only enriched in the CTNNB1 class, validating the existence of liver-related Wnt-target-genes at the protein level. Although the importance of Wnt signaling for the Wnt-TGFB class has been proven experimentally (8), it remains unclear why pathway activation at the protein level is not detected in our set of samples. Although some studies reported pathway activation without detection of nuclear β-catenin, others described negative feedback loops that activate pathway repressors or suppress pathway activators in Wnt signaling (33). However, it has recently been reported that positive β-catenin and glutamine synthetase staining are good predictors of CTNNB1-mutation (34), further strengthening the idea that CTNNB1 class is primarily driven by CTNNB1-mutation.

The fact that half of the HCC samples displayed activation of Wnt signaling points to the importance of this signaling cascade in hepatocarcinogenesis and reinforces the need for drugs targeting this pathway. However, all efforts to develop and translate effective and safe Wnt-pathway modulators into the clinical setting have been complicated by the multifaceted nature of Wnt signaling (4, 15). Wnt-inhibitor toxicity has been reported for several cancers and might be intolerable in cirrhotic patients with limited hepatocyte regenerative capacity. The only FDA-approved drug for advanced HCC, sorafenib, is well tolerated in cirrhotic patients (2) and inhibits several tyrosine kinases (e.g., VEGF, PDGFR, ERK/AKT, Ras, etc.), and their downstream oncogenic pathways (35). Although the Wnt-pathway has not been described as a direct target of sorafenib, there is increasing evidence of potential crosstalk between, for example, Wnt/β-catenin and PI3K/AKT signaling pathways (36, 37). The current data reveal a potential off target effect of sorafenib, and show its ability to reduce β-catenin/Wnt signaling in several experimental models of liver cancer. For HepG2 cells, which resemble the CTNNB1 class in their expression profile, sorafenib-mediated Wnt modulation was observed both in culture and in vivo. It remains unclear whether Wnt modulation contributes to sorafenib’s antitumor benefits in patients with HCC, and in particular those with Wnt-pathway aberrations. Therefore, profiling of HCC patients treated with sorafenib would shed light on potential correlations between Wnt-related molecular classes and treatment response. Certainly, the mechanism by which sorafenib modulates Wnt signaling needs to be characterized in greater detail to determine whether pathway crosstalk or a direct interaction with Wnt-pathway members leads to the observed effect on β-catenin and Wnt-target genes.

Disclosure of Potential Conflicts of Interest

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