Deep Multiomics Profiling of Brain Tumors Identifies Signaling Networks Downstream of Cancer Driver

Genes

Wang et al



Supplementary Figure 1 | MS-based quantification is accurate

MS measurements of phosphorylation events are highly consistent with previous immunoblot assays

comparing PDGFRA-driven HGGs to normal cortex (Paugh BS et al. 2013).





Supplementary Figure 2 | MS-based proteomic analyses specify particularly small variations

(a) Pairwise standard deviation matrix of whole proteome and phosphoproteome displays particularly small variations between replicates.

(b) Examples show distributions of representative Log₂ level whole proteome and phosphoproteome variations between *PDGFRA*-driven HGG biological replicates and between two HGG tumors.



Supplementary Figure 3 | HGG mouse models are reproducible

(a, b) Comparisons of RNAseq results on batch2 and batch1 HGG mouse models show highly consistent transcript abundance from tumors generated by independent experiments. Scatterplots show FPKM values of genes in batch1 and batch2 HGG mouse models, Pearson correlations were performed and R² values were shown on the upper left side of the panels.

(c, d) Mouse tumors express levels of mutated PDGFRA or NTRK fusion genes that are relevant to the levels expressed in primary human tumors. Normalized RNAseq data was used to compare the expression of the human NTRK1 (c) or PDGFRA (d) expression in mouse HGGs, control mouse cortex, or human HGGs from Wu et al, Nat Genet, 2014, categorized as HGGs carrying amplified wild-type PDGFRA, mutated PDGFRA, amplified and mutated PDGFRA, TPM3-TRK1 fusion gene, fusion genes involving NTRK2 or 3, or HGGs without mutation in PDGFRA or NTRK genes. Expression of mouse orthologs for NTRK1 (c) or PDGFRA (d) are shown in red. Analysis of variant allele frequency from whole genome sequencing data showed that human tumors with TPM3-NTRK1 fusion gene carried the fusion gene as a subclonal population comprising only 9-20% of the tumor cells. Therefore, expression of the TPM3-NTRK1 from bulk tumor RNA is a significant underestimate of expression of the TPM3-NTRK1 fusion in human tumor cells. Boxplot center line, median; box limits, upper and lower quartiles; whiskers, 1.5x interquartile range; points, outliers.



Supplementary Figure 4 | Proteome-transcriptome comparison shows deep proteomic coverage

(a) High percentage (84%) of expressed transcripts detected by MS. In the histogram of transcript log₂FPKM, MS-detected proteins are highlighted in grey. To ensure data stringency, FPKM >1 were applied in this analysis.

(b) The number of peptides identified for each protein in this MS analysis. > 96% of proteins were identified by at least 2 distinct peptides.

(c) The MS coverage of theoretically observable amino acid sequences in a histogram. In average, 42% of observable amino acid sequences were detected in this MS study.



and compare NTRK to PDGFRA



and compare NTRK to PDGFRA

Supplementary Figure 5 | Proteomic analyses detect a global trend of increased expression and phosphorylation of regulatory protein families in HGG tumors

(a - h) Deep proteomic data analyses show a global increase of protein expression and phosphorylation of most of regulatory protein families (Kinase, epigenetic genes, transcription factors and cancer genes) in HGG tumors compare to cortex, with higher magnitude of increase in *NTRK*-driven HGG than *PDGFRA*driven HGG. Scatter-histogram graphs of regulatory family proteins expression and phosphorylation comparing both HGG tumors to cortex and *NTRK*-driven HGGs to *PDGFRA*-driven HGGs. B.H. adjusted Student T test *p* values of 0.05 in both pairwise comparisons plus fold change (FC) cutoff of FC_{(NTRK/PDGFRA}) * FC_(HGG/cortex) > 1.5² for proteome and >2 for phosphoproteome are applied for differential expression analyses. DE genes are shown in red. Magnitude of change is represented by dot size. Top five altered proteins and phosphoproteins with the largest magnitude of alterations are labeled. The distributions of pairwise differences are shown by histograms.



Supplementary Figure 6 | AGC, CAMK and CMGC kinase superfamilies display higher activity in both HGG tumors compare to cortex

(a) Kinome tree circular map shows kinase activity comparing *NTRK*-driven HGG to cortex. Pairwise comparisons of kinase superfamilies with a Chi-square *P* values < 0.05 are labelled. Magnitude of kinase activity difference is represented by length of bars located outside of the kinome tree circles. Blue bar indicates higher activity in *NTRK*-driven HGG compare to Cortex. Green bar indicates higher activity in cortex compare to *NTRK*-driven HGG

(b) Kinome tree circular map shows kinase activity comparing *PDGFRA*-driven HGG to cortex. Pairwise comparisons of kinase superfamilies with a Chi-square *P* values < 0.05 are labelled. Magnitude of kinase activity difference is represented by length of bars located outside of the kinome tree circles. Red bar indicates higher activity in *PDGFRA*-driven HGG compare to Cortex. Green bar indicates higher activity in cortex compare to *PDGFRA*-driven HGG



Supplementary Figure 7 | Evaluation of AKT regulated substrates

(a) A high percentage of known AKT substrate sites were identified in the deep phosphoproteome. Bar graph shows reported substrate sites, quantified phosphosites and differentially phosphorylated sites.

(b) Differentially phosphorylated substrates that are regulated by AKT in HGGs. Scatter plot of log₂ level changes compare HGG tumors to cortex and compare *NTRK*-driven HGG to *PDGFRA*-driven HGG. Substrates with the same phosphorylation pattern as the AKT active site (S473, located at upper right) were accepted as AKT regulated substrates.

PRKAA1 active substrates



CDK5 active substrates



MAPK3 active substrates





ATM active substrates



PAK1 active substrates



FYN active substrates



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Supplementary Figure 8 | Heatmaps display differentially phosphorylated substrates (with up-regulated phosphorylation in HGG tumors) of other active kinases derived from kinase-substrate analysis. Color key presents the Z score of kinase substrates



Supplementary Figure 9 | Combination of mouse and human HGG data prioritizes putative cancer genes. (a) Overview of multiomics analysis across species

(b) Cross species integrative analysis reveal cancer driver responsive genes, many of which are wellreported cancer genes in HGG. The cancer driver responsive genes were classified according to their functions. Mouse transcripts and proteins/phosphorylations that follow the expression order of *NTRK* > *PDGFRA* > Cortex were accepted to be oncogene responsive. Human transcripts that show *NTRK* > *PDGFRA* were intersected with mouse oncogene responsive genes for prioritization. Gray fills indicate consistent changes identified, while white boxes indicate no significant change or data not available.

(c, d) Boxplots show Epha2 and CD74 expression in multiomics data in human and mouse. Human mRNA boxplots compare expression levels of EPHA2 or CD74 in pediatric HGGs with mutated PDGFRA or NTRK fusion genes. Boxplot center line, median; box limits, upper and lower quartiles; whiskers, 1.5x interquartile range.

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Quality controls and experimental optimization for transEDIT-dual CRISPR-Cas9 genomic screening

Primary HGG cells	 Culture medium and condition optimization
	2. Determination of cell morphology (floating or adherent)
	3. HGG genotyping
	4. Doubling time
	 Culture medium and condition optimization Determination of cell morphology (floating or adherent) HGG genotyping Doubling time Cas9 construct design Optimization of lentivirus transduction Selection curve for antibiotics Immunoblotting to confirm CAS9 expression gRNA construct design Optimization of lentivirus transduction Transduction medium and condition optimization ZeG fluorescence to confirm lentivirus transduction Real experiment with 3 repeats DNA extraction optimization Agarose gel to confirm amplicon size Sanger sequencing to confirm gRNA expression NGS deep sequencing
CA59	2. Optimization of lentivirus transduction
expression	3. Selection curve for antibiotics
	4. Immunoblotting to confirm CAS9 expression
	☐ 1. gRNA construct design
	2. Optimization of lentivirus transduction
gRNA expression	3. Transduction medium and condition optimization
	4. ZsG fluorescence to confirm lentivirus transduction
	5. Real experiment with 3 repeats
	6. DNA extraction optimization
	7. PCR optimization
	8. Agarose gel to confirm amplicon size
	9. Sanger sequencing to confirm gRNA expression
	10. NGS deep sequencing
	L 11. Data analysis



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Genomic functional screening using customized mini-gRNA library targeting cross-species conserved regulators



Supplementary Figure 10 | Quality controls for CRISPR-CAS9 functional genomic screening using transEDIT-dual system

(a) Work flow and QC steps for CRISPR-Cas9 functional genomic screening using transEDIT-dual system

(b) Simplified cartoon shows transEDIT-dual CRISPR construct with primer binding sites. Two gRNAs targeting different regions of the same genes were designed in a signal CRISPR construct. Primer binding sites for PCR amplification of gRNA 1 were shown. Fwd: forward; rev: Reverse.

(c) Image of primary HGG cells show mCherry fluorescence which co-express with *TPM3-NTRK1* cancer driver gene

(d) Stable expression of Cas9 in HGG cells as validated by immunoblot assay.

(e) gRNA integration was validated by fluorescence detection of ZsG. Image shows the detection of ZsG fluorescence in the primary HGG cells 3 days of transEDIT-dual gRNA lentivirus transduction.

(f) PCR products show amplicon size and band intensities corresponding to different PCR cycle. Theoretical amplicon size is 240bp. PCR cycle 29 was used for later analysis.

(g) Close to even distribution of gRNA counts before selection. Bar plot shows the Log2 level counts of each gRNA collected at 1 day after gRNA transduction. 3 x dual gRNAs were designed for each target gene, and the CRISPR-Cas9 experiment was repeated 3 times. Error bars indicate the s.e.m of 3 experiments.

(h) CRISPR-CAS9 genomic screening targeting top candidate targets derived through cross-species data integration. Bar pot shows the log2 level gRNA relative fold changes of top cross-species conserved regulators after selection for 15 days. Dash line indicates the mean level of all non-targeted gRNA controls. The gRNA counts of cross-species conserved regulators were compared with non-targeted gRNA controls, and red bars indicate gRNA changes that show statistical significance with a Student t test P value smaller than 0.05. Two asterisks: p < 0.01.

Fold change distribution and Z-value transformed distribution of MS2 and MS3 based peptide quantification comparing *NTRK-d*riven HGG to Cortex using TMT labeling.



Supplementary Figure 11 | TMT-based quantification using MS2 method has essentially no impact on protein differential expression analysis after Z-scale normalization

a. Quantitative ratio compression occurred in TMT labeling strategy using MS2 compare to MS3 method. MS3 strategy can essentially eliminate ratio compression with the cost of more duty cycles and the use of low resolution MS2 data for identification, which often compromise peptide/protein identification. Comparison between MS2 and MS3 methods on the same sample using TMT labeling shows smaller difference/variance measured in MS2 method compare to MS3 method, suggesting quantitative ratio compression in MS2 analysis of TMT labeling.

b. Z scale normalization essentially eliminates the effect of ratio compression on protein differential expression analysis. Z scale transformation of the same data shows almost exact same Z value distribution in MS2 and MS3 method, suggesting similar amount of DE proteins with the same Z value cutoffs for differential expression analysis comparing MS2 and MS3 methods.

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