

Figure 1. Immunoblot profiling of liver and tumor samples. *A*, expression of Myc, β -catenin, and YAP in livers and HCCs. Tissues examined included control, D3 and D7 livers, initial tumors, R3 and R7 regressing tumors, and recurrent tumors. The adjacent Myc blot is a repeat of the one shown to the left with both tumor samples removed to allow a longer exposure and reveal the lower but still overexpressed levels of Myc in D3 and D7 liver samples. See [supplemental Fig. S2A](#) for additional immunoblots and [supplemental Table S6](#) for band density quantification. *B*, phosphorylation-dependent regulation of PDH by the inhibitory kinase Pdk1 and the stimulatory phosphatase Pdp2 (27). High ATP/ADP ratios concurrently stimulate Pdk1 and inhibit Pdp2, thereby down-regulating PDH activity and limiting pyruvate's conversion to AcCoA (27). *C*, immunoblots showing total PDH and phospho-Ser²⁹³ PDH E1 α subunit (pPDH), an indicator of relative PDH inactivity. Also examined were total amounts of PDK1 and PDP2. See [supplemental Fig. S2B](#) for additional immunoblots and [supplemental Table S6](#) for band density quantification. *D*, PDH enzymatic activity in fresh tissue samples. Groups were compared with ANOVA, and each group was compared with control liver with two-sided *t* tests with Welch's correction due to heteroskedasticity and Bonferroni correction to control for false discovery (adjusted 0.0083). $n = 4-5$ samples/group except control liver, with 29 samples. The release of ¹⁴C₂ during the conversion of ¹⁴C-labeled pyruvate to AcCoA was monitored as described previously (30). *E*, ATP levels were quantified with 3–4 samples per group, each assayed in triplicate, and were compared with ANOVA and Tukey's multiple comparisons two-sided *t* tests.

tumorigenesis. It also allows a determination of how these factors respond during tumor regression following Myc silencing and tumor recurrence following Myc re-expression. This approach therefore provides insight into the evolutionary processes that accompany primary and relapsed tumorigenesis in response to the precisely regulated overexpression of a single well defined oncogenic stimulus.

Results

HCC evolution, regression, and recurrence

In contrast to a previous report (22), liver weights remained unchanged for at least 1 week following Myc induction. Tumors also appeared earlier than previously reported, reaching maximal size within 3–4 weeks. Tumors then regressed by ~50–75% within a week of resuming doxycycline and disappeared over the ensuing 1–2 weeks ([supplemental Fig. S1A](#)). Upon re-inducing Myc 2–3 months later, tumors recurred more rapidly than initial tumors, attaining sizes approximating those of the original neoplasms within 11–15 days. We refer to these hereafter as “initial tumors” (or “tumors”) and “recurrent tumors,” respectively. Cell cycle analyses performed on nuclei isolated from initial and recurrent tumors showed distinctly different patterns compared with livers and with one another ([supplemental Fig. S1B](#)). For example, nuclei isolated from normal liv-

ers, although primarily diploid, also contained the expected subpopulations of tetraploid and octaploid nuclei (23). The paucity of S-phase populations was consistent with these cells being uniformly quiescent. In contrast, nuclei from initial tumors were diploid (or pseudo-diploid) with a significant S-phase population. Recurrent tumors showed a greater fraction of S-phase cells and a more disperse DNA distribution, suggesting that many of the nuclei were aneuploid, despite the known clonal derivation of these from the initial tumors (22). Both initial and recurrent tumors expressed high but otherwise equivalent levels of proliferating cell nuclear antigen (Pcna). Thus, the faster growth of recurrent tumors correlated with a higher S-phase population and aneuploidy ([supplemental Fig. S1C](#)). Histologically, all tumors resembled moderately differentiated HCCs with numerous mitotic cells observed in most high-power fields ([supplemental Fig. S1D](#)) (22).

Both initial and recurrent tumors expressed high Myc levels (Fig. 1A and [supplemental Fig. S2A](#)). Much lower levels were detected in livers prior to the appearance of tumors (D3 and D7), and virtually none was detected in regressing tumors (R3 and R7). Thus, in response to doxycycline removal, and as reported previously (24), Myc accumulated gradually, achieved maximal expression only in fully developed tumors, and disappeared rapidly following return to the “Myc off” state.

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Many HCCs co-express β -catenin, YAP, and Myc, and the *MYC* gene is a target for both the β -catenin and Hippo/YAP pathways (14, 15, 17–19, 25, 26). Whether Myc in turn influences the expression of the other two oncoproteins is unknown as is the order in which these proteins appear during disease progression. β -Catenin expression was detected at low levels in liver, accumulated gradually following Myc induction, and was maximally expressed in initial tumors or R3 tumors. It then tended to decline by R7 and did not reappear in recurrent tumors (Fig. 1 and supplemental Fig. S2A). In contrast, YAP was highly up-regulated in and remained confined to initial and recurrent tumors, although expression in the latter group was lower. This suggested that, at least in initial tumors, the deregulation of YAP and β -catenin was more dependent on sustained Myc expression than on the actual tumor environment. It further indicated distinct biochemical differences between initial and recurrent tumors with regard to the expression of these two critical oncoproteins.

Metabolic changes accompanying HCC evolution

Pyruvate dehydrogenase (PDH) is a mitochondrial enzyme complex that links glycolysis and the TCA cycle by catalyzing pyruvate's irreversible conversion to acetyl coenzyme A (AcCoA) ($\Delta G^0 = -7.5$ kcal) (27). Reduced PDH activity in some cancers may allow the accumulation of upstream glycolytic intermediates that serve as anabolic precursors while simultaneously diverting pyruvate away from the TCA cycle in cells that are heavily reliant on the Warburg effect (28, 29). In experimental murine hepatoblastomas (HBs), however, we previously showed PDH to be highly up-regulated and to correlate with decreased expression of the inhibitory kinase Pdk1 and increased expression of the stimulatory phosphatase Pdp2 (Fig. 1B) (27, 30). To determine whether this was also true in HCCs, we measured total and phosphorylated (Ser²⁹³) PDH E1 subunit (pPDH), with the latter indicating a less active form of the enzyme (27, 29). In keeping with our earlier HB results (30), both initial and recurrent HCCs showed significantly reduced E1 subunit phosphorylation as well as decreased Pdk1 and increased Pdp2, with changes in the latter being seen as early as D3–D7 (Fig. 1C and supplemental Fig. S2B). Consistent with these findings, PDH enzymatic activity was also markedly increased at these times (Fig. 1D). Despite similarly appearing pPDH/Pdk1/Pdp2 profiles (Fig. 1C and supplemental Fig. S2B), recurrent tumors had lower PDH activity than even control livers (Fig. 1D), perhaps because of an overall lower level of total PDH. In keeping with the notion that high PDH activity might maximize ATP generation in the face of limiting amounts of pyruvate, initial tumors but not recurrent tumors contained significantly more ATP than livers (Fig. 1E). Indeed, recurrent tumors contained even less PDH activity than livers. Collectively, these results showed that PDH activity was highly sensitive to initial fluctuations in Myc levels but that this response was lost in recurrent tumors in a manner that correlated with ATP levels.

To identify additional metabolic changes, we next evaluated certain mitochondrial properties and functions, including mass, oxidative phosphorylation (Oxphos), electron transport chain (ETC) function, and fatty acid β -oxidation (FAO), that

are deregulated in HBs (30, 31). These and other human tumors often have significantly reduced mitochondrial mass relative to their corresponding normal tissues, thus partially, if not entirely, explaining the Warburg effect (30, 32). Similarly, an ~ 30 –40% reduction in mitochondrial mass was noted in initial tumors and persisted during regression and recurrence (Fig. 2A).

Complex I activity of livers remained unchanged in the immediate D3–D7 period but increased significantly in initial tumors as described previously for HBs (Fig. 2B) (30). This persisted in R3 tumors, normalized in R7 tumors, and remained low in recurrent tumors. Complex II activity was also initially unchanged in response to Myc activation but then decreased by $\sim 50\%$ in initial tumors as also seen previously in HBs (30). Rapid normalization of complex II activity occurred by R3 and persisted, even in recurrent tumors. These results indicate that, relative to livers, initial tumors altered their Oxphos profile so as to favor complex I over complex II as a source of electrons for complex III. In contrast, the overall mitochondrial function in recurrent tumors more closely resembled that of normal livers. Together, these results indicate that, despite the similarity of their mitochondrial mass, initial and recurrent tumors are metabolically distinct with regard to ETC function.

Blue native gel electrophoresis (BNGE) and *in situ* enzymatic assays (30, 31, 33) revealed no significant differences in the amounts or activities of ETC complexes I, III, and IV or of complex V (ATPase) over the course of HCC evolution (Fig. 2C and supplemental Fig. S3). In contrast, complex II activity was increased nearly 4-fold in initial tumors and somewhat less so in recurrent tumors. These results, together with the lower complex II activity of initial tumors as measured by respirometry (Fig. 2B), indicated that the electron transport function of complex II, as measured by OCR (Fig. 2B), was uncoupled from its succinate dehydrogenase (SDH) activity, which oxidizes succinate to fumarate using flavin adenine dinucleotide (FAD) as the electron donor (34).

The profoundly reduced FAO activity in HBs has suggested that, rather than being metabolized to AcCoA and re-synthesized into lipids, *de novo*, free fatty acids are instead preferentially incorporated directly into new lipids to support rapid tumor growth, thus presumably increasing tumor reliance on glucose as an energy source (30, 31). Similarly, we observed a gradual decline in FAO in response to Myc activation, culminating in a nearly 7-fold reduction in initial HCCs (Fig. 2D). FAO activity in recurrent tumors was also low, suggesting that this pathway does not contribute significantly to the ATP supply in either case.

Coordinated transcriptional dynamics accompany tumor evolution, regression, and recurrence

We employed RNA-seq to profile the transcriptional landscapes of the above seven sets of tissues. Because of the large number of differentially expressed genes that were identified (Fig. 3A), we used Ingenuity Pathway Analysis (IPA) to categorize them into 13 “super pathways” that were further composed of multiple, functionally related “member pathways” (Fig. 3B and <https://prochownik.pitt.edu/hcc>). In keeping with the ability of de-regulated Myc to rapidly induce reactive oxygen spe-

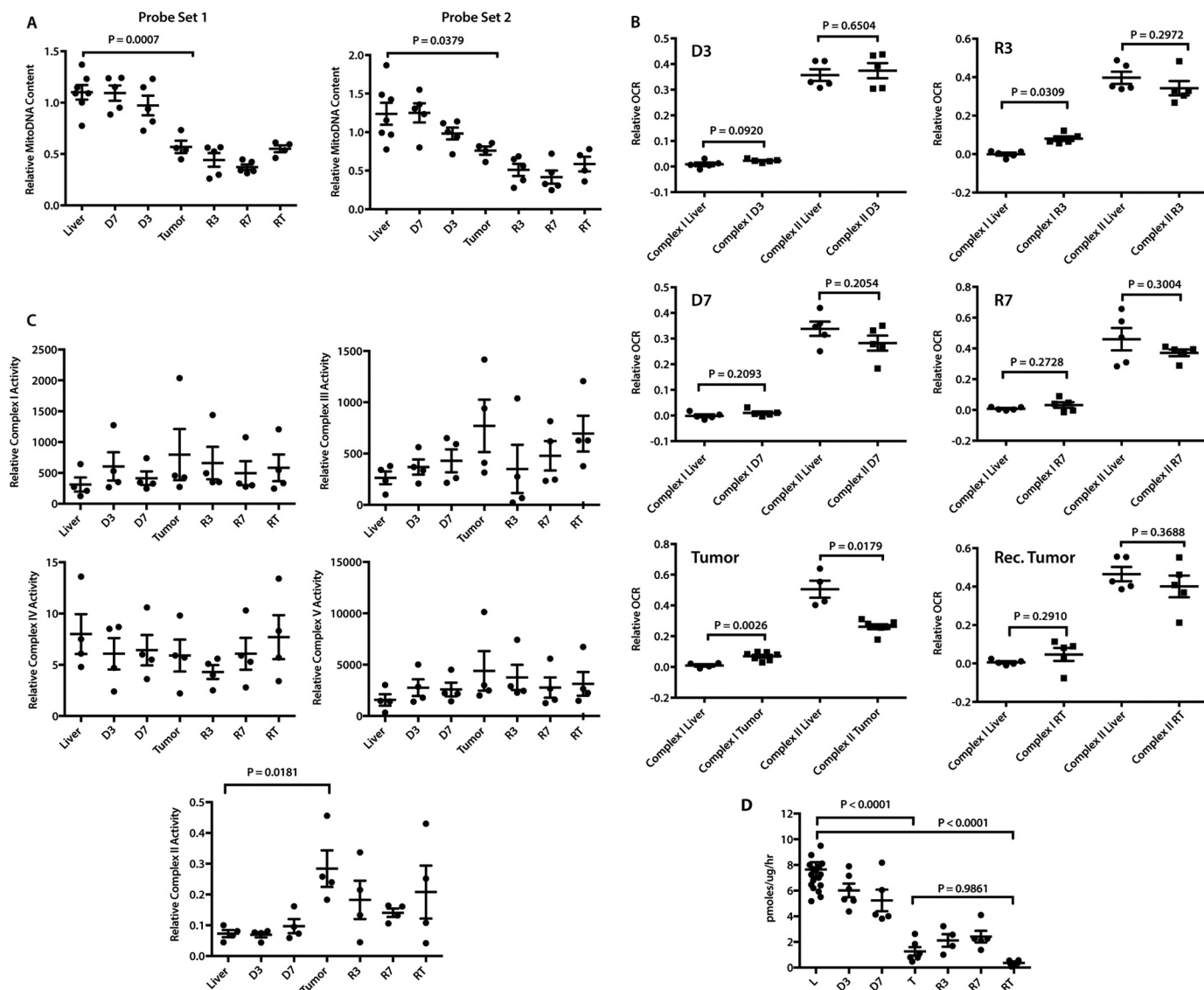


Figure 2. Changes in mitochondrial structure and function during HCC evolution. A, mitochondrial mass. Mitochondrial DNA (mtDNA) content was assessed with two TaqMan probe sets (probe set 1 and probe set 2) designed to amplify distinct regions of the mitochondrial genome. Results were normalized to those of a parallel PCR that amplified a segment of the apolipoprotein B gene (30) and compared with ANOVA and Tukey's multiple comparisons two-sided *t* tests. $n = 4-7$ samples/group. B, activities of ETC complexes I and II in intact mitochondria. OCRs were quantified on liver and tumor mitochondrial suspensions by respirometry (30, 31) and compared with Mann-Whitney U tests. Combined complex I + II activities were evaluated by measuring peak OCRs in response to pyruvate, malate, glutamate, and ADP followed by the addition of succinate. After adding rotenone to inhibit complex I, the individual contributions of complex I and complex II were calculated and normalized to protein content. $n = 5-7$ samples/group. C, ETC complex function. BNGE was performed on liver and tumor samples from each of the indicated groups (30, 31) and compared with ANOVA and Dunnett multiple comparisons two-sided *t* tests. The relative abundance of each complex was determined by densitometric scanning (supplemental Fig. S3A) and used to normalize *in situ* enzymatic activities. Complexes I-V measured NADH ubiquinone oxidoreductase, succinate dehydrogenase, decylubiquinol cytochrome *c* oxidoreductase, cytochrome *c* oxidase, and ATPase, respectively (31). SDH activity of complex II was measured on total tissue lysates rather than by *in situ* assay and was log-transformed for analysis to satisfy ANOVA's homoskedasticity requirement. $n = 4$ samples/group. D, reduced FAO oxidation by initial and recurrent tumors. [^3H]Palmitate conversion to water-soluble products was measured from the indicated groups of mice and compared with ANOVA and Tukey's multiple comparisons two-sided *t* tests. $n = 4-5$ samples/group except in control liver, which had 18 samples.

cies (ROS), unscheduled DNA synthesis, and structural chromosomal and genomic damage *in vitro* (35-38), D3 livers showed extensive but selective deregulation of three super pathways ("cancer signaling/proliferation," "cell cycle," and "DNA damage"), composed of 16 individual member pathways devoted almost exclusively to cell cycle and mitotic checkpoint control, growth arrest, and the response to and repair of DNA damage (*i.e.* ATM and Brca1 signaling and base excision, mismatch, and double-stranded DNA repair) (Fig. 3B).

Together, these member pathways contained 156 deregulated transcripts. Consistent with Myc's ability to rapidly induce aberrant DNA synthesis (37, 38), transcripts for virtually all major cyclins, cyclin-dependent kinases and cyclin-dependent kinase inhibitors, were significantly induced at this time (supplemental Fig. S4). A fourth de-regulated super pathway pertained to "small molecule biochemistry" within which only a single member pathway related to pyrimidine deoxyribonucleotide biosynthesis and involving four tran-

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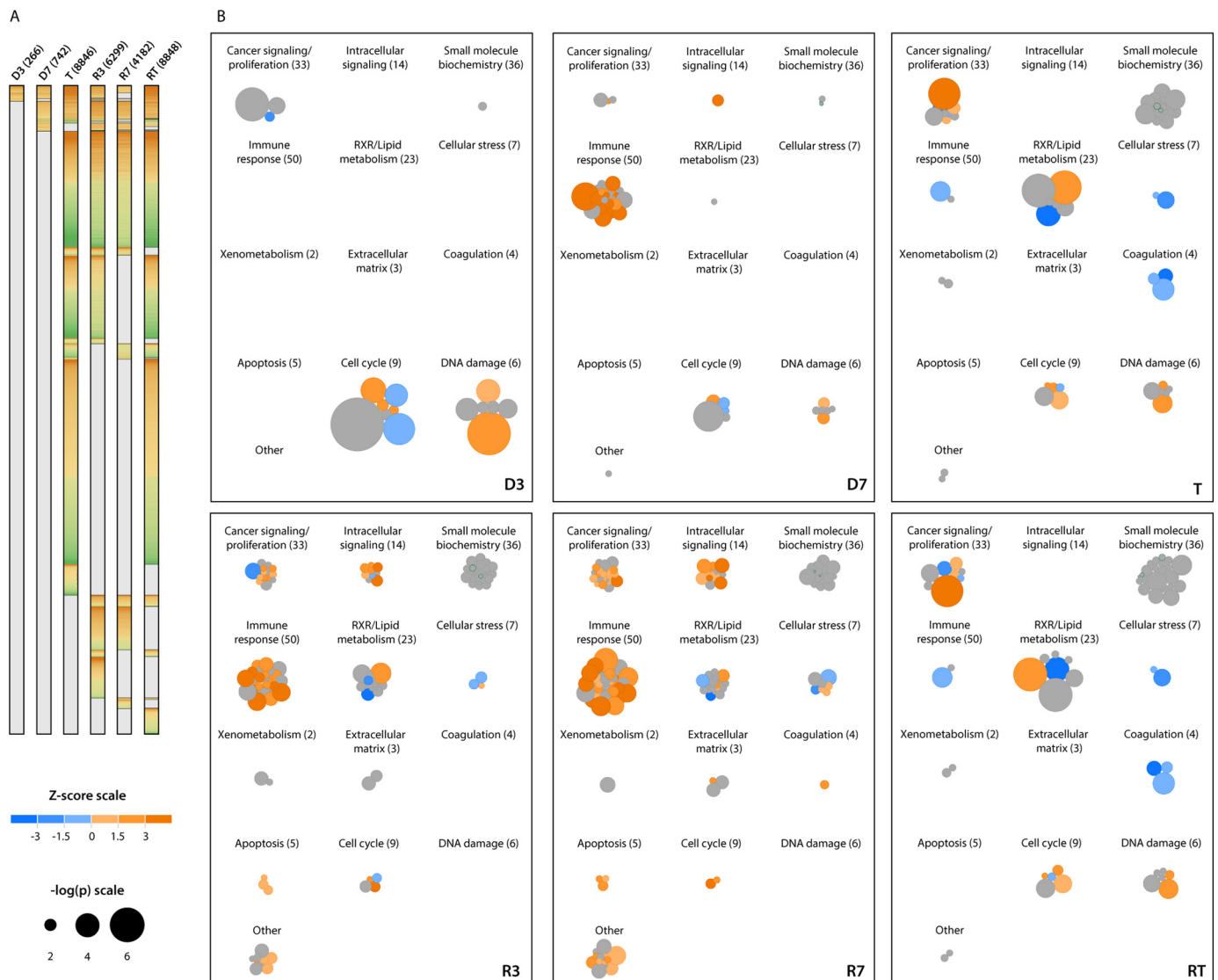


Figure 3. Patterns of gene expression changes during the course of HCC evolution. *A*, differentially expressed transcripts. Numbers in parentheses indicate the number of significant gene expression differences between the indicated samples and control livers, per DESeq FDR-adjusted $q < 0.05$. $n = 5$ samples/group. Transcripts that were overexpressed relative to control liver are colored orange, and those that were under-expressed are colored green. *B*, super pathways of gene expression changes during HCC evolution, regression, and recurrence. Differentially expressed transcripts from *A* were analyzed with IPA (<http://www.ingenuity.com>), which identified a total of 205 pathways significantly involved across the experimental groups. Individual pathways identified by an IPA comparison of each experimental group with control liver are represented by circles, sized by $-\log_2(p)$, and colored according to their predicted activation or inhibition (positive or negative Z-score, respectively). Pathways without prediction information from IPA are colored gray. These individual pathways were grouped into 13 super pathways utilizing IPA's category scheme. Numbers in parentheses indicate total number of member pathways in each super pathway. As an example, the overall changes seen in D3 livers were dominated by three major super pathways as follows: cancer signaling/proliferation, cell cycle, and DNA damage. Among the member pathway groups contained therein were those composed of transcripts regulating activities such as Brca1 and ATM signaling; DNA mismatch, base excision, and double-stranded DNA breakage repair; cell cycle control of DNA replication, and G₂/M checkpoint regulation. To access super pathways interactively and to explore their component pathway groups and transcripts, see <https://prochownik.pitt.edu/hcc>.

scripts was de-regulated as also described for Myc overexpression *in vitro* (39).

Disruption of the above individual pathways persisted in D7 livers but were now displaced by 30 of 50 components comprising an “immune response” super pathway. Nineteen of these 30 members were predicted to be up-regulated, and none were predicted to be down-regulated. The vast majority of the 436 differentially expressed constituent transcripts encoded chemokines, cytokines, and other proteins involved in the innate immune response (Fig. 4A). We postulated that these originated from inflammatory cells infiltrating the liver in response to Myc-induced genotoxic or oxidative stress injury

(35–38, 40) and/or neoantigens produced prior to actual transformation. To examine this, we isolated single cells from control and D7 livers and characterized the resident hematopoietic (CD45⁺) cell populations. Strikingly, the mononuclear infiltrate recovered from D7 livers was increased more than 4-fold (Fig. 4B) and comprised both myeloid and lymphoid cells. IL-10-secreting CD68⁺CD11b⁺ macrophages were disproportionately enriched in this population, consistent with the idea that this infiltrate represented a primarily acute inflammatory response (Fig. 4C). This was also consistent with the finding that a number of the most highly up-regulated chemokines and cytokines, including *Ccl2* (MCP1), *Ccl3* (MIP1 α), *Ccl4*

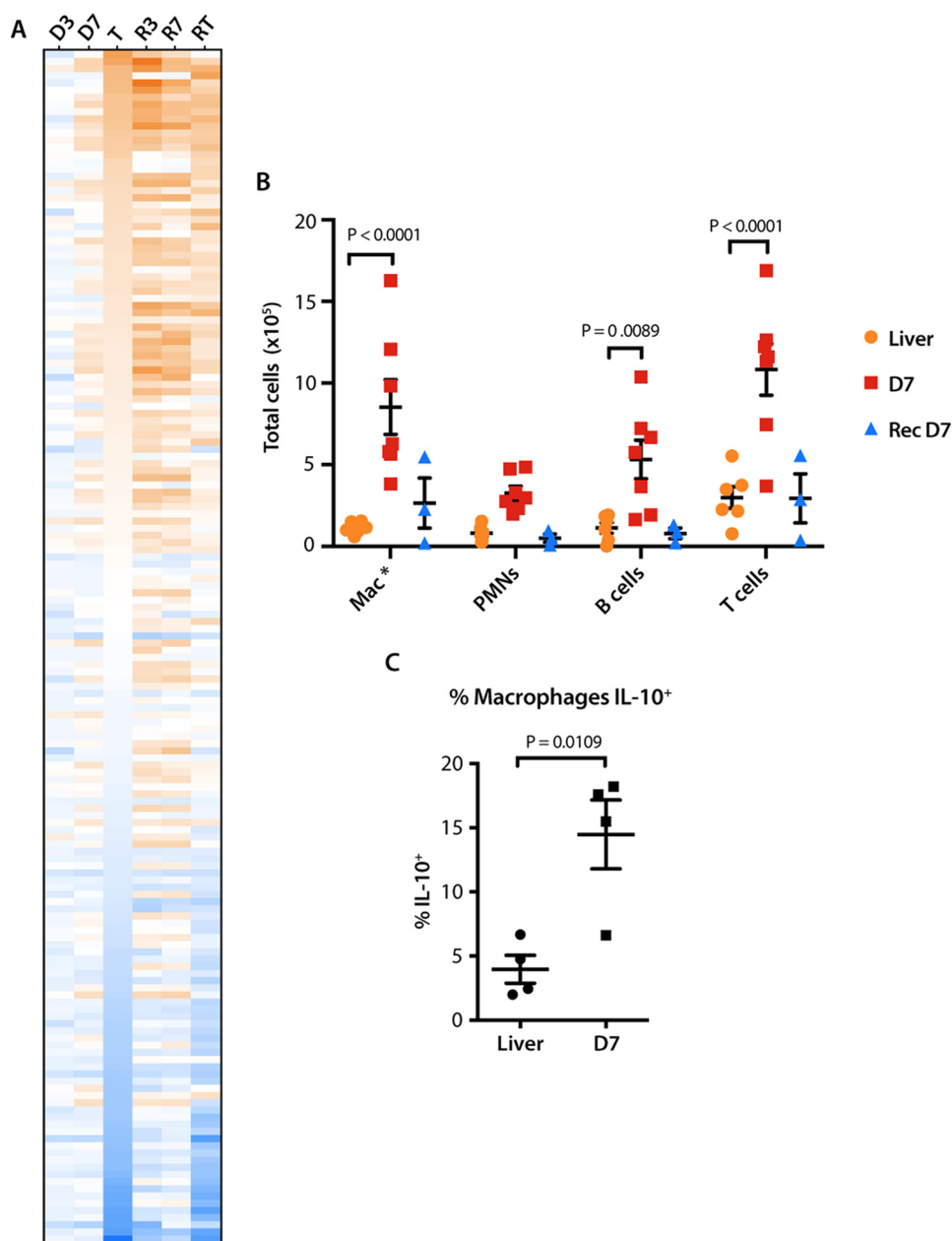


Figure 4. D7 livers demonstrate up-regulation of markers of the innate immune response and infiltration by inflammatory cells. *A*, heat map showing changes in the expression of transcripts encoding inflammatory cytokines, chemokines, and chemotactic factors pertaining to innate and adaptive immunity. The transcript order is arranged from highest to lowest levels of expression in initial tumors. *B*, total numbers of hematopoietic cells isolated from control livers, D7 livers, or regressed livers, which were to again overexpress *Myc* for 7 days (recurrent D7). Compared with two-way ANOVA and Dunnett's multiple comparison two-sided *t* tests. $n = 3-7$ samples/group. *C*, intracellular cytokine staining for IL-10 was performed on infiltrating (CD68⁺, CD11b⁺) macrophages following brief exposure to lipopolysaccharide as described under "Experimental procedures" and compared with a two-sided *t* test assuming homoskedasticity. $n = 4$ samples/group.

(MIP1 β), *Ccl7* (MCP3), and *Ccl22*, are known to be chemotactic for, or activators of, monocytes and/or macrophages (41–45).

Following a 2–3-month period of complete tumor regression, we re-activated *Myc* and re-characterized the liver-associated cellular infiltrates on day 7 by flow cytometry. Surprisingly, there was minimal increase in the number of cells recovered on day 7 post-recurrent *Myc* expression (Fig. 4*B*). Taken together, these results indicated that the extent and nature of the immune response to hepatic *Myc* induction prior to the appearance of any tumors was highly dependent upon whether prior tumorigenesis had occurred.

The "RXR/lipid metabolism" super pathway in D7 livers also demonstrated changes, although only one of its 23-member component pathways, involving 17 transcripts related to phosphoinositide metabolism, was significantly deregulated. This gained increased prominence in initial tumors in which additional six-member pathways, totaling 335 transcripts, were deregulated (Fig. 3). Together, these encoded participants in FAO, lipid, and bile acid synthesis and xenobiotic functions normally ascribed to differentiated hepatocytes (46, 47). Consistent with the idea that tumorigenesis leads to a global state of de-differentiation and the assumption of more fetal-like behaviors (48), we observed widespread down-regulation of the

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“coagulation” super pathway, which involved four member pathways containing transcripts for virtually all clotting factors synthesized by mature hepatocytes (supplemental Fig. S5A). Further supporting the loss of differentiated functions by the normal liver was the >100-fold down-regulation of all detectable transcripts encoding members of the major urinary protein family, which play important roles in pheromone sensing and chemical recognition (supplemental Fig. S5B) (49). A separate “cellular stress” super pathway contained a large set of mostly down-regulated transcripts encoding various acute-phase reactants, which are also typically synthesized by highly differentiated cells (supplemental Fig. S5C) (50). Finally, we noted marked down-regulation of numerous transcripts encoding both class I and class II members of the major histocompatibility complex (supplemental Fig. S5D). These results were consistent with the idea that, beginning as early as 3–7 days after *Myc* induction, hepatocytes dramatically down-regulate multiple transcripts involved in numerous highly differentiated functions, including immune recognition.

Additional pathways within the RXR/lipid metabolism super pathway, including those involved in fatty acid and cholesterol homeostasis (47), were also highly deregulated in tumors. This agreed with the actual changes in FAO observed in these tumors and previously observed in HBs (Figs. 2D and 5, A and B) (30). Additionally consistent with these findings and with the presumptive induction of the Warburg effect was the up-regulation in tumors of virtually all transcripts involved in glucose transport and glycolysis (Fig. 5C and, <https://prochownik.pitt.edu/hcc>, “small molecule biochemistry”). Further noted were changes in the expression of transcripts involved in glutamine uptake and glutaminolysis (Fig. 5, D and E). Specifically, those encoding the Slc1A5 glutamine transporter were up-regulated by 42-fold in HCCs relative to control livers, whereas transcripts encoding glutaminase 2 (*Gls2*), the rate-limiting enzyme of glutaminolysis and glutamate dehydrogenase (*Glud1*), were down-regulated by 5- and 3-fold, respectively (Fig. 5E).

Unlike *Gls2*, which is the liver-specific glutaminase isoform, *Gls1* is expressed at low levels in the liver but at much higher levels in tissues such as kidney and brain (51). Indeed, *Gls2* transcripts were >60-fold more abundant than *Gls1* transcripts in control liver tissues (14,793 reads per kilobase per million (RPKM) versus 241 RPKM) (Fig. 5E). However, *Gls1* is also expressed in fetal liver and by transformed cells and is a *Myc* target (51, 52). Whereas *Gls2* transcripts decreased in HCCs by 5-fold as mentioned above, those for *Gls1* increased by nearly 25-fold (241 RPKM versus 5918 RPKM). When taken together, however, total *Gls1* plus *Gls2* transcripts were decreased by 40% in initial HCCs and by 43% in recurrent HCCs. Immunoblots for *Gls1*, *Gls2*, and *Gludh1* confirmed the results obtained by transcriptional profiling (Fig. 5F). Thus, although *Gls1* and *Gls2* transcripts and proteins underwent significant changes during the course of initial and recurrent tumorigenesis, our results suggest that a switch to glutaminolysis, as a way of providing a glycolysis-independent alternative fuel source for the TCA cycle, does not play a major role in sustaining HCCs as it does in some tumors (53). Rather, the presumptive increase in glutamine content mediated by *Slc1A5* up-regulation is likely channeled into anabolic pathways.

The deregulation of the cancer signaling/proliferation super pathway, first noted in D3 livers and primarily involving cell cycle control and DNA damage repair pathways, assumed an expanded role in initial tumors (396 versus 34 deregulated transcripts). Notably, virtually all transcripts encoding ribosomal proteins, factors involved in ribosomal assembly and mRNA-ribosomal engagement (eIFs), and mTOR signaling now dominated the transcriptional landscape and involved 210 of the 396 de-regulated transcripts (Fig. 6, A–C). Cell cycle and DNA damage transcripts remained de-regulated but were composed of <20% of the differentially expressed transcripts within this super pathway. The rapid tumor regression accompanying *Myc* inactivation (supplemental Fig. S1A) was associated with equally rapid and marked down-regulation of ribosomal protein-, eIF-, and mTOR-related transcripts (Fig. 6, A–C). Also notable was the reappearance of the immune response super pathway. Individual pathways within this category increased from two in initial tumors to 36 at R3 and 46 at R7 (Fig. 3 and <https://prochownik.pitt.edu/hcc>). Many of these pathways were those previously identified only transiently in D7 livers (Fig. 3B).

In humans, predisposing factors for the development of HCC are extremely varied and include hepatitis B and C infection, non-alcoholic steatohepatitis, and alcoholic and non-alcoholic cirrhosis. Not unexpectedly, these tumors also tend to be biochemically and molecularly heterogeneous; in addition to *MYC*, oncoprotein drivers for these cancers include β -catenin, Ras, TERT, and members of the insulin-like growth factor and transforming growth factor families (10, 54–56). To correlate the findings in our *Myc*-driven murine HCC model with human HCCs, we used The Cancer Genome Atlas to compare mRNA expression data between normal liver, non-*MYC*-amplified tumors, and *MYC*-amplified tumors. *MYC*-amplified tumors had 2.2-fold greater expression of *MYC* transcript than non-*MYC*-amplified tumors ($p = 4.4 \times 10^{-11}$). Compared with normal livers, ribosomal protein, eIF, glycolytic, and glutaminolytic transcripts were deregulated in human HCCs much in the same manner as seen in murine HCCs, and this effect was greater in *MYC*-amplified tumors (supplemental Fig. S7).

Transcriptional differences between initial and recurrent tumors

Despite their metabolic differences, initial and recurrent tumors appeared quite similar at the transcriptional level in terms of the composition of their deregulated super pathways and the component member pathways (Fig. 3A and <https://prochownik.pitt.edu/hcc>). However, 441 individual gene expression differences were nonetheless noted between these two cohorts (Figs. 3A and 7A). 409 of these (92.7%) were already deregulated in initial tumors but were disparate in the two tumor groups (Fig. 7B). The remaining 26 transcripts, *i.e.* those falling on or close to the abscissa in Fig. 7B, were those which, although not differentially expressed between livers and initial tumors, were differentially expressed between the two tumor groups (Fig. 7C). Among this small group of transcripts were several whose encoded proteins might participate in the accelerated growth of recurrent tumors (supplemental Fig. S1A). These included *Gdf3*, a member of the TGF- β family of growth

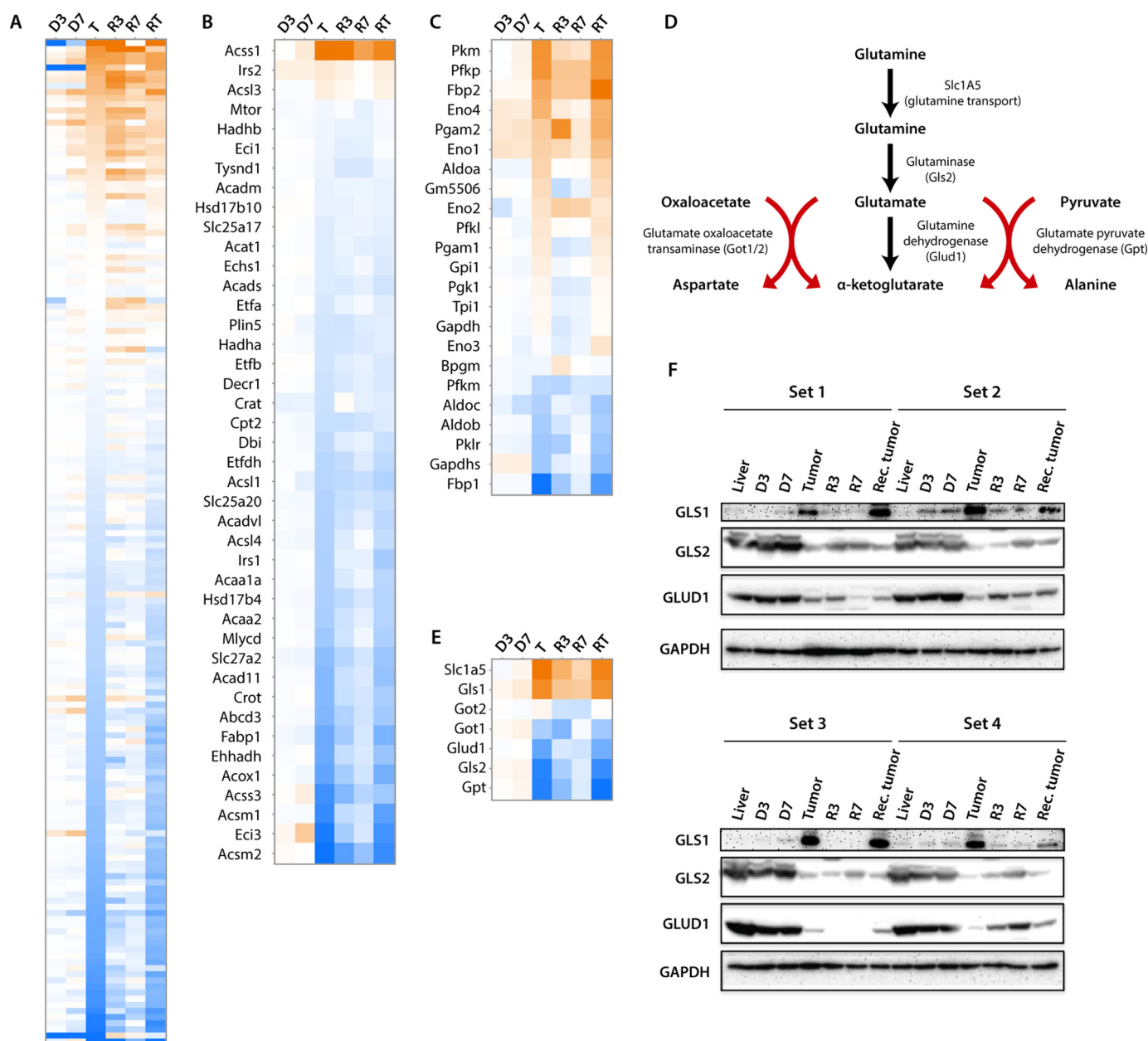


Figure 5. Deregulation of select pathways pertaining to lipid, glucose, and glutamine metabolism in initial HCCs. *A*, heat map of transcripts comprising the combined LXR/FXR/RXR/PXR transcription factor pathways. *B* and *C*, heat maps comprising FAO and glycolytic pathways, respectively. *D*, pathway of glutaminolysis showing the uptake of exogenous glutamine by its transporter, Slc1A5, and its direct conversion to glutamate and α -ketoglutarate mediated by the rate-limiting enzyme glutaminase and glutamate dehydrogenase, respectively. The conversion of glutamate to α -ketoglutarate can also occur via deamination during the glutamate oxaloacetate transaminase-mediated conversion of oxaloacetate to aspartate and the glutamate-pyruvate dehydrogenase-mediated conversion of pyruvate to alanine. *E*, heat map for the six transcripts encoding the enzymes depicted in *D*, plus *Gls1*, which is not normally expressed in normal liver but has been shown to be increased in fetal tissues and tumors and is a known *Myc* target (51, 52). *F*, immunoblots for glutaminase 1 and 2 (*Gls1* and *Gls2*) and glutamate dehydrogenase (*Glud1*) in four sets of tissues from independent groups of mice. See supplemental Table S6 for band density quantification.

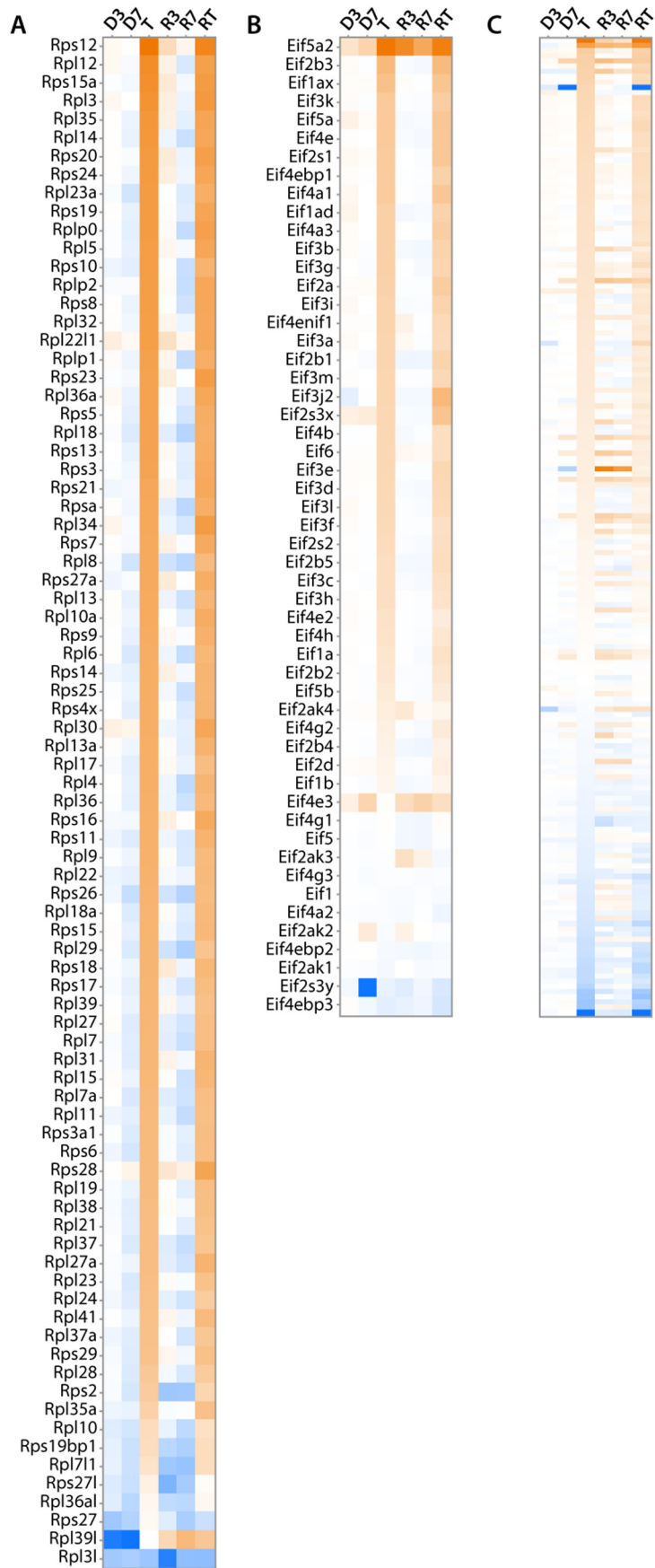
factors, which was up-regulated (58); *Ildr2*, *Pcx*, and *Dhtkd1*, which regulate lipid and carbohydrate metabolism and which were all down-regulated (59–61); and three putative tumor suppressors, *Plk2*, *Ptpn14*, and *Tshz2*, which were also down-regulated (62–64).

Fifty nine percent of the transcripts shown in Fig. 7A encoded metabolic enzymes with 21% of these specifically involved in lipid metabolism. The matching of these with individual metabolic pathways pointed to several potential metabolic differences that were at least partially confirmed by experimental

data. This included a reduction in total PDH protein levels in recurrent tumors that was consistent with the reduced level of overall enzymatic activity (Fig. 1, *C* and *D*).

Finally, an unbiased assessment of the most highly up- and down-regulated transcripts (100 in each group) that distinguished livers and initial tumors (and in which a signal was detectable in each tissue) showed a median up-regulation in tumors of 331-fold (range, 144–3910-fold) and a median down-regulation of 219-fold (range, 95–145,000-fold). Among the former groups of transcripts were those encoding several solute

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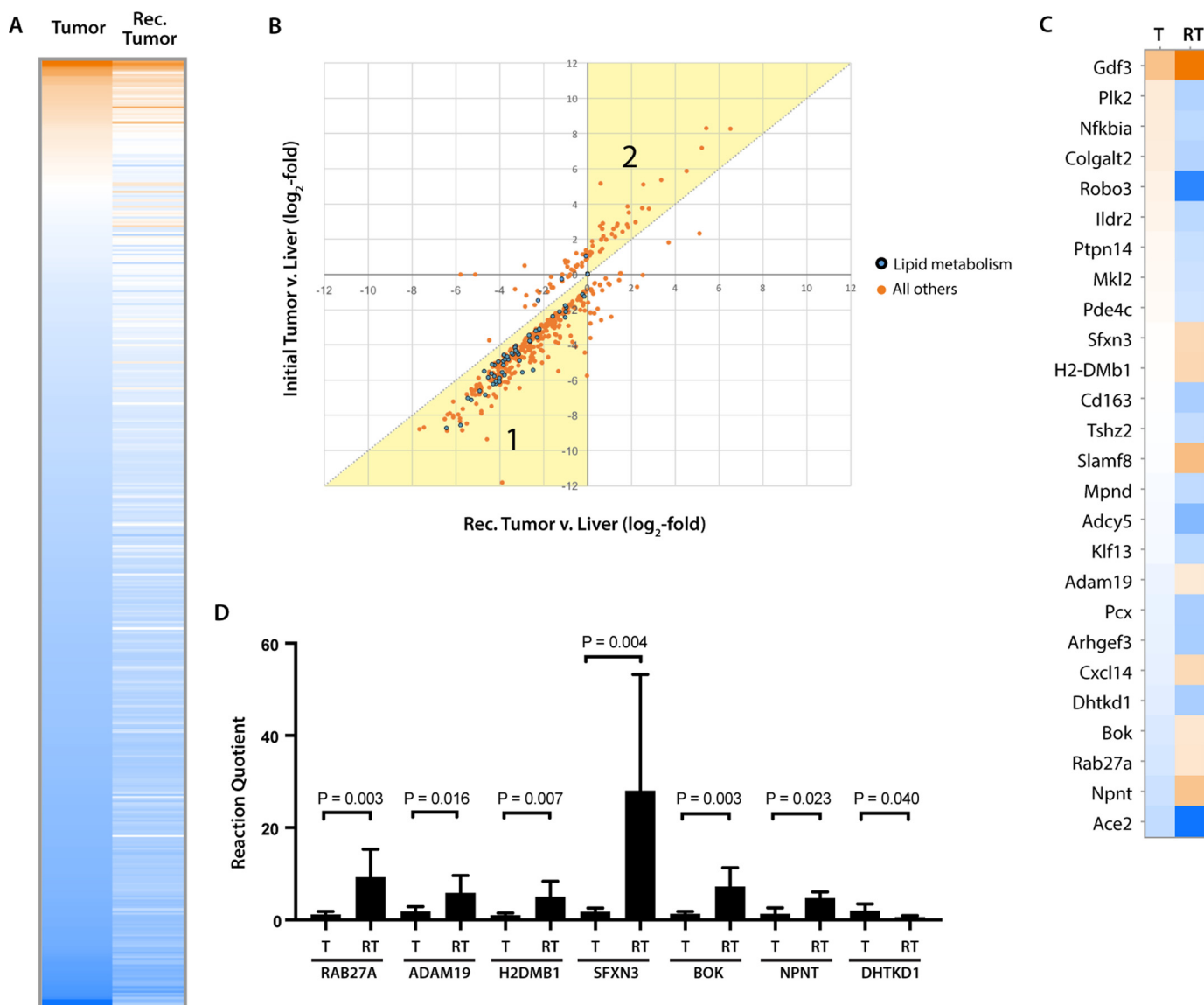


Figure 7. Gene expression differences between initial and recurrent tumors. *A*, heat map of the 441 gene expression differences between initial and recurrent tumors, according to DESeq FDR-adjusted q -value < 0.05 . Transcript levels are arranged so as to reflect their relative abundance in normal liver, with those being the most highly expressed (orange) displayed at the top of the heat map and those being the least highly expressed (blue) displayed at the bottom. *B*, transcripts from *A* showing absolute relative expression in initial tumors (ordinate) versus recurrent tumors (abscissa). The dotted diagonal line denotes the position on which individual transcripts would lie if each tumor group expressed them equally in comparison with liver. Points lying within region 1, highlighted in yellow, indicate transcripts that were relatively overexpressed in recurrent tumors versus initial tumors; and points in region 2 indicate transcripts that were overexpressed in initial tumors versus recurrent tumors. *C*, heat map of the 26 transcripts from *B* lying on or near the ordinate, which were not significantly different between initial tumors and livers but did differ significantly between the two tumor groups. *D*, results of qRT-PCR analysis of selected transcripts from *C*. The high intragroup variability seen with *Sfxn3* reflects similar results obtained from RNA-seq data that showed a 5.5-fold variation in transcript levels. *T*, initial tumor; *RT*, recurrent tumor. See supplemental Table S5 for primer sequences. Each bar represents the mean of triplicate reactions \pm S.E.

carriers dedicated to amino acid and carnitine transport and synthesis (*Slc7a3*, *Slc7a9*, *Slc22a16*, and *Phgdh*) and proteins typically confined to transformed, undifferentiated, or stem cells (*Dlk2*, *Afp*, *Glis1*, and *Peg10*) (supplemental Fig. S6 and supplemental Table S2). Transcripts for insulin-like growth factor 2 (*Igf2*), the long non-coding RNA *H19* and Lin-28B were also up-regulated by 2000-, 3000-, and 1200-fold, respectively

(supplemental Table S4). Although not among the top 100 deregulated transcripts, those encoding *Igf2bp3* and *Igfbp1*, which enhance *Igf2* translation and stabilize the protein, were increased by 23- and 20-fold, respectively. Similar results were obtained for recurrent tumors (supplemental Table S4) although, as a group, the median fold up-regulation of these top 100 transcripts was only 232-fold ($p = 0.012$).

Figure 6. Changes in select transcripts comprising the cancer signaling/proliferation super pathway. In contrast to the 34 differentially expressed transcripts comprising members of the "hereditary breast cancer," "gadd45," and "aryl hydrocarbon receptor" signaling pathways (19, 6, and 9 transcripts, respectively) seen in D3 livers, a total of eight pathways involving a total of 396 transcripts were deregulated in the super pathway in initial tumors. Most notably, these involved marked up-regulation of transcripts encoding ribosomal proteins (A), factors that facilitate mRNA engagement of and translation by the ribosome (B), and members of the mTOR pathway, which integrates a variety of growth-promoting signals and serves as a nutrient sensor (C) (57).

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Again consistent with the previously noted loss of highly differentiated functions by tumors, the 100 most highly down-regulated transcripts encoded several classes of proteins involved in xenobiotic metabolism (e.g. 17 members of the Cyp450 family and four members of the Ugt family), amino acid catabolism (*Ido2* and *Tat*), solute carriers involved in bile salt and non-amino acid ion transport (14 Slc and Slco transporters), and multiple members of the major urinary protein family (supplemental Fig. S6 and supplemental Table S3). As was the case with up-regulated transcripts, the median down-regulation of these 100 transcripts in recurrent tumors was only 55-fold versus 219-fold ($p < 10^{-4}$).

To independently confirm at least some of our RNA-seq results using a different method, we used qRT-PCR to compare transcript levels for a subset of the transcripts listed in Fig. 7C. As shown in Fig. 7D, we found excellent agreement with all tested transcripts.

Discussion

These studies demonstrate that the metabolic and molecular alterations associated with HCC pathogenesis and evolution mediated by Myc overexpression occur gradually, sequentially, and cumulatively. The timing and ordering of these events are defined as they are during the course of tumor regression. Finally, we show that recurrence, following a period of tumor regression and sustained quiescence, differs from initial tumors in a number of ways that involve a relatively small number of key genes and pathways.

Whereas initial tumors reached maximal size within 21–28 days, recurrent tumors required about half this time (supplemental Fig. S1A). Potential reasons include an inherently faster tumor growth rate and/or the presence of residual dormant tumor cells that survive after total regression and remain quiescent until being stimulated by Myc re-expression (22). Such cells, not present in naive livers, might represent a particularly susceptible and responsive population, having previously accumulated additional genetic lesions necessary for initial tumor induction but now requiring only Myc re-expression (22). The larger S-phase fraction present in recurrent tumors (supplemental Fig. S1B) is consistent with the former model but does not rule out additional contributing factors. In fact, these studies indicated that initial tumors were almost exclusively diploid or pseudodiploid and thus representative of a subpopulation of hepatocytes that are otherwise of variable ploidy (supplemental Fig. S1B) (23). This is consistent with findings that polyploid hepatocytes tend to have lower replicative capacity, that fetal and embryonic hepatocytes and human HCCs are almost exclusively diploid, and that tumors are often enriched for self-renewing diploid stem cell populations (65–68). Thus, diploid hepatocytes are either more susceptible to Myc-mediated transformation or, more likely, outcompete their hyperdiploid counterparts during tumor expansion, possibly because of their higher stem cell content and/or inherently faster doubling times. Recurrent tumors contained more cells in S-phase than their initial tumor counterparts yet appeared to be more aneuploid (supplemental Fig. S1B), suggesting a causal relationship between these two features and implying that Myc overexpression can lead to aberrant DNA synthesis without intervening

mitoses. Increased DNA synthesis without mitosis results in the generation of tetraploid or pseudotetraploid cells, particularly if Tp53 is absent or mutated (37, 38, 69). Inactivation of the p19^{ARF}/MDM2/p53 tumor suppressor pathway appears to be a relatively late event in initial HCC tumorigenesis but persists in recurrent tumors,⁴ suggesting that aneuploidy may develop following prolonged Myc overexpression in the absence of this protective tumor suppressor background (37, 38). The deregulation of certain cell cycle regulators, particularly cyclin B1 (supplemental Fig. S4), may further enhance the development of tetraploidy (37). Indeed, it is notable that transcripts encoding cyclin B1 and cyclin B2 were the most highly up-regulated of all cyclins in both initial and recurrent tumors (supplemental Fig. S4). Although aneuploidy is not uncommon in human HCCs, neither its relationship to Tp53 mutation/loss nor its prognostic implications have been firmly established (supplemental Fig. S1B) (28, 67, 68, 70).

Both initial and recurrent tumors tended to be associated with increased expression of β -catenin and even higher levels of YAP (Fig. 1A). Although the regulation of Myc by β -catenin and less so by YAP is well known (25, 71), the reverse has been less well studied. A positive autoregulatory loop by which Myc regulates YAP in liver cancer has been described, although it is not clear whether this is a direct consequence of Myc overexpression or instead involves β -catenin as an intermediate (72, 73).

Recent work has documented that most recurrent human cancers are descendants of the initial malignant clonal population but contain additional genetic or epigenetic changes with distinct functional consequences (16, 74–81). However, it remains unclear how and to what extent the emergence and evolution of these recurrent clones are influenced by therapeutic interventions. Moreover, little is known regarding how these newly acquired alterations relate to specific biochemical and metabolic differences between primary and recurrent tumors. The current model allows correlations among these differences while avoiding the confounding factors introduced by genotoxic therapies and permits the identification of what are likely to be only the most relevant and direct changes associated with HCC recurrence.

Although recurrent HCCs are also clonal descendants of the initial tumors (22), we noticed that many of their metabolic and molecular abnormalities were less pronounced relative to those of initial tumors. For example, YAP expression was consistently lower in recurrent tumors, and the levels of β -catenin were virtually indistinguishable from those in control livers. Indeed, even the levels of doxycycline-driven human Myc expression tended to be lower (Fig. 1A and supplemental Fig. S1A). Additionally, the PDH enzymatic activities, ATP levels, and ETC function of recurrent tumors more closely resembled those of normal livers (Figs. 1, D and E, and 2B). Finally, as a group, the most highly deregulated transcripts were less altered in recurrent tumors than in initial tumors (supplemental Tables S2 and S3). In contrast, certain changes were indistinguishable between the two tumor sets, including the dysregulation of Pdk1

⁴ S. Kulkarni, J. M. Dolezal, H. Wang, L. Jackson, J. Lu, B. Frodey, A. Dosunmuu-Ogunbi, Y. Li, M. Fromherz, A. Kang, L. Santana-Santos, P. V. Benos, and E. V. Prochownik, manuscript in preparation.

and Pdp2, the suppression of FAO, and the down-regulation of the glutamine catabolic pathway (Figs. 1C, supplemental Fig. S2, B and D, and Fig. 5F). Elsewhere, we have shown the loss of p19^{ARF} expression is similar in the two tumor types.⁴ Collectively, these results indicate that, while growing even more rapidly than initial tumors, recurrent tumors can “normalize” some aspects of their metabolism. This suggests that some of the changes seen in initial tumors are less important for driving the growth of recurrent tumors and/or that they are normalized as a result of compensatory changes in other pathways.

A possible example of such compensatory mechanisms and how they communicate with and reinforce one another may exist with regard to energy production by initial tumors. Despite their high ATP content and PDH activity (Fig. 1, D and E), their overall mitochondrial function was markedly lower than that of normal and proliferatively quiescent livers. This reduced activity was at least partially reflected by a reduction in mitochondrial mass, which we have also witnessed in HBs and is a common feature of many tumor types (Fig. 2B) (30, 32). Moreover, initial tumors demonstrated significantly lower activity of complex II, which serves as the major electron donor in both livers and HCCs (Fig. 1B). In contrast, the levels and activities of ATP, PDH, and complex II in recurrent tumors more closely resembled those of normal liver. It seems likely that the lack of mitochondrially derived ATP in initial tumors is overcompensated for by an increase in glycolysis (the Warburg effect), which also supplies necessary anabolic precursors in the form of glycolytic intermediates. As a result of this metabolic diversion, initial tumor cells may thus be relatively depleted of pyruvate despite being relatively oversupplied with ATP. We have observed a somewhat analogous PDH response in murine HBs induced by the overexpression of mutant forms of β -catenin and YAP, and we have proposed that this might represent an attempt to convert the remaining limiting amounts of pyruvate to AcCoA, which can then either supply the TCA cycle or anabolic pathways (30). Although the precise reason why recurrent HCCs normalize these features remains to be determined, our results demonstrate that, even in the absence of selective pressures such as chemotherapeutic intervention, tumors can reprogram their metabolic circuits to their benefit. Moreover, the nature of these changes appears to be dependent upon the tumor's past history.

One exception to the above-noted metabolic rewiring was the marked reduction in FAO by both initial and recurrent tumors, indicating that fatty acids are not a major source of energy. Rather, it is likely that exogenous fatty acids, such as the palmitate used to conduct these studies, are utilized directly by tumors for *de novo* lipid biosynthesis rather than being transported into the mitochondria, oxidized to AcCoA, and re-synthesized into lipids. Glutamine does not appear to be directed into the TCA cycle in the form of α -ketoglutarate in either initial or recurrent tumors (Fig. 5, E and F) or in HBs (30), suggesting that glutaminolysis does not compensate for the anabolic diversion of glycolytic intermediates (53). The coordinate up-regulation of numerous transcript encoding transporters for glutamine and other amino acids further supports this notion (supplemental Table S2). In the case of glutamine, however, this interpretation must be made cautiously due to the

complex nature of the changes in Gls1 and Gls2 transcripts and proteins. Gls1 is not typically expressed at high levels in liver, whereas Gls2 is a liver-specific isoform (51). However, these relative levels of expression can change in fetal liver as well as in tumors as we and others have demonstrated (Fig. 5, E and F) (51, 52, 82). Despite such changes, the cumulative levels of *Gls1* and *Gls2* transcripts in tumors were reduced by ~40% in tumors (Fig. 5E). It is also noteworthy that high levels of Gls2 expression have recently been shown to impede the growth of HCCs, thereby suggesting that diverting the supply of glutamine too much in favor of catabolic rather than anabolic pathways can be detrimental to tumor growth (83). In Myc-induced HCCs similar to those described here, Yuneva *et al.* (82), have shown total glutaminase activity to be increased in isolated mitochondria under conditions that provided maximal levels of glutamine and other TCA cycle substrates. However, that same study was unable to measure intratumoral glutamine levels, leaving open the possibility that low levels of the amino acid reflect its immediate use for *de novo* protein synthesis. High levels of enzyme activity therefore could reflect a compensatory response to a relative glutamine deficit without necessarily increasing the rate of conversion of glutamine to glutamate. Similarly increased demands on glutamate for protein synthetic purposes might further impair the ultimate conversion of glutamine to α -ketoglutarate.

The uncoupling of complex II's electron transport function from SdhA's catalytic activity in initial tumors, and beginning quite early after Myc induction (Fig. 2, B and C), leaves open the possibility that other amino acids such as valine, isoleucine, and methionine may circumvent the glutamine degradation pathway and furnish succinate, thus providing an alternative source of electrons in the form of FAD. However, such a compensatory pathway, particularly when considered in light of the overall lower rate of Oxphos in initial tumors (Fig. 2B) (30), would be expected to extract a metabolic cost in the form of enhanced ROS production as the electrons generated during the succinate to fumarate conversion are now donated to non-physiological acceptors such as molecular oxygen (34). It will be of interest in future work to determine whether the uncoupling of SDH represents the primary means by which Myc overexpression generates ROS (36–38, 40).

Only 157 transcripts were deregulated in response to Myc overexpression at the earliest time point examined (D3) with the vast majority of these (144 (92%)) being involved in cell cycle control, DNA damage response, and DNA repair (Fig. 3, A and B, and supplemental Fig. S4). Indeed, within the cell cycle super pathway (Fig. 3B) was a component IPA-derived member pathway designated “p53 signaling,” 9 of whose 111 members were up-regulated with none being down-regulated. Three of these component transcripts (*Tnfrsf10b*, *Apaf*, and *Birc5*) encoded proteins involved in the regulation of apoptosis, whereas the remaining six (*Brca1*, *Chek1*, *Chek2*, *Pcna*, *Topbp1*, and *E2f1*) encoded proteins involved in DNA damage and cell cycle arrest by virtue of their communication with TP53.

The numerous transcripts encoding hematopoietic chemokines and cytokines expressed by D7 livers correlated with a pronounced acute inflammatory infiltrate. Although tumor cells themselves can express similar factors for purposes such as

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angiogenesis (84, 85), it seems more likely that the ones we have identified were in fact of hematopoietic origin. The inflammatory infiltrate may arise as a result of hepatocyte death mediated by Myc overexpression, local necrosis, and/or from the expression of neo-antigens arising from ongoing mutations induced by replication errors, chromosomal breaks, and ROS production (36–38, 40, 86, 87). The absence of such an infiltrate in the livers of mice with previously regressed tumors following an otherwise identical 7-day re-induction of Myc (Fig. 4B), indicates either that the initial intrahepatic stimulus that prompted the inflammatory infiltrate was no longer present or that hematopoietic cells were unresponsive. Although we did not perform RNA-seq on this particular cohort, the presence of a similar pattern of immune-related transcript dysregulation in both initial and recurrent tumors (Fig. 4A) suggests that the rate of hematopoietic infiltration during the re-induction phase may be slower relative to that of the initial induction phase.

Relevant to these findings is a recent report by Guo *et al.* (88) who utilized a different murine HCC model induced by the hydrodynamic tail injection of mutant YAP. Although these animals do not develop tumors until 12–16 weeks, the homing of macrophages to YAP⁺ tumor-initiating hepatocytes could be detected as early as 2 days post-injection and well before the development of any discernible tumors. Eliminating these macrophages strongly inhibited tumorigenesis. The authors speculated that these macrophages play a decisive role in the survival of single tumor-initiating cells *in vivo* by inhibiting their clearance and facilitating their expansion. Our failure to detect macrophages after tumor re-induction suggests that they may no longer be needed because previously selected tumor-initiating cells from the initial tumor persist during regression and need not be generated anew (22). A similar phenomenon with regard to mast cell recruitment, as well as macrophage infiltration, in a Myc-driven model of pancreatic islet cell tumors has also been described (89). Taken together, these findings suggest that both Myc and non-Myc-driven tumors may participate in a symbiotic relationship with the innate immune system.

Igf2 and *H19*, both of which were highly up-regulated in initial and recurrent tumors (supplemental Table S4) are imprinted and physically adjacent loci but are independently transcribed from paternal and maternal alleles, respectively. Both are highly expressed in some human fetal tissues and in lung cancer, Wilms' tumor, and HCC, with IGF2 being a particularly potent driver of the latter (90–93). *H19* is a direct Myc target as is *Lin-28b*, which positively regulates *Igf2* expression (20, 94). *Lin-28b* also down-regulates several let-7 micro-RNAs with tumor suppressor functions (95). Although not as highly up-regulated as transcripts for *Igf2*, *H19*, and *Lin-28b*, those encoding *Igfbp1* and *Igf2bp3* were also up-regulated significantly in both initial and recurrent tumors (20- and 23-fold, respectively). *Igfbp1* binds to and prolongs the serum half-life of *Igf2*, and both it and *Igf2bp3*, which binds to and stabilizes *Igf2* mRNA, have been reported to drive cancer growth and to be associated in some cases with an adverse prognosis (96). In contrast, *Igfbp5* and *Igfbp7*, which were down-regulated in HCCs by 5- and 2-fold, respectively, are negative regulators of IGF2 and thus possess potential tumor suppressor functions

(97–99). These findings, together with the 25-fold down-regulation of *Igf1* transcripts (supplemental Table S4), support the idea that HCC tumor growth is at least supported by, if not entirely dependent upon, complex and coordinate activities of members of the H19-Igf2 axis.

In summary, our analyses identified a number of molecular, metabolic, and biochemical pathways that are perturbed and/or re-programmed over the course of HCC tumor evolution in response to a single oncogenic stimulus (Myc). Importantly, this has enabled us to capture some of the earliest events that respond to Myc overexpression, including the induction of many cell cycle-regulated and DNA repair genes and the reprogramming of specific metabolic pathways, all of which occur prior to any changes in liver mass or appearance. Some of these deregulated pathways, particularly those pertaining to glycolysis and glutaminolysis, have been previously identified at the transcriptional level (24). By examining the fate of [¹³C]pyruvate during the course of tumor evolution, that study also showed that this end-stage glycolytic intermediate is first converted into alanine during the earliest pre-neoplastic stages of Myc induction but that, at later times, and as glycolysis becomes a primary mode of providing energy and anabolic substrates, pyruvate is re-directed into lactate production (24). This study adds to the molecular, biochemical, and metabolic changes that accompany evolution, regression, and recurrence by integrating these and additional pathways, particularly those pertaining to fatty acid and pyruvate oxidation and mitochondrial oxidative phosphorylation, electron transport chain function, and the adaptation of mitochondrial mass to accommodate these changes. The analyses of initial, regressing, and recurrent tumors are also unencumbered by the confounding factors that would otherwise be introduced in response to chemotherapeutic interventions. Approaches of this type should be useful for the evaluation of other tumor models.

Experimental procedures

Animals

All animal studies were reviewed and approved by the University of Pittsburgh Institutional Animal Care and Use Committee. FVB/N-Tg(tetO-MYC)36aBop/J and LAP-tTA mice (B6.Cg-Tg[Cebpb-tTA]5Bjd/J) (The Jackson Laboratory) (22) were genotyped according to published protocols and as recommended by the vendor (31). Animals were housed and fed *ad libitum* except that drinking water contained doxycycline (100 μg/ml) unless otherwise stated. For tumor induction studies, six groups of 3–4-week-old male and female mice were provided with doxycycline-free water. Three and 7 days later (D3 and D7), well before the appearance of any tumors, two groups of mice were sacrificed. Liver samples were flash-frozen for molecular and biochemical analyses or used immediately for enzymatic and metabolic assays requiring fresh tissues. The remaining mice were maintained off doxycycline for a further 3–4 weeks by which time all had developed large and aggressive HCCs. One group was sacrificed and used to duplicate the above studies, and the remaining three groups were again provided doxycycline-containing drinking water to suppress Myc expression and initiate tumor regression. Regressing tumors

were then sampled 3 and 7 days later (R3 and R7 groups), at which point they were at least 50% smaller. Doxycycline was maintained for an additional 2–3 months in the remaining group and was then again removed to induce recurrence.

Cell cycle analysis

Excised tissue was fractionated with a NE-PER isolation kit using the directions provided by the supplier (Thermo Fisher Scientific, Rockford, IL). Tissue was first washed with ice-cold PBS, minced into small pieces, then homogenized in cytoplasmic extraction buffer (~100 mg tissue/1 ml of buffer), and passed through a 70- μ m nylon mesh cell strainer (Thermo Fisher Scientific, Pittsburgh, PA). Homogenates were then centrifuged at $500 \times g$ for 5 min at 4 °C; the resultant pellet was the nuclear fraction. The final nuclear pellet was resuspended in 1 ml of ice-cold 10 mM Tris-HCl, pH 8.0; 10 mM NaCl; 0.1% Nonidet P-40; 0.07 mM propidium iodide (Sigma). RNase A (Sigma) was added to a final concentration of 10 μ g/ml, and the nuclei were incubated on ice for 10–15 min. Flow cytometry and downstream analyses were performed as described previously (99).

PDH assays and ATP assays

PDH complex activity on fresh tissue samples was determined as described previously (30). Briefly, ~50 mg of tissue was minced in ice-cold DMEM lacking glucose, pyruvate, and glutamine and gently homogenized by several rounds of passage through an 18-gauge needle. 0.5-ml aliquots were then placed into 0.5 ml of $2 \times$ assay buffer (30) containing 0.075 μ Ci of [1- 14 C]pyruvate (PerkinElmer Life Sciences). The rubber stoppers used to seal the tubes contained a central hanging basket (Kimble-Chase Life Science and Research Products, Rockwood, TN) with a 25-mm glass microfiber filter (Whatman/GE Healthcare) saturated with 0.5 M KOH. Reactions were incubated at 37 °C with continuous motion and terminated by injecting 0.5 ml of 4.0 M perchloric acid through the stopper. 14 CO₂ was collected onto the filters for 40 min at 37 °C and quantified by scintillation counting.

ATP levels were quantified on tissue lysates using the ATP Lite ATP detection system (PerkinElmer Life Sciences) according to the instructions of the supplier. Each point was assayed in at least triplicate and then normalized to protein content. All points were then expressed relative to the ATP levels measured in control liver tissues (30). Tissues from at least five animals were assayed for each time point.

Fatty acid β -oxidation and glutaminase assays

FAO was measured as described previously (30, 31). Briefly, ~50 mg of tissue was minced in SET buffer and gently disrupted with a Potter-Elvehjem homogenizer on ice. Triplicate ~5- μ l aliquots of the homogenate were added to 195 μ l of FAO reaction buffer containing 1 μ Ci of BSA-conjugated 3 H-labeled palmitate (PerkinElmer Life Sciences) (30, 31) and incubated for various times at 37 °C. Reactions were terminated by adding 40 μ l of 1 M KOH, and newly synthesized acyl-carnitine esters were hydrolyzed at 60 °C for 1 h. 40 μ l of 4 mM perchloric acid was added for an additional hour on ice. After organic extraction, the water-soluble 3 H-labeled FAO products were quantified by scintillation counting.

Immunoblotting

Western blotting was performed as described previously (30, 31, 100) with at least 4–5 independent groups of tissues as indicated in the appropriate figures and supplemental figures. All antibodies, their sources, and conditions for their use are listed in supplemental Table S1. Quantification of Western blotting band densities was performed using ImageJ 1.50 (101). Relative band densities are reported in supplemental Table S6 relative to liver, except for blots with no detectable liver band.

Measurements of Oxphos

Respirometry studies were performed as described previously (30, 31). For all measurements, livers from littermate control mice of approximately the same age were evaluated in parallel. Briefly, ~50 mg of tissue was disrupted in ice-cold MiR05 buffer. 300 μ g of the lysate was then used to quantify OCR using an Oroboros Oxygraph 2k instrument (Oroboros Instruments, Inc., Innsbruck, Austria). Following cytochrome *c* addition to stabilize mitochondrial membranes (31), malate, pyruvate, ADP, and glutamate were sequentially added to initiate electron flow via complex I followed by the addition of succinate to measure the combined activities of ETC complexes I + II. Rotenone was then added (0.5 μ M final concentration) to inhibit complex I and to allow for the quantification of pure complex II activity. Final activities were normalized to total protein content.

BNGE and *in situ* enzymatic assays

BNGE and visualization of ETC complexes by non-denaturing gel electrophoresis was performed as described previously (31, 100, 102). Briefly, purified complexes were electrophoresed on a 3–12% gradient native PAGE Novex BisTris gel (Invitrogen) using the buffer provided, stained with Bio-Safe Coomassie G-250 (Bio-Rad), and then de-stained with water. Band intensities were quantified using an AlphaEase FC2200 scanner and software (Alpha Innotech, Inc., San Leandro, CA). *In situ* enzymatic assays were performed for complex I (NADH ubiquinone oxidoreductase), complex III (decyubiquinol cytochrome *c* oxidoreductase), complex IV (cytochrome *c* oxidase), and complex V (ATPase) as described previously (30, 31, 100). Band intensities were quantified using ImageJ software (National Institutes of Health) (101) and normalized to their corresponding bands on the Coomassie-stained gel. Because *in situ* assays for complex II (succinate dehydrogenase) were poorly reproducible, measurements were performed *in vitro* on total lysates (100). Enzymatic activities were assessed at 1-min intervals over a span of 1 h on a BMG LabTech FLUOstar Omega spectrophotometer (500 nm) (Cary, NC). $\Delta A_{340}/\text{min}$ was obtained using the maximum linear rate over a period of 20 min.

Mitochondrial DNA (mtDNA) quantification

Cellular DNAs were purified from all samples, and mtDNA was quantified with a TaqMan-based assay, as described previously (30, 103), using two separate sets of PCR primers designed to amplify short segments of the cytochrome *c* oxidase I gene (probe set 1) and the D-loop region (probe set 2), respectively. PCR products were normalized to a similarly sized product of the

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nuclear apolipoprotein B gene in a parallel reaction. All primers and probes were synthesized by IDT, Inc. (Coralville, IA).

RNA purification, RNA-seq, and qRT-PCR

RNA-seq acquisition and data analyses have been previously described (30, 31). Briefly, liver and tumor RNAs (five samples per time point) were purified using RNA Easy columns (Qiagen, Inc., Valencia, CA). RNA concentrations were determined with a Nanodrop ND-1000 instrument (NanoDrop Technologies Inc., Wilmington, DE), and RNA integrity was assessed with an Agilent 2100 Bioanalyzer (Agilent Technologies, Santa Clara, CA). Single-end sequencing was performed on an Illumina NextSeq 500 sequencer (Illumina, Inc., San Diego). Read counts were normalized across samples and assessed for significance using DESeq, which determined transcript abundance (expressed as RPKM) and statistical significance when comparing transcript abundance levels between experimental groups and control liver (FDR-adjusted q -value < 0.05). Heat maps of specific gene sets were generated by comparing expression values in each experimental group to control liver, expressed as \log_2 -transformed fold-change.

Pathway analysis for each experimental group was performed using IPA (www.qiagen.com/ingenuity). A maximum of 5000 genes could be accommodated by the IPA program at any given time; thus, the following criteria were used in the experimental groups to select genes for pathway analysis: D3 *versus* liver, 266 significant differences (selected criteria included q -value < 0.05 (266 genes)); D7 *versus* liver, 742 significant differences (selected criteria included q -value < 0.05 (742 genes)); initial tumor *versus* liver, 8846 significant differences (selected criteria included q -value < 0.01 and $|\log_2$ -fold change| > 1.4 (4801 genes)); R3 *versus* liver, 6299 significant differences (selected criteria included q -value < 0.01 (4774 genes)); R7 *versus* liver, 4182 significant differences (selected criteria included q -value < 0.05 (4182 genes)); recurrent tumor *versus* liver; 8848 significant differences (selected criteria included q -value < 0.01 and $|\log_2$ -fold change| > 1.3 (4796 genes)).

Significant pathways were identified by a Bonferroni-Hochberg corrected (FDR-adjusted) p value of < 0.01 . For each pathway in each experimental group, the following were recorded for later display: overall significance (p value), ratio of dysregulated genes in the experimental sample compared with all possible genes associated with that pathway (ratio), and predicted pathway activation or inhibition (Z-score, where applicable). Not all pathways included the ability to estimate a Z-score.

Following the above analyses, pathways were grouped into 13 larger super pathways using IPA's category classification scheme. This included such broad groups as cancer signaling/proliferation, small molecule biochemistry, and apoptosis. RXR, LXR, and FXR signaling pathways were combined with those involving lipid metabolism into the larger super pathway termed RXR signaling/lipid metabolism. Although some pathways could be classified into more than one super pathway (e.g. many cancer-related pathways could be classified into both DNA damage and cancer signaling/proliferation), the final inclusion was based on the first-listed IPA category under the category description.

To compare initial and recurrent tumors in an unbiased manner, the top and bottom 100 expressed transcripts in initial tumors were compared with the same transcripts in recurrent tumors. Mean and median expression were compared between the two groups, the former of which was formally tested for significance with a one-sided t test assuming unequal variance.

qRT-PCR analyses were performed for a small subset of relevant transcripts to confirm the results obtained by RNA-seq. For these analysis, we used a Power SYBR Green RNA-to-CtTM 1 Step kit (Applied Biosystems, Inc., Foster City, CA) using the conditions recommended by the supplier. PCR primers were synthesized by IDT, Inc. (Coralville, IA). Sequences of these primers and the conditions used for amplification are listed in [supplemental Table S5](#). All qRT-PCRs were performed with an Applied Biosystems Step One Plus Real Time PCR System.

Transcriptional profiling of human HCCs and livers

Gene expression data for human HCCs were obtained from The Cancer Genome Atlas (TCGA) using the University of California Santa Cruz Xena Browser (UCSC Xena, <http://xena.ucsc.edu>). Data were stored under the cohort name "TCGA Liver Cancer (LIHC)" with the data tag "gene expression RNA-seq (poly(A)⁺ IlluminaHiSeq)." *MYC* gene amplification data were accessed via the data tag "copy number (gistic2thresholded)" and queried for "Myc." Tissues and normal tissues were identified with the data tag "phenotypes – sample type." Heat maps of gene expression were generated by comparing average gene expression in each group [*MYC*-amplified (High Myc) and *MYC*-non-amplified (Low Myc)] to the average gene expression in normal tissues.

Isolation and antibody labeling of liver-associated leukocytes

Fresh livers were finely minced and disrupted by pressing through a 70 μm nylon mesh screen into 20 ml of ice-cold PBS. 10 ml of 100% Percoll (GE Healthcare Biosciences, Pittsburgh, PA) and 300 units heparin (Thermo Fisher Scientific, Hampton, NH) were then added. The homogenate was centrifuged at $800 \times g$ for 20 min at 22 °C, after which the supernatant was aspirated, and red blood cells were eliminated by incubating for 3 min in ammonium chloride lysis buffer. Cells from the Percoll phase were washed once in FACS buffer (2% FBS in PBS), passed through a 40 μm mesh screen, and enumerated on a Vi-cell cell viability analyzer (Beckman Coulter, Inc. Brea, CA).

Leukocytes were labeled for flow cytometry by first blocking the Fc receptor with anti-mouse CD16/CD32 (BD Biosciences) for 5 min at 4 °C and then incubating with the antibodies listed in [supplemental Table S1](#) for 15 min at 4 °C. Viable cells were enumerated using a Live/Dead Fixable Aqua staining for 30 min according to the manufacturer's instructions (Thermo Fisher Scientific), followed by two washes in FACS buffer and data acquisition on an LSRII Fortessa (BD Biosciences). For intracellular staining, cells were stimulated with lipopolysaccharide (1 $\mu\text{g}/\text{ml}$) for 4 h in the presence of brefeldin A (10 $\mu\text{g}/\text{ml}$), followed by extracellular antigen staining and intracellular fixation and permeabilization as directed (Affymetrix/eBioscience, San Diego). Flow data were analyzed using Flowjo Version 10.0 (Tree Star, Ashland OR).

Statistics

GraphPad Prism 7 (GraphPad Software, Inc., La Jolla, CA) and Stata SE (StataCorp LP, College Station, TX) were used for all statistical tests. Relative PDH activity was compared between groups with analysis of variance (ANOVA) and Brown-Forsythe tests to check for unequal variance, and each group was compared with control liver with two-sided *t* tests with Welch's correction due to heteroskedasticity and Bonferroni correction to control for false discovery (adjusted $\alpha = 0.008\bar{3}$) (Fig. 1D). ATP content was compared among groups with ANOVA, Brown-Forsythe tests, and Tukey's multiple comparisons two-sided *t* tests (Fig. 1E). Relative mitochondrial DNA content was compared with ANOVA, Brown-Forsythe, and Tukey's multiple comparisons two-sided *t* tests (Fig. 2A). OCR for complex I and II were compared between each experimental group and control liver using non-parametric Mann-Whitney U tests, as data were not normally distributed (Fig. 2B). BNGE *in situ* enzymatic assay activity was compared between experimental groups with ANOVA, Brown-Forsythe tests, and Dunnett's multiple comparisons two-sided *t* tests. Complex II SDH activity was log-transformed to satisfy the equal-variance requirement of the ANOVA test; after log transformation, the Brown-Forsythe *p* value was 0.0681 (Fig. 2C). FAO was compared between experimental groups with ANOVA, Brown-Forsythe, and Tukey's multiple comparisons two-sided *t* tests (Fig. 2D). Total immune cell population differences were compared with a two-way ANOVA and Dunnett's multiple comparisons two-sided *t* tests (Fig. 4B). Percent IL-10 positivity was compared with an *F*-test to ensure homogeneity of variance and a two-sided *t* test assuming homoskedasticity (Fig. 4C). Statistically significant differences for RNA-seq data were defined as FDR-adjusted *q*-value of < 0.05 , as provided by the DESeq software package and as described previously (Fig. 7) (28, 29). S-phase content was compared between liver and tumors with ANOVA, Brown-Forsythe, and Tukey's multiple comparisons two-sided *t* tests (supplemental Fig. S1). Expression differences in the top and bottom 100 transcripts between initial and recurrent tumors were compared with non-parametric Mann-Whitney U tests, as expression levels for these transcripts were not normally distributed (supplemental Fig. S6). Degrees of dysregulation (absolute value of fold-change in tumors relative to normal liver) between *Myc*-amplified and non-*Myc*-amplified human HCCs were compared with Wilcoxon signed-rank tests, as data were not normally distributed. All error bars in figures represent mean \pm S.E.

Study approval

All animal studies were reviewed and approved by the University of Pittsburgh Institutional Animal Care and Use Committee, IACUC No. 14104886.

Author contributions—J. M. D., H. W., S. K., L. J., J. L., S. B., and K. B. performed, analyzed, and interpreted experiments. J. M. D. developed software and analyzed data sets. E. S. G. and C. A. B. analyzed and interpreted data. S. R. performed histopathologic examinations. E. V. P. devised experiments, supervised and coordinated research, and analyzed and interpreted data. E. V. P. and J. M. D. wrote the paper.

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