

# FGFR signaling regulates resistance of head and neck cancer stem cells to cisplatin

## SUPPLEMENTARY MATERIALS

### RNA extraction

Collected cells from flow cytometry were centrifuged at 337 g for 5 min at 4°C. Excess volume was removed with a p1000 pipette and the remaining sample transferred to 1.5 mL microcentrifuge tube. Tubes were centrifuged at 300 g for 5 min at 4°C. Supernatant was removed with a p200 pipette until <30 µL sample remained. RNA was extracted using the Qiagen RNeasy Mini Kit following the manufacturer instructions. Briefly, 350 µL RLT was added to cell pellets (in <30 µL) and vortexed for 30 sec. 350 µL 70% EtOH was added to each sample and mixed until no viscosity remained. The sample was transferred to spin column, centrifuged 8000 g × 15 sec. Then 700 µL RW1 was added to spin column, centrifuged 8000 g × 15 sec. Then 500 µL RPE buffer to spin column, centrifuged 8000 g × 15 sec. Then 500 µL RPE buffer to spin column, centrifuged 8000 g × 2 min. Spin columns were placed on new collection tube and centrifuged 8000 g × 1 min. Spin columns were placed on 1.5 mL RNase-free centrifuge tubes. 15 or 30 µL of RNase-free water added to membrane and then centrifuged 8000 g × 1 min. 1 µL of sample was used to measure RNA concentration with a Nanodrop.

### Microarray

Microarrays and RNA processing were performed by the University of Michigan Microarray Core facility. The quality of total RNA for each sample was confirmed with a Bioanalyzer with either a Eukaryote Total RNA Pico or Nano kit. RNA Integrity Numbers (RIN) for samples were from 7.8–10.00 indicating high quality RNA. Over 4 separate experiments, approximately 13.5 pg RNA/cell was recovered. At least 10ng of total RNA was processed with the Ovation Pico whole transcriptome amplification kit and hybridized with Human Gene ST 2.1 plate using manufacturer instructions.

### Bioinformatic analysis

CEL files processed using R and Bioconductor after loading the following, and associated, packages: oligo [1], genefilter [2], affycoretools [3], limma [4], ReportingTools [5], sva [6], gplots [7], hwriter [8], pd.hugene.2.1.st

[9], hugene21sttranscriptcluster.db [10]. Files were preprocessed by Robust Multiarray Average (RMA) with background subtraction, quantile normalization, and median polish. Data was fitted with robust probe level linear models to all the probesets for quality control analysis. Since microarrays were performed on two different days and flow cytometry performed on four separate days, batch effects (experiments) were accounted for using ComBat [11] in the sva package. Boxplot and histogram analysis showed similar profiles for the 16 microarray samples. Principal component analysis (PCA) following experiment adjustment showed clustering of similar populations (i.e. control ALDH<sup>high</sup>CD44<sup>high</sup>). Differentially expressed genes were identified using univariate comparisons [12]. First, the relative reliability of each array is estimated by measuring how well the expression values for that array follow the linear model. Second, data is fitted to a linear model (limma package) [12]. Third, the estimated coefficients and standard errors are computed. Finally, initial statistics are determined using an empirical Bayesian model. Multiple testing comparisons were adjusted using Benjamini and Hochberg (aka FDR) [13]. Probes with an adjusted *p*-value < 0.05 were considered statistically significant. Unsupervised hierarchical clustering with complete linkage and Euclidean distance was performed on only statistically significant probes. The dataset (CEL files and ComBat intensity values) has been deposited in Gene Expression Omnibus (GEO; <http://www.ncbi.nlm.nih.gov/geo/>) of the National Center for Biotechnology Information with accession number GSE72384.

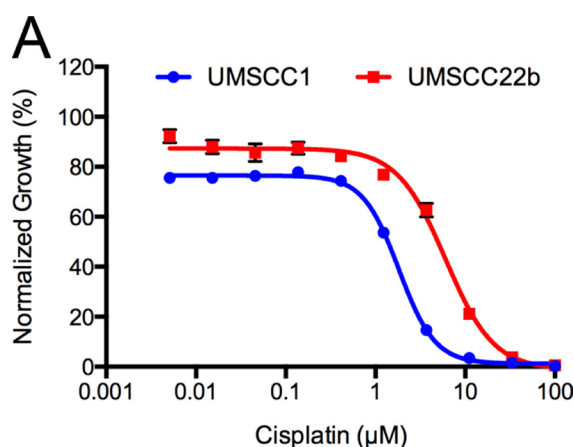
### Pathway analysis

Gene Set Enrichment Analysis (GSEA) and iPathway analysis platforms were used. GSEA is a computational method that determines whether an a priori defined set of genes shows statistically significant, concordant differences between two biological states (e.g. phenotypes) [14]. The iPathway software analysis tool implements an 'Impact Analysis' approach that takes into consideration the direction and type of all signals on a pathway, the position, role and type of every gene, etc. [15, 16]. The Impact Analysis develops two *p*-values using two orthogonal approaches based on over-representation

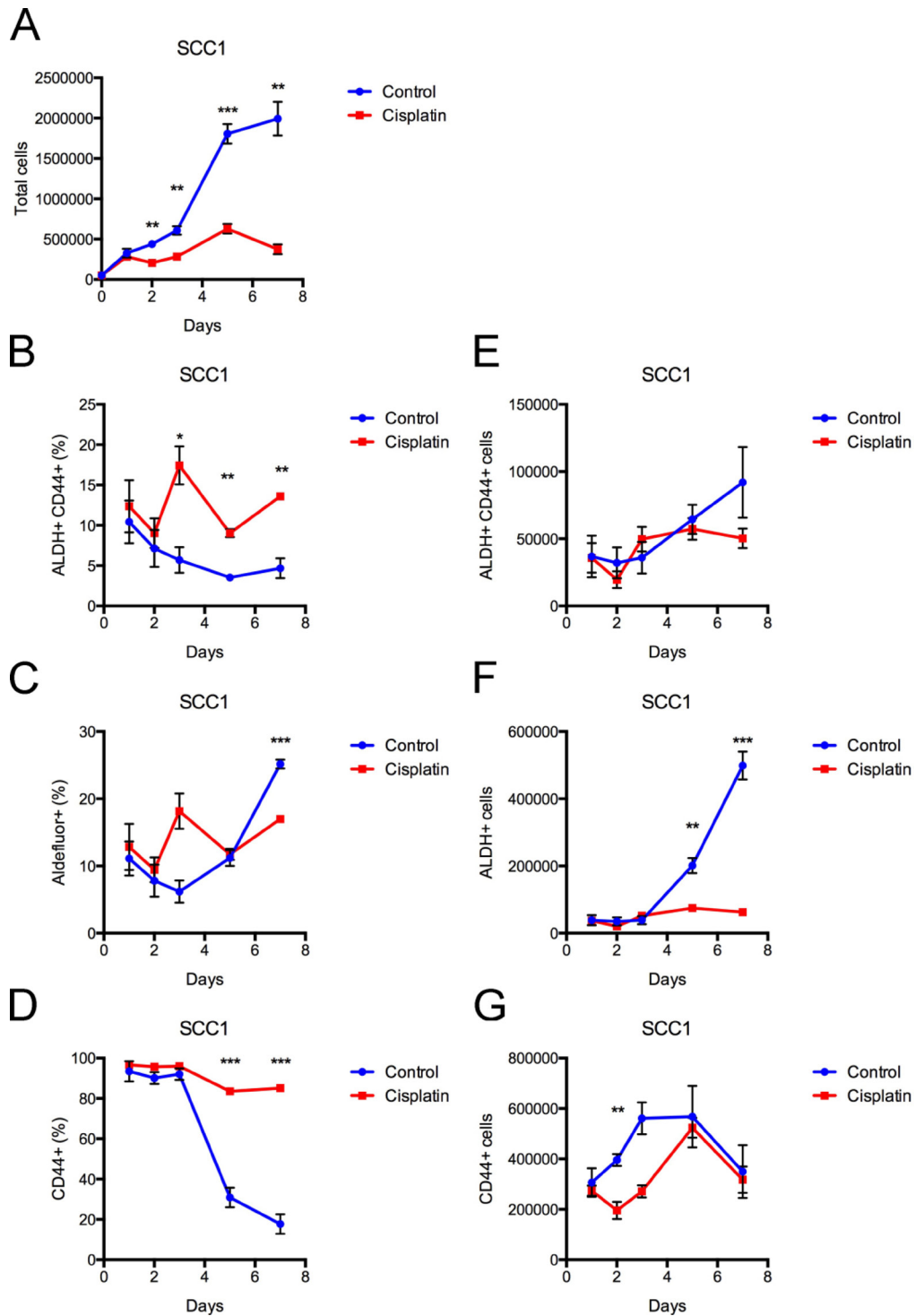
and the accumulated perturbation. These two  $p$ -values are combined into a global  $p$ -value for each pathway and iPathway adjusts for multiple testing based on Bonferroni and FDR approaches [15, 16].

## REFERENCES

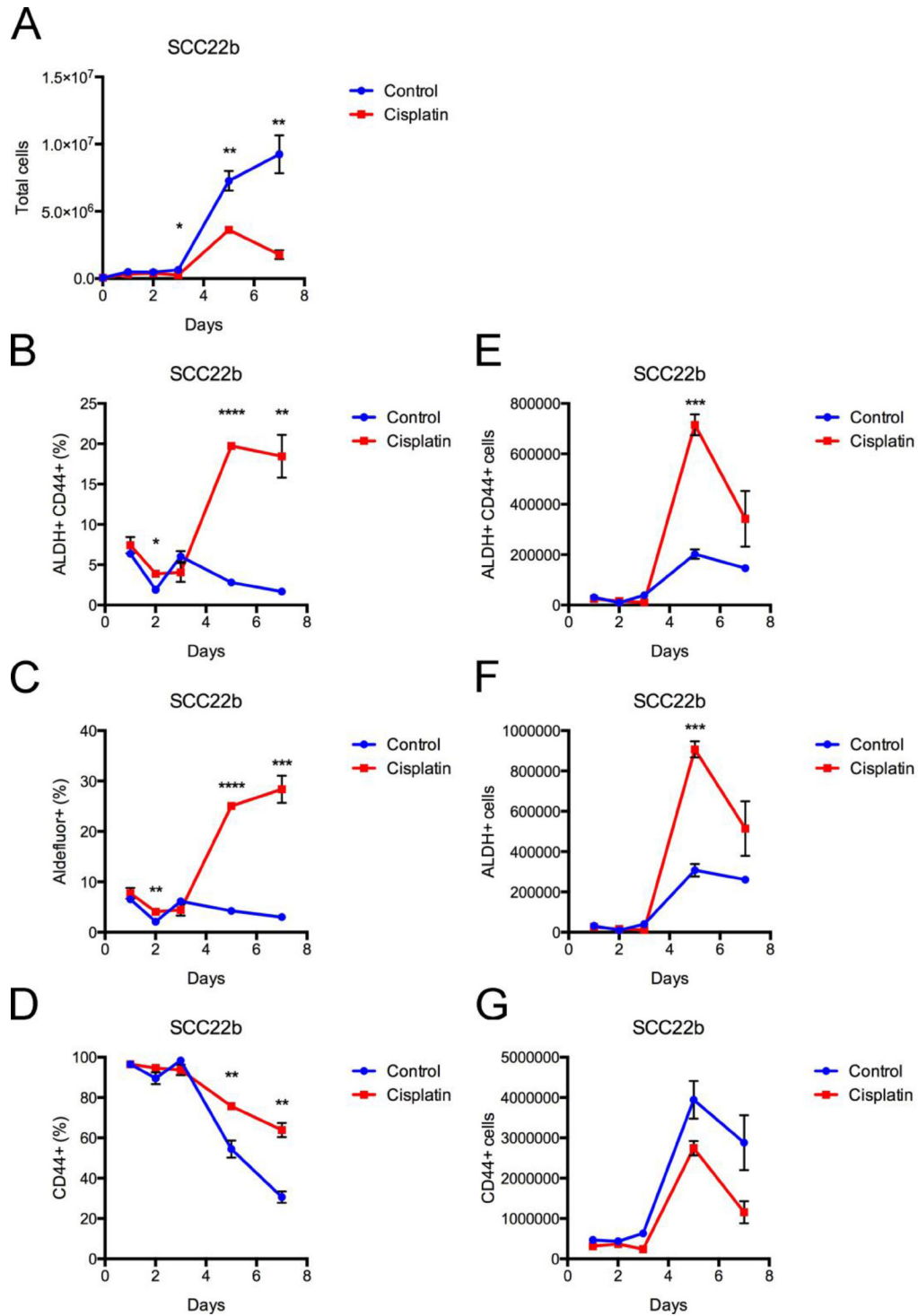
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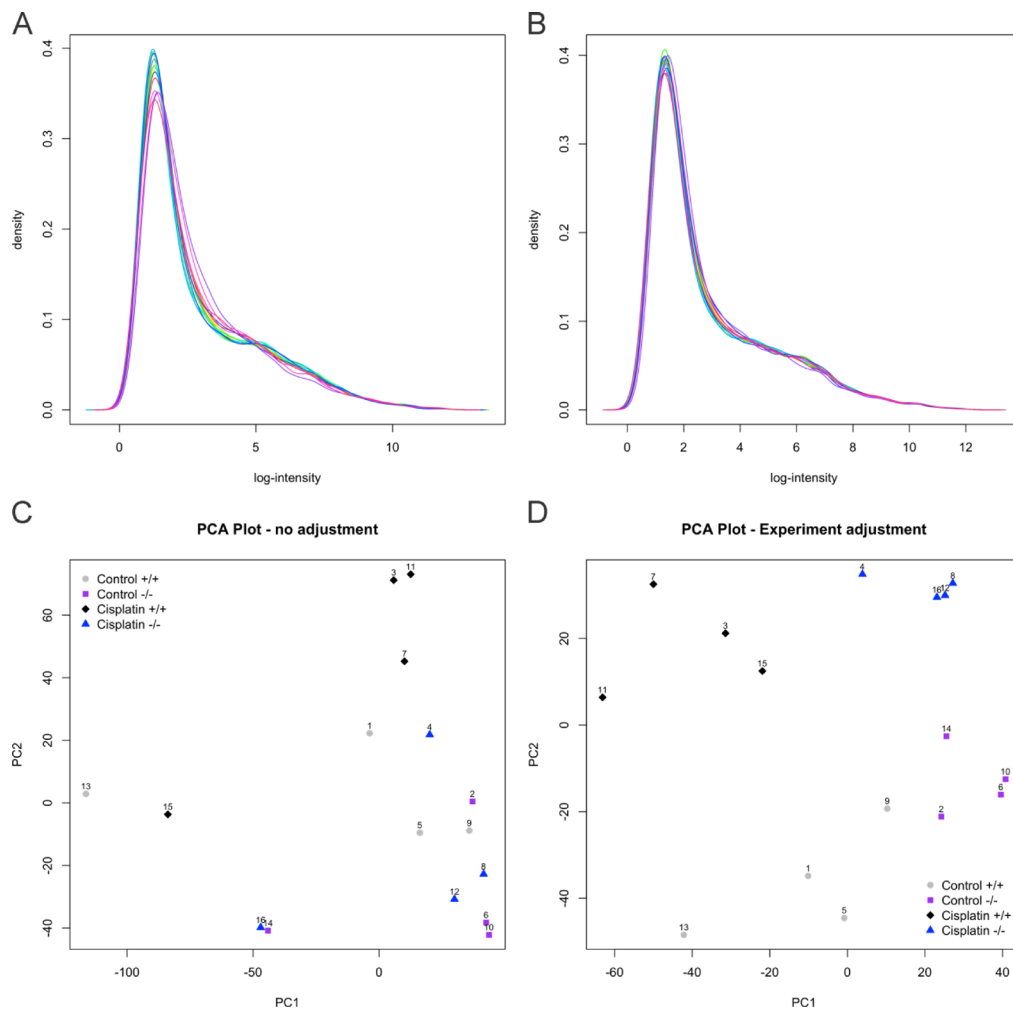
**Supplementary Figure 1: IC<sub>50</sub> analysis of cisplatin-treated UM-SCC-1 and UM-SCC-22b cells.** (A) UM-SCC-1 (blue circles) and UM-SCC-22b (red squares) were plated in 96-well plates and treated for 3 days with increasing concentration of cisplatin. Growth was measured by Alamar Blue assay, normalized to untreated cells (100%) and wells containing only media (0%), and curve-fitted with non-linear regression.



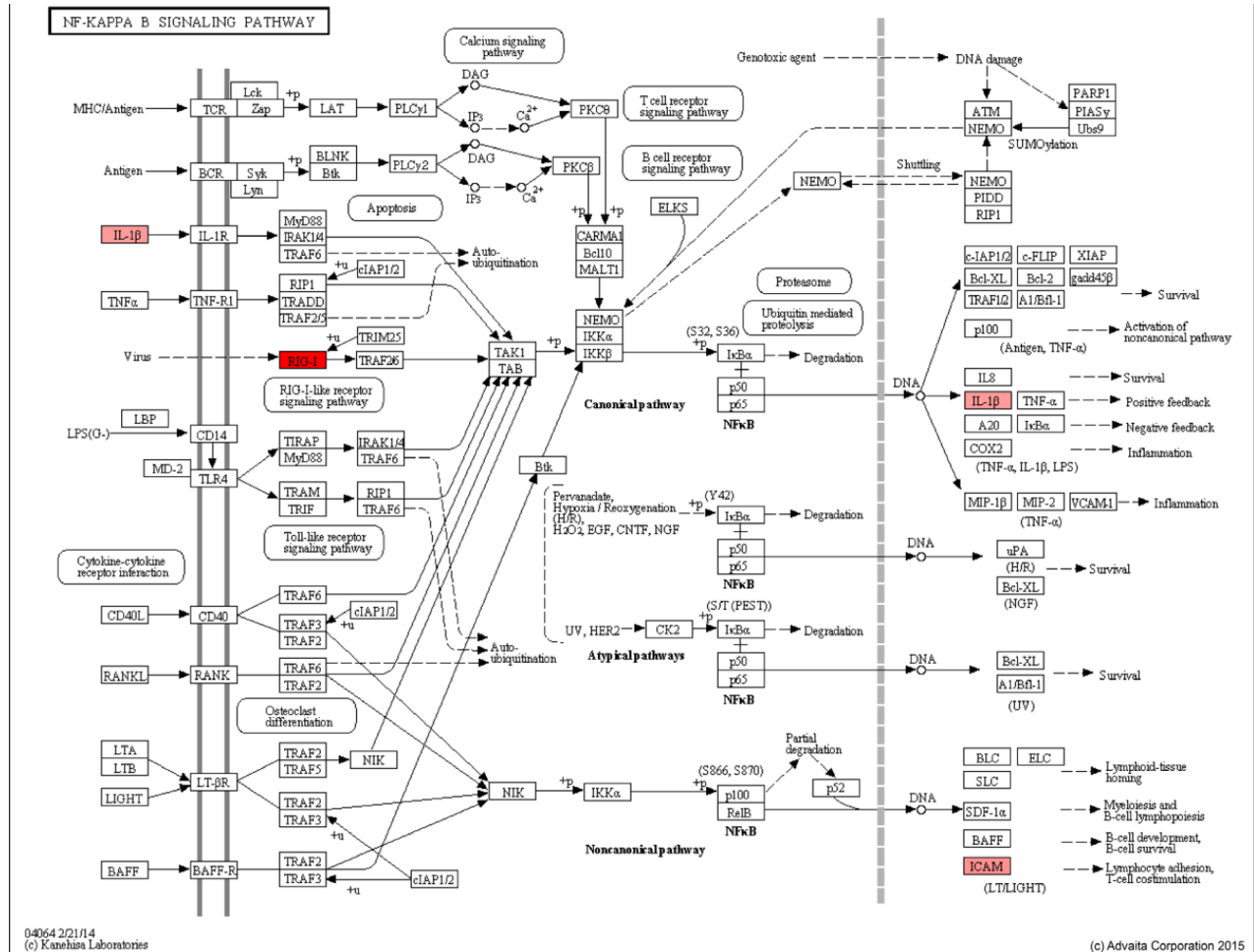
**Supplementary Figure 2: Response of UM-SCC-1 cells to cisplatin treatment.** (A) Total cell counts for control (blue circles) and 2 $\mu$ M cisplatin (red squares). The frequency of (B) ALDH<sup>high</sup>CD44<sup>high</sup>, (C) ALDH<sup>high</sup>, and (D) CD44<sup>high</sup> cells using gating based on DEAB sample. The absolute number of (E) ALDH<sup>high</sup>CD44<sup>high</sup>, (F) ALDH<sup>high</sup> and (G) CD44<sup>high</sup> cells. Mean  $\pm$  standard deviation.  $N = 3$ . Two-sided t-tests with Bonferroni multiple testing correction. \* $p < 0.05$ , \*\* $p < 0.01$ , \*\*\* $p < 0.001$ , \*\*\*\* $p < 0.0001$ .



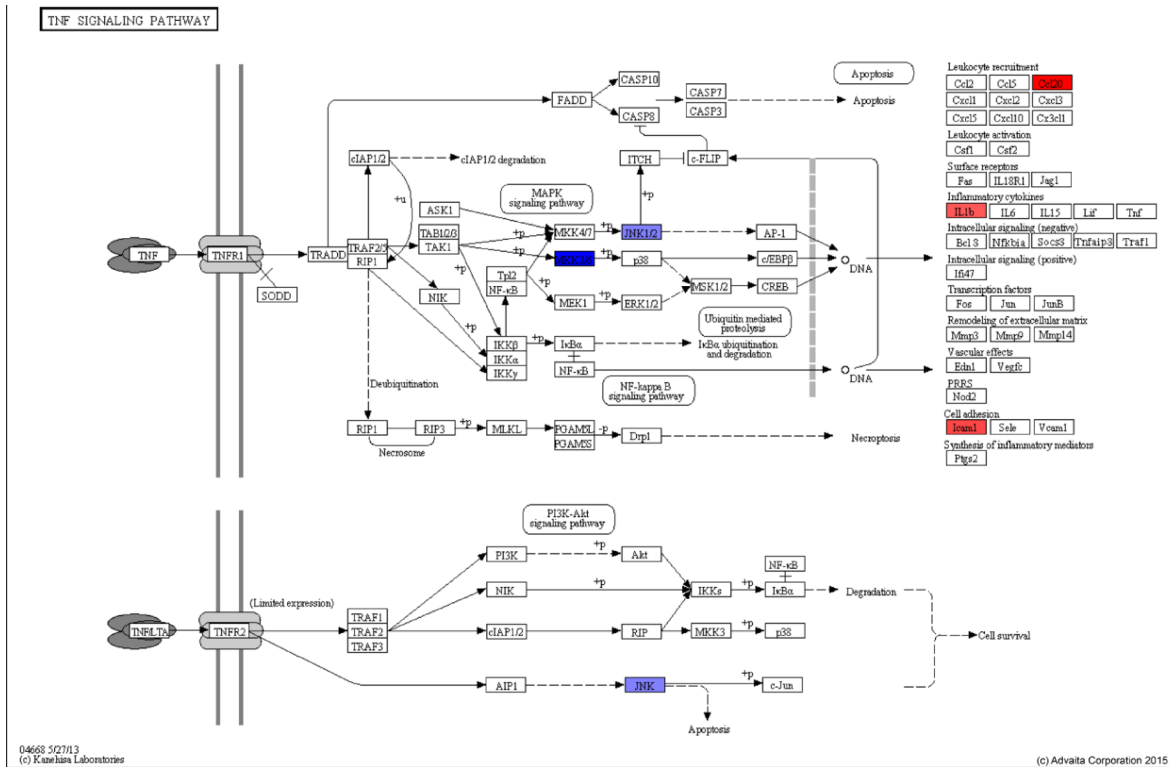
**Supplementary Figure 3: Response of UM-SCC-22b cells to cisplatin treatment.** (A) Total cell counts for control (blue circles) and 2 $\mu$ M cisplatin (red squares). The frequency of (B) ALDH<sup>high</sup>CD44<sup>high</sup>, (C) ALDH<sup>high</sup>, and (D) CD44<sup>high</sup> cells using gating based on DEAB sample. The absolute number of (E) ALDH<sup>high</sup>CD44<sup>high</sup>, (F) ALDH<sup>high</sup>, and (G) CD44<sup>high</sup> cells. Mean  $\pm$  standard deviation.  $N = 3$ . Two-sided t-tests with Bonferroni multiple testing correction. \* $p < 0.05$ , \*\* $p < 0.01$ , \*\*\* $p < 0.001$ , \*\*\*\* $p < 0.0001$ .



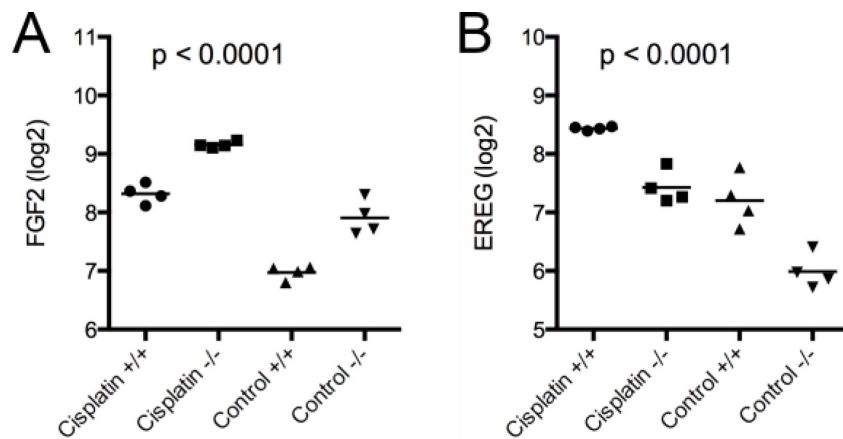
**Supplementary Figure 4: Adjustment of microarray batch variable.** (A, B) Histogram analysis of 16 microarray samples (A) before or (B) after 'Combat' batch adjustment. (C, D) Principal component analysis of 16 microarray samples (C) before or (D) after 'Combat' batch adjustment. Samples 13-16 are the pilot microarray and represent the largest variance.



**Supplementary Figure 5: NF-κB Signaling Pathway Enriched via iPathway Analysis.** iPathway analysis identified NF-κB signaling (via KEGG database) enriched in the comparison of cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> vs. control ALDH<sup>high</sup>CD44<sup>high</sup>. Genes enriched in cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> are depicted in red; genes enriched in control ALDH<sup>high</sup>CD44<sup>high</sup> are depicted in blue.

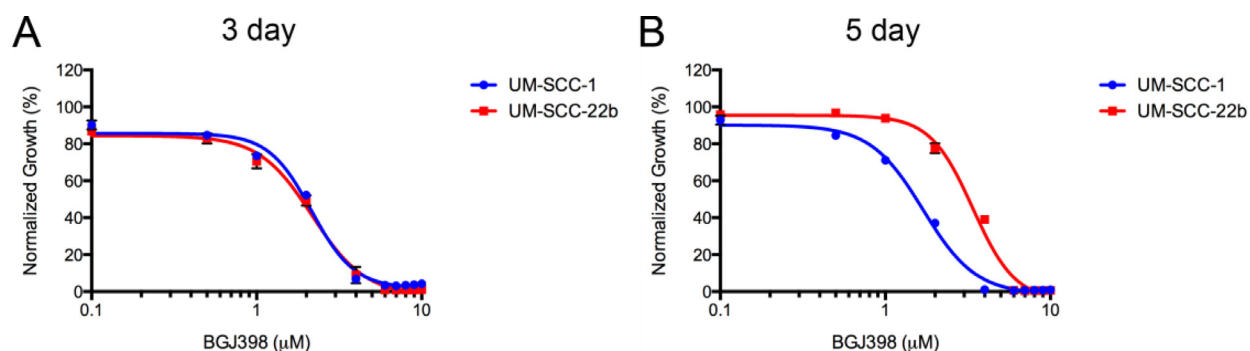


**Supplementary Figure 6: TNF Signaling Pathway Enriched via iPathway Analysis.** iPathway analysis identified TNF signaling (via KEGG database) enriched in the comparison of cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> vs. control ALDH<sup>high</sup>CD44<sup>high</sup>. Genes enriched in cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> are depicted in red; genes enriched in control ALDH<sup>high</sup>CD44<sup>high</sup> are depicted in blue.



**Supplementary Figure 7: Evaluation of FGF2 and EREG from microarray.** Log2 expression values from the microarrays for (A) FGF2 and (B) EREG. *P*-values represent overall one-way ANOVA.





**Supplementary Figure 8: BGJ398 IC<sub>50</sub> analysis.** 1000 UM-SCC-1 (blue circles) or 2000 UM-SCC-22b (red squares) cells were plated for (A) 3 or (B) 5 days in 96-well plates. BGJ398 was added for a final concentration from 0.1 to 10  $\mu\text{M}$ . Growth was measured by Alamar Blue assay, normalized to untreated cells (100%) and wells containing only media (0%), and curve-fitted with non-linear regression.

**Supplementary Table 1: RNA integrity of microarray samples**

Exp	Treatment	Population	RIN	RNA (ng)	# cells	RNA (pg)/cell
Pilot	Control	ALDH <sup>high</sup> CD44 <sup>high</sup>	9.00	88	8,960	9.8
Pilot	Control	ALDH <sup>low</sup> CD44 <sup>low</sup>	9.90	2,561	426,880	6.0
Pilot	Cisplatin	ALDH <sup>high</sup> CD44 <sup>high</sup>	9.20	159	10,700	14.9
Pilot	Cisplatin	ALDH <sup>low</sup> CD44 <sup>low</sup>	9.80	4,583	376,980	12.2
1	Control	ALDH <sup>high</sup> CD44 <sup>high</sup>	9.50	79	2,770	28.4
1	Control	ALDH <sup>low</sup> CD44 <sup>low</sup>	8.10	551	89,250	6.2
1	Cisplatin	ALDH <sup>high</sup> CD44 <sup>high</sup>	8.60	70	2,730	25.7
1	Cisplatin	ALDH <sup>low</sup> CD44 <sup>low</sup>	8.20	85	8,650	9.8
2	Control	ALDH <sup>high</sup> CD44 <sup>high</sup>	9.00	163	13,250	12.3
2	Control	ALDH <sup>low</sup> CD44 <sup>low</sup>	10.00	6,488	766,900	8.5
2	Cisplatin	ALDH <sup>high</sup> CD44 <sup>high</sup>	9.00	104	2,170	48.0
2	Cisplatin	ALDH <sup>low</sup> CD44 <sup>low</sup>	10.00	2,183	325,060	6.7
3	Control	ALDH <sup>high</sup> CD44 <sup>high</sup>	9.20	300	32,800	9.1
3	Control	ALDH <sup>low</sup> CD44 <sup>low</sup>	10.00	9,126	1,090,000	8.4
3	Cisplatin	ALDH <sup>high</sup> CD44 <sup>high</sup>	7.80	117	3,700	31.7
3	Cisplatin	ALDH <sup>low</sup> CD44 <sup>low</sup>	10.00	7,501	667,840	11.2
4	Control	ALDH <sup>high</sup> CD44 <sup>high</sup>	8.50	26	405	63.3
4	Control	ALDH <sup>low</sup> CD44 <sup>low</sup>	8.40	614	115,130	5.3
4	Cisplatin	ALDH <sup>high</sup> CD44 <sup>high</sup>	9.10	45	2,650	17.2
4	Cisplatin	ALDH <sup>low</sup> CD44 <sup>low</sup>	10.00	4,423	512,470	8.6

The RNA Integrity Number (RIN), total RNA (ng), number of FACS-collected cells, and amount of RNA (pg) per cell for the FACS experiments used for microarray analysis.



**Supplementary Table 2: Leading Edge Analysis of Gene Sets Enriched in Cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> vs. Control ALDH<sup>high</sup>CD44<sup>high</sup>**

Gene	HALLMARK INTERFERON ALPHA RESPONSE	HALLMARK TNF $\alpha$ SIGNALING VIA NFKB	HALLMARK IL6 JAK STAT3 SIGNALING
CXCL10	X	X	X
CXCL11	X	X	X
IRF1	X	X	X
RIPK2	X	X	
IL15RA		X	X
IL6		X	X
IL4R	X		X
LY6E	X	X	
BST2	X	X	
RTP4		X	X
IL15	X		X
IRF7	X		X
IFIT2		X	X
IFIH1		X	X
IL1B		X	X
ICAM1		X	X

Gene Set Enrichment Analysis was performed for cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> vs. control ALDH<sup>high</sup>CD44<sup>high</sup>. Leading Edge Analysis was performed using three of the “Hallmark” gene sets enriched in cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> cells. Some of the genes are in common within the three Hallmark gene sets, Interferon Alpha Response, TNF $\alpha$  Signaling Via NF- $\kappa$ B, and IL6-JAK-STAT3 Signaling.

**Supplementary Table 3: iPathway analysis of Cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> vs. Control ALDH<sup>high</sup>CD44<sup>high</sup>**

Pathway Name	<i>P</i> -value	Bonferroni Adj <i>P</i> -value	FDR Adj <i>P</i> -value
Influenza A	2.49E-07	3.61E-05	2.33E-05
Herpes simplex infection	3.21E-07	4.65E-05	2.33E-05
Measles	9.42E-05	0.013654	0.004551
Malaria	0.000209	0.030366	0.007591
Hepatitis C	0.000382	0.055426	0.011085
TNF signaling pathway	0.005782	0.838410	0.126821
NF-kappa B signaling pathway	0.007406	1	0.126821
Epstein-Barr virus infection	0.007813	1	0.126821
Cytosolic DNA-sensing pathway	0.007872	1	0.126821
Osteoclast differentiation	0.012901	1	0.185015
ErbB signaling pathway	0.014036	1	0.185015
Toll-like receptor signaling pathway	0.020263	1	0.236966
Hepatitis B	0.021245	1	0.236966
RIG-I-like receptor signaling pathway	0.023531	1	0.243713
Nicotinate and nicotinamide metabolism	0.026829	1	0.245565
Leishmaniasis	0.027410	1	0.245565
Pentose phosphate pathway	0.028790	1	0.245565
Amoebiasis	0.036477	1	0.293841
African trypanosomiasis	0.044945	1	0.343000

Pathway analysis was performed with iPathway using the genes with adjusted *P*-value < 0.05 in the cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> vs. control ALDH<sup>high</sup>CD44<sup>high</sup> comparison. Unadjusted *p*-value, bonferroni and FDR adjusted *p*-values are shown for the pathways with an unadjusted *p*-value < 0.05.

**Supplementary Table 4: Genes higher in cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> cells vs. control ALDH<sup>high</sup>CD44<sup>high</sup> cells.** RMA normalized microarray data was fitted to a linear model and initial statistics were determined using an empirical Bayesian model. Multiple testing comparisons were adjusted using Benjamini and Hochberg (aka FDR). Probes with an adjusted *p*-value < 0.05 were considered statistically significant. The Affymetrix ProbeID, EntrezID, HUGO gene symbol, gene name, log2 fold change, and adjusted *p*-value are shown for the 115 genes higher in cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> vs. control ALDH<sup>high</sup>CD44<sup>high</sup>. See Supplementary\_Table\_4

**Supplementary Table 5: Genes lower in cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> cells vs. control ALDH<sup>high</sup>CD44<sup>high</sup> cells.** RMA normalized microarray data was fitted to a linear model and initial statistics were determined using an empirical Bayesian model. Multiple testing comparisons were adjusted using Benjamini and Hochberg (aka FDR). Probes with an adjusted *p*-value < 0.05 were considered statistically significant. The Affymetrix ProbeID, EntrezID, HUGO gene symbol, gene name, log2 fold change, and adjusted *p*-value are shown for the 116 genes lower in cisplatin ALDH<sup>high</sup>CD44<sup>high</sup> vs. control ALDH<sup>high</sup>CD44<sup>high</sup>. See Supplementary\_Table\_5

**Supplementary Table 6: IC<sub>50</sub> analysis of BGJ398-treated UM-SCC-1 and UM-SCC-22b cells**

IC50 (μM)	3 day		5 day	
	UM-SCC-1	UM-SCC-22b	UM-SCC-1	UM-SCC-22b
Average	2.211	2.210	1.909	3.569
Std. Deviation	0.084	0.056	0.308	0.258
Std. Error of Mean	0.060	0.040	0.218	0.183

The average BGJ398 IC<sub>50</sub> (μM), standard deviation, and standard error of mean from representative experiment of 2 plates/ experiment with 3 wells per dose. Cells were treated for 3 or 5 days with 0.1 to 10μM BGJ398 and growth measured by Alamar Blue assay.