Profiling status epilepticus-induced changes in hippocampal RNA expression using high-throughput RNA sequencing

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Status epilepticus (SE) is a life-threatening condition that can give rise to a number of neurological disorders, including learning deficits, depression, and epilepsy. Many of the effects of SE appear to be mediated by alterations in gene expression. To gain deeper insight into how SE affects the transcriptome, we employed the pilocarpine SE model in mice and Illumina-based high-throughput sequencing to characterize alterations in gene expression from the induction of SE, to the development of spontaneous seizure activity. While some genes were upregulated over the entire course of the pathological progression, each of the three sequenced time points (12-hour, 10-days and 6-weeks post-SE) had a largely unique transcriptional profile. Hence, genes that regulate synaptic physiology and transcription were most prominently altered at 12-hours post-SE; at 10-days post-SE, marked changes in metabolic and homeostatic gene expression were detected; at 6-weeks, substantial changes in the expression of cell excitability and morphogenesis genes were detected. At the level of cell signaling, KEGG analysis revealed dynamic changes within the MAPK pathways, as well as in CREB-associated gene expression. Notably, the inducible expression of several noncoding transcripts was also detected. These findings offer potential new insights into the cellular events that shape SE-evoked pathology.

Status epilepticus (SE) is defined as persistent, unremitting seizure activity lasting 5–30 minutes1. This potentially lethal bout of sustained excitatory synaptic activity is caused by a variety of conditions, including, hemorrhage, stroke, viral infection and the withdrawal of anticonvulsive medication. In rodent models, pilocarpine-evoked SE is commonly utilized to profile neuroprotective and cell death signaling pathways, as well as potential cellular and molecular processes that underlie epileptogenesis. With respect to epileptogenesis, pilocarpine initiates a very well characterized step-wise pattern of pathophysiological changes that ultimately lead to the development of spontaneous seizure activity. Hence, the acute pilocarpine-induced SE event manifests at a behavioral level as tonic-clonic seizures that persist for many hours; within this time period, actuation of cell death signaling pathways and reactive gliosis are observed (reviewed in Turski et al., 1989). Subsequent to the SE phase, mice enter a ‘seizure silent’ period that persists for several weeks. During this period, alterations in synaptic organization are observed, including the well-characterized formation of recurrent granule cell collaterals2,3. Following this quiescent phase, animals enter a chronic epileptic-like state. At the cellular level, marked changes in neuronal excitability, cellular morphology, and synaptic reorganization are observed within the hippocampus. In addition, increased angiogenesis, granule cell dispersion and aberrant neurogenesis within the subgranular zone of the dentate gyrus are also observed3,4.

At a molecular level, SE triggers an array of changes in cell death, neuroprotective, and plasticity-associated signaling pathways. Along these lines, there is also a rapid rise in the generation of reactive oxygen species (ROS) levels, resulting in oxidative stress and damage to DNA, RNA, proteins, and lipids5. Likewise, SE stimulates the activation of apoptotic signaling pathways6. Parallelizing this rise in cell death signaling, SE also drives a robust neuroprotective response, including the expression of phase II detoxifying enzymes and antiapoptotic signaling cascades. Finally, consistent with neuroanatomical changes resulting from epileptogenesis, a number of studies have shown that there is an increase in expression of genes involved in axonal outgrowth and synapse formation7.
Many of these molecular events are likely resulting from altered transcriptional activity. Consistent with this, seizure activity is associated with an increase in SRF-, ARE-, and CREB-dependant transcription. Further, the activation state of upstream kinase pathways, including the p42/44 MAPK, p38(MAPK), JNK2/3, AKT, and PKC is increased at distinct phases of epileptogenesis, and in a cell-type-specific manner. Notably, in addition to changes in the activation state of kinase pathways, epileptogenesis has also been shown to alter kinase expression patterns.

To gain a clearer picture of the transcriptional events that could contribute to SE-evoked brain pathology, a number of studies have utilized high-throughput, discovery-based approaches, such as proteomics and large-scale microarray transcriptome profiling. The development of next-generation deep sequencing technologies provides a new approach to such profiling, potentially allowing for the discovery of novel genes, noncoding transcripts, and methylation changes in a high-throughput manner. Because of its relatively recent advent, the statistical tools available for RNA-seq analysis are still evolving, but the insights garnered from next-generation sequencing make it a compelling method for identifying novel biomarkers. Thus, through its ability to probe expression changes on a transcriptome-wide basis, RNA-seq approaches may yield additional therapeutic targets by highlighting novel genes and cellular signaling networks.

Here, we took advantage of these strategies to examine the differential expression and molecular regulation characterizing SE-evoked hippocampal pathology across epileptogenesis. We hypothesize that the discrete and tightly-regulated phases of the epileptogenic process are mediated by the regulation of transcriptionally-inducible gene expression patterns that subsequently give rise to the next phase of pathogenesis, ultimately manifesting themselves in spontaneous seizure activity. To this end, we performed whole transcriptome profiling to identify differentially expressed mRNAs at 12 hrs, 10 days and 6 weeks post-SE. These time points where selected to approximate the distinct phases of the epileptogenic process (acute, seizure silent, and spontaneous-seizure phases) as described above, and parallel the timepoints chosen for previous transcriptomic examinations of epileptogenesis. The data presented here identify distinct alterations in gene expression, and functional analysis provides potential clues regarding how these changes may contribute to the epileptogenic process.

Results
Altered gene expression across phases of epileptogenesis. To determine the transcriptional regulation that occurs over the course of epileptogenesis, we used Illumina-based high-throughput sequencing. Hippocampi were dissected 12 hours, 10 days, and 6 weeks after induction of SE, as well as from vehicle-treated controls (Fig. 1A, B). These time points were selected to profile each stage of the epileptogenic process and are consistent with timepoints used in previous studies (acute, seizure silent, and spontaneous-seizure phases). Concordantly, animals sacrificed during the acute timepoint exhibited robust excitotoxic cell death within the hippocampus (Fig. 1C), followed later by evidence of substantial mossy fiber synaptic reorganization within the dentate gyrus (Fig. 1D). Further, animals sacrificed at the 6-week time point exhibited spontaneous seizure activity, occurring every 10–20 minutes and lasting several seconds each. Libraries were sequenced to an average depth of ~3.4 million reads per group across the three timepoints (Supplemental Table 1). Because our cDNA libraries are 3’ biased, this depth was adequate to detect genes with low levels of transcription (1–10 copies per cell). Approximately 2.2 million reads mapped to unique locations on the M. musculus UCSC mm9 reference genome, with ~50% of these mapping to exons of known RefSeq genes (Supplemental Fig. 1).

To examine temporal regulation of gene expression, we performed an unsupervised clustering analysis of the 300 most significantly-regulated genes (Fig. 1E). Hierarchical clustering (hclust) showed distinct patterns with significant clusters of genes showing an increase at 12 hr and 10 days. The 6 week timepoint showed a mixed pattern of gene expression with most gene blocks showing a decrease relative to the 10 day time point. Cell death resulting from SE may contribute to reports of down regulated transcripts. Given the potential difficulty that cell death poses for the interpretation of the downregulated transcript set, our subsequent analysis focused more heavily on those transcripts that were upregulated after SE induction relative to untreated animals (downregulated genes, their fold changes, and subsequent analysis are included for reference in Supplemental Fig. 1 and Supplemental Fig. 4). Of the total list of upregulated genes, only 197 genes were upregulated in all three timepoints, whereas the majority of genes in each timepoint were unique to their respective group (Fig. 1F). Of note, data corresponding to the 12-hour, 10-day, and 6-week timepoints are represented by magenta, yellow, and cyan (respectively) throughout our report, with overlapping datasets denoted by the resultant mix of colors presented in Figure 1F.

The transcriptional patterns revealed by our analysis offer a good degree of overlap with previously reported datasets, as well as reveal new avenues for investigation. SE is known to trigger robust excitotoxicity and cell death throughout the hippocampus. In response, a range of neuroprotective and reparatory mechanisms are activated, which was reflected in our transcriptional analysis of epileptogenesis (Supplemental Fig. 1 and Supplemental Fig. 4N). For example, metallothionein, a Golgi-associated metal binding protein with neuroprotective effects after seizure insult, was upregulated at both the 12 hour (Fold Change [FC] = 2.54, q < 0.05) and 10 day (FC = 3.03, q < 0.05) timepoints. Similarly, upregulation of hypoxia response genes like Hmox2 during the seizure-silent phase (FC = 3.31, q < 0.05) contribute to recovery after excitotoxic insult. We also saw an increase in neurotrophic and growth factors (e.g., Bdnf: FC = 3.03, q < 0.05; Fgf1: FC = 3.31, q < 0.05) that play a protective role after induction of SE. Induced vascularization is highly associated with the post-SE response, and numerous genes linked with blood vessel development were upregulated across all three epileptogenic phases, including Vegf (FC = 4.64 [12 hours], 5.36 [10 days], 4.00 [6 weeks]), q < 0.05) the expression of which has a protective effect in the hippocampus after seizure activity. Furthermore, although cell death is widespread throughout pathogenesis, an upregulation of expression related to axonogenesis was also prevalent across all three timepoints (Supplemental Fig. 1E, largest changes presented and discussed in Fig. 4 below). Aberrant mossy fiber sprouting leads to recurrent excitatory circuits that exacerbate seizure activity. We found numerous genes associated with axonal growth that were upregulated across epileptogenesis. Some are known to be upregulated with seizure activity (e.g., Srf: FC = 5.74 [12 hours], 5.58 [10 days], 6.53 [6 weeks], q < 0.05; Cxcl12: FC = 15.77 [12 hours], 24.67 [10 days], 24.15 [6 weeks], q < 0.05)23,24, but several were yet to be described within the context of epileptogenesis up to this point (e.g., Dyps1: FC = 11.94 [12 hours], 15.50 [10 days], 14.95 [6 weeks], q < 0.05; Cyfip1: FC = 2.13 [12 hours], 2.42 [10 days], 2.48 [6 weeks], q < 0.05). Given the heterogeneous nature of cell-types within hippocampal tissue, interpretation of altered pathways should be undertaken with caution. Nevertheless, these genes sets may provide substantial insight when used in conjunctions with established literature on the signaling mechanisms underlying epileptogenesis.

The twenty unique genes with the greatest increases in expression at each timepoint are presented in Table 1, as well as the top twenty upregulated genes that were common to all three groups. Values reflect the average of statistically-significant fold changes across all replicates and splice-variant transcripts. Several genes, such as Cxcl12 and Notch3, had previously been reported to be involved in...
Figure 1 | Differential expression of mRNA across phases of epileptogenesis. (A) Timeline depicts tissue collection across the progression of epileptogenesis. Animals were sacked 12 hours, 10 days, and 6 weeks after pilocarpine induction of status epilepticus (SE). These timepoints correspond to the acute, seizure-silence and spontaneous-seizure phases, respectively, paralleling the time course of pathogeneses. (B) Schematic outline of processing and analysis procedures. Hippocampal tissue was collected from each time point, and RNA was isolated and cDNA libraries were created. Libraries were sequenced, and subsequent datasets were filtered and compared to controls to yield fold changes across experimental groups. Genes showing changes in regulation were examined for differential functionality using the Database for Annotation, Visualization, and Integrated Discovery (DAVID) v6.7. (C) At 2 days post-SE, Fluoro-Jade B labeling revealed cell death within CA1, CA3 subfields and the hilar region (Hil) of the hippocampus. GCL, granule cell layer. Scale bar: 200 μm. (D) Timm staining revealed reorganization of mossy fiber projections within the hippocampus 6 weeks after the induction of SE. Arrows indicate mossy fiber sprouting within the intermolecular layer. Mol, molecular layer; MF, mossy fibers. Scale bars: 200 μm (left); 100 μm (right). (E) Hierarchical cluster analysis of differentially expressed genes between control, 12-hour, 10-day, and 6-week samples (q < 0.05). Increases in expression are indicated in red hues and decreases are in green hues. (F) Venn diagram showing the number of genes with significant upregulation (q < 0.05) in each time point. Color schemes representing each time point are used in all subsequent figures.
seizure-related neuronal activity and SE-affected tissue. Again, several others had not been previously associated directly with seizure activity. These genes represent a wide array of functional categories, including transcriptional regulation (e.g., Aif7), regulation of cell survival (e.g., Cflar, Mul1), and neuronal outgrowth (e.g., Ltbp3, Dnaja3).  

Pathogenic modification of molecular and cellular signaling. To investigate the functional effects of gene induction at each timepoint, we examined Gene Ontology (GO) terms across three domains: molecular functions, cellular components, and biological processes (see Supplemental Fig. 2 for a glossary of terminology). In addition, we examined gene enrichment within Kyoto Encyclopedia of Genes and Genomes (KEGG) and BioCarta pathways. As detailed in the Methods section, we created an array of possible ontologies for five functional domains of interest (‘Molecular Functions’, ‘Cellular Components’, ‘Biological Processes’, and ‘Signaling Pathways’), with each ontology represented as cubed within the array, to produce a colorimetric profile of the gene functionalities upregulated at each timepoint. In addition to the static overview presented in Figure 2, Supplemental Fig. 3 offers an interactive digital version, which allows for dynamic exploration of each category. Each square can be zoomed in to view its ontological term, as well as the names of all the genes upregulated in that category at each timepoint. The total number of upregulated genes at each timepoint is also provided. The greatest diversity of changes appeared at the 6-week timepoint, whereas dysregulated genes at the earlier two timepoints showed more focused functionality. Interestingly, the greatest overlap in functional categories appeared between the 12-hour and 6-week timepoints. This pattern is reflective of the temporally distinct pathological changes that occur across epileptogenesis, and highlights the unique nature of the molecular signaling events within the quiescent period compared with the hyperexcitability that occurs during the acute and chronic phases. The ontological categories that correspond to each cube, as well as the associated genes from each timepoint and the statistical strength of each category are available and searchable in Supplemental Fig. 4.

As part of our functional analysis, we also identified the ontologies of the genes uniquely expressed at each timepoint. Mutually-exclusive lists of genes were used to produce clusters of enriched categories for each timepoint. The top 20 clusters of upregulated categories are presented for each timepoint in order of DAVID Enrichment Score, along with the number of genes within each cluster (Fig. 3). The overlaid line graphs depict the –log(EASE), which indicates the strength of gene associations within each cluster. Of note, the acute timepoint (12 hours) was enriched for genes associated with synaptic activity (e.g., ‘Synapse’, ‘Synaptic Transmission,’ and ‘Protein Kinase Regulator Activity’), morphological modifications (e.g., ‘Neuron Projection,’ and ‘Microtubule’), and transcriptional regulation (e.g., ‘Chromatin Modification,’ ‘Transcription Factor Binding,’ ‘mRNA Processing’, ‘Methyllation’). In contrast, the 10-hour timepoint showed heightened metabolic regulation (e.g., ‘Mitochondrion’, ‘Oxidative Phosphorylation’), and homeostatic damage response signaling (e.g., ‘Response to Oxidative Stress’, ‘Negative Regulation of Apoptosis’, ‘Cell Redox Homeostasis’). By six weeks, the transcriptional responses appear to have reached a new equilibrium reflective of adaptive cellular function that show fewer significant cluster differences compared to control (Supplemental Fig. 4). However there remained evidence of heightened excitability and morphogenesis reflected in the transcriptome at the chronic phase (e.g. ‘Synapse,’ ‘Regulation of Cytoskeleton Organization,’ ‘Neuron Projection Morphogenesis,’). Once again, these timepoint-specific gene lists yielded greater functional similarity between the 12-hour and 6-week timepoints than with those from the 10-day timepoint. This pattern is likely reflective of the hyperexcitability observed at both acute and chronic epileptogenic phases that is absent during the quiescent seizure-silent phase. A detailed examination of the genes associated with these enriched ontologies may yield greater insight into the signals that drive the pathological progression along each phase of epileptogenesis.

Table 1 | Genes showing strongest upregulation in each phase of epileptogenesis and in all three phases (left 3 gene columns) The 20 genes with the largest increases in expression that were unique to each time point. (right gene column) The 20 genes common to all three time points with the largest increases in expression relative to control tissue. (FC = Fold Change; q = q-value)

| Genes | FC; q | Genes | FC; q | Genes | FC; q | Genes | FC; q |
|-------|-------|-------|-------|-------|-------|-------|-------|
| Nt5c3l | 240; 0.020 | Pdlim2 | 367; <0.001 | Pgtt1b | 269; <0.001 | Ralgps1 | 38; <0.001 |
| Dusp22 | 185; <0.001 | Lbp | 234; 0.030 | Spag6 | 221; <0.001 | Afs311 | 70; <0.001 |
| Paqr9 | 161; <0.001 | Lgals9 | 193; <0.001 | Naip2 | 190; <0.001 | Elf1 | 145; <0.001 |
| H2-Ke2 | 157; <0.001 | 2310009 | 1B15Rik | 152 | <0.001 | Gpmb | 179; <0.001 |
| Iftp27 | 155; 0.005 | Echdc2 | 166; <0.001 | BoC067068 | 141; <0.001 | Tmem186 | 164; <0.001 |
| Dusp14 | 151; <0.001 | Cd72 | 153; <0.001 | Rsad1 | 140; 0.003 | Notch3 | 23; 0.001 |
| Wdr46 | 121; <0.001 | Prrg2 | 153; <0.001 | Dysf | 138; <0.001 | Atf7 | 46; 0.002 |
| 281003 | 119; <0.001 | Pigw | 152; <0.001 | Sco1 | 135; <0.001 | Mul1 | 63; 0.001 |
| Grin2d | 110; <0.001 | Cfbf | 151; 0.001 | Hpgds | 127; 0.004 | Cflar | 47; <0.001 |
| Fam184b | 109; <0.001 | Adpgk | 150; <0.001 | H110ra | 127; 0.010 | Fhl1 | 68; <0.001 |
| Tek5 | 105; 0.040 | Rnps1 | 147; <0.001 | Samd10 | 127; <0.001 | Zfp143 | 70; 0.002 |
| Spaq1 | 103; <0.001 | Mgs1 | 144; >0.001 | Ikbip | 111; <0.001 | Gramd3 | 17; 0.010 |
| Yars2 | 103; <0.001 | Sat1 | 142; 0.030 | Cxcr4 | 110; <0.030 | Fam20a | 24; 0.001 |
| Prss36 | 101; <0.001 | Dolk | 137; >0.001 | Spint1 | 111; <0.001 | Lass2 | 10; <0.001 |
| Glt1d1 | 101; 0.030 | S100A13 | 135; <0.001 | Gulp1 | 108; <0.001 | C4b | 20; 0.001 |
| Wdr60 | 98; 0.003 | Slc16a9 | 132; <0.001 | Fgl2 | 104; <0.001 | Dnaja3 | 65; 0.003 |
| 1110018 | 97; 0.002 | Gdpd5 | 130; 0.020 | Esam | 103; <0.001 | Fam173a | 89; 0.003 |
| J18Rik | 96; <0.001 | 1190002 | H23Rik | 124; <0.001 | Trhr2 | 100; <0.001 | Asb7 | 39; 0.041 |
| S1C7a3 | 93; 0.010 | Med30 | 124; <0.001 | Sell13 | 99; <0.001 | Ltbp3 | 5; 0.003 |
| Larp6 | 93; 0.004 | Crybb1 | 122; <0.001 | Nol12 | 96; <0.001 | Syl3 | 14; <0.001 |

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In addition to a functional examination of those transcripts unique to each phase of pathogenesis, we probed the subset of genes that were upregulated across all three timepoints for functional patterns and commonalities (Supplemental Fig. 1E). Because of the limited number of genes that were upregulated, this analysis yielded significantly fewer enriched ontological categories. Nevertheless, among the resulting clusters ‘blood vessel development’ and ‘axonogenesis’ were both highly enriched. Given the massive cellular reorganization and vascularization that occurs during epileptogenesis, it is not surprising that these clusters would be strongly represented. Here, we present the relative fold changes across time for five angiogenesis cluster-associated genes upregulated across experimental groups (Fig. 4). Of note is the particularly robust enrichment of Cxcl12 (FC = 15.77 [12 hours], 24.67 [10 days], 24.15 [6 weeks], q < 0.05), a chemokine known to play a significant role in angiogenesis and progenitor cell migration acutely after seizure activity, though its role in chronic SE has yet to be fully examined.

Among the signaling pathways identified from lists of upregulated genes, enrichment within the MAPK cascade (57 genes out of 118 in the KEGG representation of the JNK, p38, and p42/44 ERK1/2...
Figure 3 | Enriched ontological clusters across epileptogenesis. The top 20 enriched clusters of DAVID categories are reported for the 12-hour, 10-day, and 6-week timepoints. Clusters were sorted by enrichment score and the number of upregulated genes within each cluster is reported as bars corresponding to the lower axis. In addition, EASE Scores (DAVID’s modified Fischer Exact P-value) for each cluster are depicted as a measure of the association strength of genes within each cluster with a line graph corresponding to the upper axis.
pathways) was of particular interest. Activation of these pathways is known to mediate many of the cellular responses to seizure activity in a time-dependent manner. Here, we characterized changes in expression across the entirety of the JNK, p38, and classical MAPK signaling pathways at all three timepoints (Fig. 5). Accompanying previously demonstrated increases in kinase phosphorylation are robust changes in transcriptional regulation within all three arms of the MAPK cascade. Positive regulation of these pathways was prominent at all three timepoints (Fig. 5, inset). Upregulation of upstream activators of Ras suggest a facilitation of subsequent Raf, ERK and CREB phospho-activation, and all four of these signaling proteins are also upregulated transcriptionally. In addition, the induction of MAPK kinase (Mkk) 6 (FC = 42.29 [12 hours], 25.79 [6 weeks], q < 0.05) and Mkk3 (FC = 67.83 [12 hours], 67.44 [6 weeks], q < 0.05) at early and late timepoints is consistent with the idea that stress-mediated cellular responses persist throughout epileptogenesis and well into the chronic phase of SE. Activation of each of these pathways may be counterbalanced by the upregulation of inhibitors such as MKPs, and protein tyrosine phosphatases. Key inhibitors, such as Pp2ca (Fig. 5, circled), that did not show concordant upregulation may serve as potential therapeutic targets to counter hyper-activation of inflammatory and apoptotic signaling.

CREB mediates activity-dependent gene expression and is known to be phosphorylated following seizure activity. Given the increased expression of both CREb and its upstream activators across multiple timepoints, we determined the proportions of upregulated genes across epileptogenesis that are potentially under control of CREB transcriptional regulation. To accomplish this, we used gene lists generated from a genome-wide screen of CREB occupancy in hippocampal neurons. In this approach, chromatin immunoprecipitation (ChIP) high throughput sequencing was performed after CREB pull down, and the resulting list of enriched genes was intersected with our datasets of upregulated transcripts from each timepoint. Thus, we generated outputs representing the subset of genes that are both CREB-associated and upregulated at each phase of epileptogenesis (1474 genes total; Fig. 6A, Supplemental Fig. 5). Altogether, 37% of upregulated genes were associated with CREB binding at each timepoint. Among these was found an enrichment of genes associated with neuronal projection and synaptic transmission during the acute phase that gave way to oxidative stress response, myelination, and homeostatic regulation in the second epileptogenic phase (Supplemental Fig. 5). The chronic phase was characterized by genes regulating cytoskeletal reorganization, and synapse formation, along with increased transcriptional regulation. Hence, these CREB-associated genes likely reflect the robust reorganization that occurs during epileptogenesis, and provide insights into key molecular mediators of these processes.

Given that our RNA seq data strongly suggest that SE drives a lasting increase in CREB-mediated gene expression, we sought to profile the activation of the CRE pathway at multiple time points following SE. To this end, we utilized a well-described transgenic mouse model in which 6 tandem CRE motifs drive the expression of a β-galactosidase reporter gene. Consistent with our prior work, we found that SE triggers a marked upregulation of CREB-dependent transcription in the dentate GCL at 4 hrs post SE onset (Fig. 6B). Double labeling with the neuronal-specific marker NeuN and the interneuron marker parvalbumin A revealed that CRE-transcriptional activity occurred in both excitatory glutamatergic neurons and inhibitory GABAergic neurons (Fig. 6C). Low level β-galactosidase expression was also detected in astrocytes (as assessed via GFAP double labeling). Of note, at the 2 day post-SE time point, the most pronounced labeling was observed in astrocytes and microglia (as assessed via CD11b double labeling), within the molecular layer and hilar region; limited labeling was observed in neurons of the GCL. Quantitative analysis of SE-evoked CRE-reporter expression within the hippocampus can be found in Lee et al, 2007. Interestingly, in mice that developed spontaneous seizure activity (6 week post-SE), marked CRE transgene expression was detected within the GCL (Fig. 6D). Together, the data are consistent with the RNA seq data; thus, 1) SE triggered a rapid and prolonged increase in CRE mediated gene expression, and 2) robust expression was detected within a multitude of cell types.

Identification of a predicted noncoding transcripts upregulated after induction of SE. An advantage of unbiased high-throughput sequencing relative to more traditional profiling approaches is its capacity to detect unannotated non-protein-coding transcripts. To this end, we identified several predicted noncoding regions that showed transcriptional upregulation after the induction of SE. A sliding window approach was used to find clusters of repeat-masked sequence reads that did not overlap with annotated protein-coding exons. Sequence tags within these regions were counted and pair-wise comparisons were conducted. Here, we highlight two examples of predicted noncoding regions that exhibited transcriptional activity (Storey Q-test) and that did not have homology with RNA or DNA repeats (Fig. 7). First, we show upregulation of a long noncoding RNA (IncRNA) whose involvement in epileptic pathophysiology has been previously established. MALAT1 regulates synaptic density in neurons and was shown to be upregulated in high-activity areas of the human epileptic cortex. We also show its increased expression across multiple phases of epileptogenesis (Fig. 7A). In addition to Malat1, we present a predicted noncoding transcript on chromosome 9 that was previously unassociated with epileptic activity. This short sequence showed robust upregulation predominantly 6 weeks after induction of SE (Fig. 7B). The predicted noncoding transcript is contained within a small EST that has only been detected in pre-implantation embryo cDNA libraries. No exonic sequences from this EST were detected in any of our hippocampal RNA-Seq data sets (unpublished observation). Thus, we are confident that the small transcript reported in Figure 7B is not associated with the EST sequence. These transcripts are representative examples of a wide array of noncoding transcriptional activity that occurs across the phases of epileptogenesis. The increased transcription of such predicted noncoding regions supports the possibility of a regulatory role for noncoding signaling over the course of epileptogenesis.

Discussion
Here, we used high-throughput deep sequencing to profile the hippocampal transcriptome across three phases of SE-evoked epileptogenesis. Comparative analysis of samples derived from each timepoint revealed significant differences in transcriptomic profiles,
reflecting temporal alterations underlying the progression of epileptic pathology. In addition, we performed functional analysis to determine the cellular processes associated with altered mRNA expression across epileptogenesis. Several patterns emerged from an ontological analysis of these datasets. Although some functional overlap is apparent, the induced functional profile at each timepoint was largely unique. Together, these data provided insight into both the temporally divergent and sustained cellular responses to pilocarpine-induced SE. Several of the cellular and molecular processes mentioned above are under the regulatory control of the MAPK-family signaling pathways. We demonstrate that the transcriptional expression of these signaling cascades is markedly altered at multiple phases of the epileptogenic process. The p42/44 MAPK cascade plays a prominent role in progenitor cell proliferation and neuronal survival following SE.\(^{38-40}\) Given this, some of the upregulation of these signaling cascades may be due to the increase in neurogenesis known to occur following SE, presenting a potential confound to our data by biasing our results toward this population of cells. Nevertheless, the density of these developing neurons is relatively low, and likely only accounts for a small subset of the transcriptional increases observed. Thus, given the robust activation of p42/44 MAPK signaling across epileptogenesis, an increase in CREB-dependant gene expression is not unexpected. Indeed, CREB-mediated gene expression is associated with SE in a cell-layer and time-dependant manner.\(^{41,42}\) Concordantly, we showed that CREB-bound genes were highly enriched across timepoints, as demonstrated by robust intersectionality with CREB ChiP-seq gene lists. This subset of CREB-associated genes yields insight into the protective mechanisms of CREB phosphorylation.

Conversely, the JNK and p38 MAPK pathways are known to contribute to the apoptotic and inflammatory responses following seizure activity.\(^{43,44}\) Within these pathways, genes have emerged out of...
our non-biased interrogation that may not have been emphasized \textit{a priori} as having therapeutic potential. The serine/threonine protein phosphatase PP2CA is one such example. Research surrounding the regulatory potential of PP2CA in epilepsy is limited, though initial studies suggest that PP2CA/CACNB4 signaling is essential to activity-dependent gene expression and is impaired in juvenile myoclonic epilepsy\textsuperscript{33}. PP2CA is a known regulator of cellular stress response, attenuating excitotoxic JNK/p38 signaling after ischemic injury\textsuperscript{45,46}. Given the absence of compensatory upregulation of PP2CA in response to JNK/p38 activation after SE, therapeutic strategies to increase PP2CA activity may prove protective against epileptic pathology as well. Nevertheless, the scale of cascade-wide transcriptional dysregulation revealed by high-throughput technology suggests that therapies focused on a single target may be insufficient for meaningful protective benefit.

Figure 6 | Temporal and cell-type profile of SE-evoked, CREB-mediated, gene expression. (A) Data from a genome-wide analysis of CREB occupancy by ChIP-Seq was used to determine the proportion of upregulated epileptogenic genes that are CREB-associated. Venn diagram depicts subsets of genes at each timepoint. (B) Representative immunohistochemical labeling of tissue from CRE-\(\beta\)-Gal transgenic mice reveals robust reporter gene expression within the GCL at 4 hrs post-SE onset. Mol: molecular cell layer; GCL: granular cell layer; Hil: hilus. Bar: 50 microns. (C) Double immunofluorescence labeling for \(\beta\)-Gal and NeuN or parvalbumin A (Parv A) revealed that the reporter gene was inducibly expressed in both excitatory and inhibitory neurons at the 4 hour time point. Limited transgene expression was also observed in astrocytes, as assessed via GFAP double labeling. At 2 days post-SE (right panels), limited \(\beta\)-gal was detected in GCL neurons; rather, the reporter was detected in astrocytes and microglia, as assessed via GFAP and CD11b double-labeling. (D) Representative data from animals that were rendered epileptic (sacrificed at 6 weeks post-SE) revealed marked CRE-mediated gene expression within the GCL.

Figure 7 | Upregulation of predicted noncoding RNAs after induction of SE. Representative predicted noncoding regions showing increased transcription across epileptogenesis. The UCSC Genome Browser was used to visualize tag counts from each timepoint. (A) Increased expression of the MALAT1 long noncoding RNA. (B) An intergenic transcript showing increased expression during the chronic phase of epileptogenesis.
epileptogenesis, others have previously made use of microarray technology to profile the epileptogenic transcriptome\(^{4,6,15}\). These studies all exhibited the greatest changes in gene expression at the earliest timepoints, which is to be expected given the profoundly heightened neuronal excitability that occurs during the acute phase. Here, we too found the greatest number of genes expressed at the acute timepoint (using similar significance and fold-change criterion). Though there remains some level of variability between datasets due to both the methods of initial approach and subsequent analysis used in each, there remains remarkable overlap between studies that lends confidence to our new findings and the new datasets provided here that expand upon previous work. Consistent among each of these studies is the robust expression of morphogenesis, blood vessel development, and stress-response signaling pathways throughout epileptogenesis\(^{22,14,47}\). Further, the MAPK family signaling cascades were a prominent feature of multiple studies, and PI3K, TGF-beta, IGF-1, mTOR signaling were all frequent regulatory themes\(^{22,14}\).

Though high-throughput screening of transcription across multiple phases of epileptogenesis is problematic in the context of human studies, microarray assays of postmortem epileptic brain tissue have proven useful in characterizing the chronically-epileptic transcriptome\(^{49}\). Arion et al. (2006) demonstrated that transcriptional regulation (including of the MAPK signaling cascade) correlated with intracranial EEG activity measured immediately prior to partial lobectomies performed in patients with temporal lobe epilepsy. Additionally, transcriptional expression within the CA3 subfield of the epileptic hippocampus was shown to correlate with patients’ initial precipitating injury\(^{49}\). That the transcriptome varied substantially between patients with differing precipitating injuries emphasizes the importance of the method of SE induction used in animal models, and the need to validate transcriptional changes found in rodent models against those observed in human patients\(^{49}\).

Though still subject to false positives, RNA-Seq has been demonstrated to be a reliable method of transcriptional analysis\(^{51,52}\). Nevertheless, several factors should be considered when interpreting the data presented here. As mentioned, the heightened neurogenesis following SE should be considered when interpreting the increases in gene transcription reported here. In addition, not only will cell death associated with excitotoxicity lead to some bias with regard to down regulated genes, the remaining spared cells (as well as those migrating into the region as part of the pathogenic response) will also create some level of expression bias in our reports of upregulated genes in a cell-type specific manner (relevant genes with the heightened susceptibility to these biases are noted in Supplemental Fig. 4N). Furthermore, the current study was performed on whole-hippocampus homogenates, resulting in a transcriptional readout from a heterogeneous cell populations, including pyramidal neurons, basket cells, granule cells, astrocytes, microglia, and oligodendrocytes. Beyond these cellular-level confounds, there is still a need for more advanced statistical tools that have a sophistication paralleling the high-throughput capacity of RNA-seq outputs. Further, some issues with regard to sample replication and normalization remain unresolved in the field.

Up to this point, RNA-Seq had only been employed to screen for specific, strain-dependent single-nucleotide polymorphisms (SNPs) across genetic animal models of epilepsy, as well as changes in transcription and methylation at the chronic phase\(^{53,54}\). By using RNA sequencing to profile the entire transcriptome across multiple phases of epileptogenesis, we were able to gain perspective on both the protein-coding and noncoding transcriptional changes that occur. While many of the functional gene associations reported here build upon previous studies of molecular signaling following induction of SE, we also reveal expression patterns that suggest new regulatory pathways for which there is untapped therapeutic potential. Collectively, the data presented here emphasize the need for a coordinated investigation into the treatment of repetitive seizure activity that takes into account both gene-by-gene and high-throughput approaches.

A final word regarding current methods of reporting next-generation sequencing studies: as the sheer volume of data output increases, better graphic representation will be essential for the effective interpretation and translation of results. To date, published figures are almost exclusively comprised of bar graph, Venn diagram, and heat-map depictions of data, which are limited in their utility for meaningful therapeutic discovery. Though the authors have made some attempts here to further the endeavor for novel methods of representation, there remains a need for network-level visualization graphics to a degree that has not yet been achieved. The ability to overlay datasets and interacting pathways in scalable, digital representations would greatly enhance the insights to be garnered from high-throughput investigations and aid in the discovery of therapeutic models that more closely approximate biological reality. By these means, a more complete picture of gene expression and molecular signaling networks will emerge.

**Methods**

**Pilocarpine-induced SE.** Adult (~10 weeks old) C57BL/6J mice were intraperitoneally injected with atropine methyl nitrate (1 mg/kg), followed 30 minutes later with an injection of pilocarpine hydrochloride (325 mg/kg)\(^{49}\). Only those animals that survived and achieved SE were included in this study. SE was defined as an animal exhibiting multiple stage 5 seizures (tonic-clonic seizure extending to all four limbs, resulting in loss of balance and falling) and subsequent persistent and spontaneous seizure activity\(^{50}\). SE was induced in 20% of animals, with a mortality rate of 40%. Pilocarpine acts as an agonist for muscarinic receptors in the entorhinal cortex, and changes within the hippocampus are a result of downstream excitatory transmission, rather than any direct confounding effects of pilocarpine itself. Thus, given that subthreshold pilocarpine still drives synaptic activity at higher than normal levels, pilocarpine-injected mice that did not develop seizures were not included in our RNA-seq analysis. Instead, control animals were injected with atropine, followed by an equal volume of saline (0.9% Sodium Chloride; B. Braun Medical Inc.) 30 minutes later. No systemic, adverse side-effects of pilocarpine administration, beyond those expected by the induction of SE, were observed in treated mice. The Ohio State University Lab Animal Use Committee approved all animal procedures and all experiments were conducted in accordance with the approved guidelines.

**Fluoro-Jade B.** Fluoro-Jade B staining was performed as previously described in Choi et al. (2007)\(^{50}\).

**Timm staining.** Timm staining was performed as previously described in Li (2010)\(^{50}\).

**RNA isolation and library preparation.** Twelve hours, ten days, or six weeks after induction of SE, animals were sacrificed by cervical dislocation and hippocampi were removed bilaterally and subjected to total RNA extraction by TRizol (Invitrogen Corp.) as per manufacturer instructions. Polyadenylated mRNA was purified by Dynabeads\textregistered m RNA purification kit (Life Technologies #61006). As per manufacturer protocol, 1 μg of total RNA was incubated at 50°C for 2 min and placed on ice. The sample was added to Dynabeads\textregistered with 50 μl binding buffer, mixed thoroughly at room temperature for 5 min, washed twice, and then eluted into 10 μl 10 mM Tris- HCl (pH 7.5) by heating to 70°C for 2 min. Purified mRNA was treated with DNase I (Life Technologies #18068015) at 37°C for 30 min followed by 70°C for 5 min. For first strand synthesis, 10 μl 0.1 M DTT, and 1 μl 10 mM DNTP mix were added, Samples were heated to 65°C for 5 minutes and incubated on ice. Four microliters of 5× First Strand Buffer, 2 μl of 0.1 M DTT, and 1 μl of RNaseOUT Recombinant RNase Inhibitor (Life Technologies #10777-019) were added and incubated at 37°C for 2 min. One microliter of MMLV reverse transcriptase (Life Technologies #28205) was added to samples, which were then incubated at 37°C for 50 minutes, followed by a five-minute inactivation at 70°C. cDNA was then treated with NEBNEXT mRNA second strand synthesis mix (NEB #6112) and incubated for 2.5 hours at 16°C, followed by sonication for 10 seconds. Samples were purified on a PCR column (Fermentas #K0702) and eluted in 22 μl 10 mM Tris pH 8.5. Blunt ends were using the DNA Terminator\textregistered End Repair Kit (Lucigen #40035). Samples were purified using AMPure xp beads (Beckman #A63880) and A-tailing of 3′ ends of eluted cDNA was performed using Exo-minus Klenow DNA polymerase (Epicentre #KLI080259) at 37°C for 30 min. Uniquely tagged Illumina genomic adaptors were ligated by Enzymatics T4 DNA ligase (Invitrogen #15224) at RT for 2 hrs. Size visualization was performed using 2% agarose gel. yielding cDNA libraries ranging in size from 200–250 bp, which were amplified by PCR (5 sec at 98°C; 10 sec at 72°C; 20–20 cycles) using Phusion High-Fidelity DNA Polymerase (ThermoScientific #F530).

**RNA-Seq and bioinformatic analyses.** cDNA from two animals was pooled into two independent biological replicates for each timepoint (ie. two sets of two animals per experimental group: control, 12 hours, 10 days, 6 weeks). Samples were sequenced...
using a Genome Analyzer II (GAII) at a concentration of 10pM in each lane. Base-calling was conducted with the standard Illumina Analysis Pipeline 1.0 (Firecrest-Bustard). Eight FASTQ sequence files (sequencing reads plus quality information) were generated and mapped to the mouse genome (UCSC mm9) using the Bowtie algorithm with default settings. A C + r program was used to count the number of uniquely mapped reads within exons of Ref-seq genes (UCSC Genome Browser mm9 annotation). The sequence data have been submitted to the NCBI Short Read Archive with accession number GSE78307. Ref-seq gene expression counts were median-normalized and overlapping gene IDs were discarded. Data was visualized using the UCSC Genome browser (NCBI37/mm9). All statistical analyses were performed in the R programming environment. RNA-Seq tag counts in Ref-seq gene regions were scaled to a weighted trimmed mean that excluded the top 5% of genes. Genes with a normalized read per million of less than 10 were excluded for differential expression analysis. Quality control analyses showed that less than 1% of exonic tags were common repetitive artifacts (e.g. satellite repeats and RNA repeats, Repeat masker4). Statistical comparisons were conducted using the likelihood-ratio statistic56 and the Storey Q-test was used to correct for multiple comparisons (Bioconductor qvalue package). Differential gene expression analysis was performed using the R/Bioconductor programming environment57. Statistical enrichment analysis was conducted using a 500 bp sliding window approach (see Lesiak et al. for details)35.

Functional analysis of gene lists

Using the web-based annotation tool DAVID v6.7 (Database for Annotation, Visualization and Integrated Discovery) to cluster genes overexpressed in each timepoint by their common functionality, enriched ontologies were identified and statistically clustered by comparison of their constituent genes57. The sequence data have been submitted to the NCBI Short Read Archive. Quality control analyses showed that less than 1% of exonic tags were common repetitive artifacts (e.g. satellite repeats and RNA repeats, Repeat masker4). Statistical comparisons were conducted using the likelihood-ratio statistic56 and the Storey Q-test was used to correct for multiple comparisons (Bioconductor qvalue package). Differential gene expression analysis was performed using the R/Bioconductor programming environment57. The sequence data have been submitted to the NCBI Short Read Archive. Quality control analyses showed that less than 1% of exonic tags were common repetitive artifacts (e.g. satellite repeats and RNA repeats, Repeat masker4). Statistical comparisons were conducted using the likelihood-ratio statistic56 and the Storey Q-test was used to correct for multiple comparisons (Bioconductor qvalue package). Differential gene expression analysis was performed using the R/Bioconductor programming environment57. The sequence data have been submitted to the NCBI Short Read Archive. Quality control analyses showed that less than 1% of exonic tags were common repetitive artifacts (e.g. satellite repeats and RNA repeats, Repeat masker4). Statistical comparisons were conducted using the likelihood-ratio statistic56 and the Storey Q-test was used to correct for multiple comparisons (Bioconductor qvalue package). Differential gene expression analysis was performed using the R/Bioconductor programming environment57.
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