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## Supplemental Information

# Neural signaling modulates

## metabolism of gastric cancer

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**Figure S1, Study design, related to Figure 1:** Drawing showing study design of translational research approach and methodology used (indicated in arrows).



**Figure S2, Gastric cancer is glutamine-dependent, related to Figure 4:** Levels of metabolites in mouse gastric cancer (GC) (marked in black), wild-type WT (blue), GC after vagotomy (VT) (red) and WT after vagotomy (VT) (purple)(**A-D**) related to **Figure 4.** Glu: L-glutamate; Gln: L-glutamine; GSH: reduced glutathione; GSSG: oxidized glutathione; Gly: glycine; Thr: threonine; Oxo: 5-oxoproline; C-at: cisaconitate; Glc: glucose; G6P: glucose-6-phosphate; F6P: fructose-6-phosphate. Bars represent relative scaled intensities with SEM and one-way ANOVA p-values. The values were as same as ones in **Figure. 4B-N.** For detailed information, see Table S6. Endogenous levels of L-glutamate and L-glutamine in gastric cancer cells AGS during culture period from 1 to 24 hrs (**E**). Gln reduction (**F,G**) and Gln or Pyr depletion (**H,I**) in the medium in AGS (**F,H**) and MKN45 (**G,I**) cell culture periods of 24, 48 and 72 hrs. Mean of n=3-12 replicates/treatment with SD. Proliferation was assessed using Cell count reagent SF and cell proliferation was calculated relative to controls.



**Figure S3, Drug target prediction, related to Figure 6:** Waterdrop diagrams showing drug-target interaction prediction and computational drug repositioning in human GC. Note: nodes of RAD001-targeted mTOR (marked in red), CPI-613 targeted PDP1 and α-KGDH (also known as OGDH, purple), BoNT-A-targeted SNAP25 (yellow) and L-DON/968/CB-839/BPTES-targeted GLS (light blue). Lines represent biological interactions between molecules that include proteins, genes, mRNAs, microRNA, lncRNAs and metabolites, generated from differentially expressed drug target genes (only drug targets differentially expressed at p<0.05, q<0.05 are shown).



**P ro life ra tio n in h ib itio n (% o f c o n tro l)**

**Figure S4,** *In vitro* **drug screening, related to Figure 7:** Proliferation inhibition rates of *in vitro* treatment of BoNT-A, RAD001, CPI-613, 5-FU and oxaliplatin either alone or in different combinations at increasing doses using MKN74 cells. Mean of n=3-12 replicates/treatment with SD. Proliferation was measured using CCK-8 Kit at 450 nm and treatments were normalized to respective vehicle controls.



## **Figure S5, Nerve-cancer metabolism in gastric cancer, related to Figure 8A:**

Transcriptome profiling of genes involved in the nerve-cancer metabolism pathways of synaptogenesis signaling pathway, WNT/β-catenin signaling, mTOR pathway and energy metabolism. Correlations between mouse GC with *vs.* without unilateral vagotomy (UVT)(**A**), between mouse GC with *vs.* without BRC (**B**), between mouse GC with RC *vs.* UVT (**C**), and between mouse GC with BRC *vs.* UVT (**D**). Linear regression lines were drawn using GraphPad Prism v6. Pearson's test for correlation was used.



**Figure S6, Single-cell atlas and glutamine pathways, related to Figure 8C-D:** Computational network modeling showing interactions within B cell gene markers (**A**), macrophage gene markers (**B**), fibroblast gene markers (**C**), mast cell gene markers (**D**) and endothelial cell gene markers (**E**) and connections between the cell types and glutaminolysis (**A-E**) based the single-cell transcriptome atlas (Zhang et al., 2019)(GSE134520).



**Figure S7,** *In silico* **modelling, related to Figure 8:** Representative prediction of downstream effect of *in silico* inhibition of the CTNNB1 node (marked in green and annotated by black arrow) in the WNT signaling cluster (left) on mTOR signaling (right)(**A**) and effects of treatment of RC for 2 months (2M) on WNT/β-catenin signaling pathway and mTOR signaling clusters (**B**). Overlay gene expression: GC *vs.* WT. MAP (molecular activity prediction) to generate predictions. Semi-quantitative method: dark blue represent -2, light blue represent -1, white represent 0, light orange represent +1 and dark orange represent +2.

# **Table S1, Genes detected by real-time PCR and RNAseq, related to Figure 1:**

List of genes detected by both RNAseq and real-time PCR and correlation analysis (figure).



**XM\_134865** Pygo1

PCR (mouse GC) inverted  $\Delta \textsf{CT}$ 

<b>Upstream</b> Regulator	Log <sub>2</sub> FC	<b>Predicted</b> <b>Activation</b> <b>State</b>	<b>Activation</b> z-score	p-value of overlap	<b>Target</b> <b>Molecules in</b> <b>Dataset</b>	<b>Target molecules</b> in mTOR pathway	<b>Molecules</b>
Tgf beta		Activated	3.848	2.99E-07	65	U	
WNT1	0.000	Activated	3.138	6.85E-03	34	3	EIF3C, MRAS, PDPK1
CD44	1.156	Activated	3.695	6.05E-08	55		
<b>JUN</b>	0.184	Activated	2.219	5.21E-03	55		<b>RHOB</b>
TGFB1	0.727	Activated	5.957	2.39E-17	236	3	PPP2CA, PRKCG, RHOA
TGFBR1	$-0.536$	Activated	2.735	5.98E-04	20		
TGFB <sub>2</sub>	1.573	Activated	3.309	4.55E-11	34		
CTNNB1	0.208	Activated	3.741	4.53E-10	165		PPP <sub>2CA</sub>

**Table S2, Upstream regulators in mouse GC, related to Figure 2F:** Upstream analysis of WNT/β-catenin pathway regulators and mTOR targets in mouse GC, related to **Figure 2F.** Predicted Activation State was Activated for z-score>2.000.

**Table S3, Metabolite signature, related to Figure 4:** List of metabolites of gastric cancer (GC) mice and wild-type (WT) mice presented in **Figure 4A.** *GC: gastric cancer; WT: wild-type; FC: Fold change.* Green: p≤0.05, fold change <1.00; Red: p≤ 0.05, fold change ≥1.0. White:  $p < 0.05$ , 1.0 ≤ fold change >1.0.





## **Table S4, Energy metabolites, related to Figure 4B-N:** Statistic data corresponding to metabolites shown in **Figure 4B-N and Figure S2A-D.**



Mean: scaled intensity of N=10 (WT) or N=6 (GC), *p*-value: One-way ANOVA test between groups, GC: Gastric cancer, WT: wildtype, UVT: Unilateral vagotomy, Glu: L-glutamate, Gln: L-glutamine, GSH: glutathione, reduced, GSSG: glutathione, oxidized, Gly: glycine, Thr: threonine, Oxo: 5-oxoproline, C-at: cis-aconitate, Glc: glucose, G6P: fructose-6-phosphate, F6P: fructose-6-phosphate.

## **Table S5, Signaling pathways involved in mouse gastric cancer (GC), related to Figure 5A:** Multi-omics integrative analysis in IPA revealed 41 signaling pathways that appeared exclusively in Mouse GC *vs*. WT.





## **Table S6, Signaling pathways involved in mouse gastric cancer (GC) after**

**vagotomy, related to Figure 5B:** Multi-omics integrative analysis in IPA revealed 24 signaling pathways that appeared exclusively in mouse GC after vagotomy *vs.* sham operation.





**Table S7, Signaling pathways involved in mouse gastric cancer (GC) with and without vagotomy, related to Figure 5C:** Multi-omics integrative analysis in IPA revealed 13 signaling pathways present in comparison between mouse GC *vs.* WT and in mouse GC after vagotomy *vs.* sham operation.



<b>Patient</b> number	Age at inclusion	TNM stage at time of diagnosis	Chemotherapy	TNM stage at inclusion time	<b>Tumor location and size</b>
	81	T4N1M1	1 <sup>st</sup> line treatment, stopped due toxic side effects	<b>T4N1M1</b>	Greater curvature, lesser curvature and anterior wall. Longest diameter 9 cm
	70	T3N0M0	Neoadjuvant chemotherapy, inoperabel due to comorbidity	<b>T3N1M0</b>	Cardia and proximal esophagus. Longest diameter 3 cm
	79	T4aN2M1	Palliativ chemotherapy with EOX	T4aN2M1	Distal part of the stomach. Circular tumor with longest diameter 8 cm
4	49	TxNxM1	1 <sup>st</sup> line treatment, 2 <sup>nd</sup> line treatment.	TxNxM1	Cardia. Extensive liver metastasis
5.	83	<b>TxNxMx</b>	No previous chemotherapy due to age and comorbidity	T4aN3M1	Cardia. Extensive liver metastasis
6	84	T4aN3M0	No previous chemotherapy due to age and comorbidity	T4aN3M0	Linitis plastica in whole stomach except the most proximal part

**Table S8, Baseline patient data, related to Figure 8L:** Baseline patient data (the first patient was recruited at September 10,

## 2014).

Note: One additional patient (i.e. patient no. 7) gave his consent for participation in this study, but pretreatment CT scan of the stomach showed no measurable tumor size. According to the study protocol (Supplementary Data Clinical study protocol), this patient was excluded from further participation in the study and did not receive BoNT-A treatment.







**Table S10, Secondary outcome measure (short term), related to Figure 8L:** Secondary outcome measure: Short-time adverse effects and ECOG status after BoNT-A injections



**Table S11, Secondary outcome measures (long-term), related to Figure 8L.** Secondary outcome measure: Long term adverse effects and ECOG status after BoNT-A injections





# **Table S12, Study groups, related to Figure 7K and Figure 8G-L.**

Endoscopic injection of BoNT-A (6) 49-84 years 2 years follow-up **# group from previous study (Zhao et al., 2014)**

**Table S13. Metabolites involved with DNA/protein synthesis, related to Figure 4 and Data S3.** Effects of vagotomy (unilateral vagotomy, UVT) on gastric tissue levels (scaled intensity) of metabolites that are involved in DNA/protein synthesis in either wildtype (WT) or gastric cancer (GC) mice



One-way ANOVA was used for comparisons between WT and WT (UVT) or between GC and GC (UVT).

**Table S14, Chemical and reagent list, related to methods.** List of reagents and chemicals used.





**Table S15, Description of Metabolon QC samples, related to methods.**



# **Transparent Methods**

## **GC patients**

Twenty-two patients (17 men aged 49-87 years and 5 women aged 51-83 years) were included. 16 of 22 patients underwent total/subtotal or distal gastrectomy because of intestinal or diffuse gastric cancer and were followed-up for 5 years since 2012 at St. Olavs Hospital, Trondheim, Norway. The study was approved by the Regional Committees for Medical and Health Research Ethics Central Norway (REK 2012- 1029). 6 of 22 patients were enrolled in a clinical trial (see below) (Table S12). Total, subtotal or distal gastrectomy was performed on 16 patients diagnosed with gastric cancer. Biopsies from 4 pre-determined positions in corpus (major and minor curvature), cardia and antrum were collected, and largest diameter of the tumor was decided. Biopsies from adjacent, normal tissue was taken 5-10 cm from the tumor site. TNM status was defined, and samples were classified according to Lauren's classification, (Intestinal, diffuse or mixed/combined type), WHO classification (tubular, papillary, mucinous and poorly cohesive), WHO grading (well, moderately or poorly differentiated), and were reviewed according to the Japanese pathological classification. Samples were assigned gastric histopathology scoring including inflammation, epithelial defects, oxyntic atrophy, epithelial hyperplasia and dysplasia and an overall GHAI score.

## **Animals**

Three hundred-twenty four mice were used and some of the mice were followed- up for more than one year to measure the overall survival rate. The mouse GC model was the transgenic INS-GAS mice which spontaneously develop GC at our own institute (Wang et al., 1996; Zhao et al., 2014) and its wild-type (WT) mice (FVB strain). Mice were housed ~5 mice per cage on wood chip bedding with a 12-hour light/dark cycle in a specific pathogen free environment with room temperature of 22°C and 40- 60% relative humidity. Mice including both INS-GAS and WT mice were age-matched and randomized into different experimental groups (Table S12). All animal experiments were approved by The Norwegian Food Safety Authority (Mattilsynet).

## **Surgery**

Vagotomy and BoNT-A injections were performed under isoflurane anesthesia as described previously(Zhao et al., 2014). The success of UVT was confirmed by reduced thickness of gastric mucosa (Zhao et al., 2014) and reduced tissue-levels of metabolites that are involved in DNA/protein synthesis in the denervated side in comparison with the innervated side of stomach (Table S13).

## **Chemicals and reagents**

For details, see chemical and reagent list in Table S14.

## **Cells and cell culture**

GC cell lines included AGS (female, 54 years, Caucasian), MKN74 (male, 37 years, Asian), MKN45 (female, 54 years, Caucasian) and KATO-III (male, 55 years, Asian). AGS cells were kindly provided by Prof. Sasakawa (Tokyo University, Japan). MKN45 cells were kindly provided by Prof. Kamiya (Kyorin University, Japan). MKN74 cells were provided by Prof. T.C Wang and KATO-III cells were purchased from LGC group.

AGS and MKN45 cells were maintained in Dulbecco's Modified Eagle's Medium (DMEM (1.0 g/l Glucose, 10 mM) with L-Gln (584.00 mg/L, 4.0 mM) and Sodium Pyruvate (110.00 mg/L, 1.0 mM)(Nacalai tesque, Japan) supplemented with 10% fetal bovine serum (FBS; ThermoFisher Scientific, Grand Island, NY) and antibioticantimycotic solution (1%) containing penicillin, streptomycin and amphotericin B (Nacalai tesque, Japan). MKN74 and KATO-III cells were maintained in RPMI-1640 medium (Sigma Aldrich, Norway) supplemented with fetal bovine serum (10%, FBS), Sodium pyruvate and penicillin streptomycin solution (1%) in a humidified incubator holding  $5\%$  CO<sub>2</sub> and  $37^{\circ}$ C.

#### *In vitro* **experiments**

#### **Gln/pyr depletion**

The cells (1.0x10<sup>4</sup>) were plated (24h) and treated with 0-2.0 mM L-glutamine and 1.0 mM pyruvate in DMEM supplemented with dialyzed bovine serum (10%) and glucose at 25 mM. In depletion testing, either glutamine or pyruvate were omitted from the medium. Proliferation was assessed using Cell Count Reagent SF or Cell counting Kit-8 reagent at 450 nm and cell proliferation was calculated relative to controls. Determination of endogenous L-glutamine and L-glutamate was performed after 1, 6 and 24 hrs in culture using a detection kit (Glutamine/glutamate determination kit, Sigma, Saint Louis, Missouri).

#### **Drug screen**

Cells (2.5x10<sup>3</sup>) were plated (24 hrs) and subjected to individual dose-response drug screens and sequential combination treatment during 3 days in culture. First, cells were treated with either serum-free medium or BoNT-A- without serum at 0.25 U BoNT-A/well and incubated for 24 hrs. CPI-613 and RAD001 were dissolved in DMSO at highest solubility before diluted in the medium. The cells were treated with RAD001, CPI-613, combination of these or vehicle (DMSO) control and incubated for 24 hrs. A combination of 5-FU and oxaliplatin or medium control was added to the cells for 24 hrs. To assess whether the drug combinations acted synergistically, we calculated Bliss synergy scores for RAD001 + CPI-613 combinations using the SynergyFinder web-application (Ianevski et al., 2017). Synergy scores were quantified as an average excess over expected drug combination effect given by the Bliss reference model (Ianevski et al., 2019). Bliss Independence model was used because the two drugs (i.e. RAD001 and CPI-613) act independently in such a manner that neither of them interferes with the other (different sites of action), but each contributes to a common result, i.e. cell proliferation.

#### *In vivo* **experiments**

GC mice were injected BoNT-A through laparoscopic procedure as described earlier (Zhao et al., 2014), treated with RAD001 (1.5 mg/kg/day for 3 weeks, i.p.), CPI-613 (20 mg/kg/week, once weekly for 3 weeks, i.p.), or combination of RAD001 and CPI-613. Saline injection (i.p.) was used as control. The mice were allowed one-week rest after the first cycle of treatment, and then the treatment cycle was repeated once, yielding a total treatment window of 8 weeks (**Figure 7H**). BoNT-A was dissolved in saline containing methylene blue (1.0 %) to visualize the injection. The achieved

concentration of BoNT-A was 0.25 U of BoNT-A/mL. Injection was performed through laparotomy into the serosa layer in the anterior side of stomach. Thus, for a mouse receiving 0.4 mL BoNT-A (0.25 U/mL) the dose corresponded to 0.10 U. 5-Fluorouracil (5-FU) was given i.p. at dose of 25 mg/kg in a volume of 0.5 mL. Oxaliplatin was given i.p. at dose of 5 mg/kg in volume of 0.5 mL. The two drugs were injected on either left or right side of abdomen at same time once weekly for 3 weeks in 2 cycles, starting one week after BoNT-A injection.

#### **Sample collection and preparation**

Mouse tissue samples were taken after the animals were killed under deep isoflurane inhalation anesthesia. The anterior and posterior parts of stomachs were collected for histopathological analysis and cryopreservation for transcriptomics of mouse GC in which mice underwent unilateral vagotomy (UVT) at 6 months of age and the stomachs were collected 6 months afterwards, the data from our previous study was re-analyzed (according to 3R principle)(Zhao et al., 2014). For metabolomics, GC and WT mice at 6 months of age underwent the same UVT or sham operation and the stomachs were collected as described previously. Six months after UVT, animals were terminated for sampling, and tissue samples from the denervated anterior stomach and tissue samples from the posterior stomach with intact innervation were analyzed with liquid chromatography/mass spectrometry and gas chromatography/mass spectrometry. Mouse tissue samples were collected for transcriptomics analysis immediately after completing two months BRC-treatment.

#### **Measurement of survival rate, body weight and tumor size**

Animals were followed up by daily inspection with scoring sheet, weighing and euthanized according to primary human endpoints. Scoring parameters included severe body weight loss (>25%), stress behavior, abdominal pain or reduced physical activity and was followed in collaboration with the responsible veterinarian at the animal facility. Body weight was measured daily (during treatment) or weekly (during follow up). Tumor volume density (% of glandular area of the stomach occupied by tumor) was measured using point count method described earlier<sup>14</sup>.

#### **Pilot clinical trial (phase II)**

Six patients were enrolled according to inclusion criteria and written consent (Supplementary Data: Clinical Trial Protocol). Inclusion criteria included 1) patients who received  $1^{st}$  line and  $2^{nd}$  line chemotherapy but no longer respond to such therapy, 2) patients who, due to toxicity of chemotherapy, could not be offered such treatment, 3) patients who, after meticulous information about chemotherapy, still did not want such treatment and 4) patients with performance status (ECOG) 0-2. Patients were elderly and diagnosed with already advanced gastric cancer which precluded surgical resection (Table S12). Exclusion criteria included 1) known allergy to any of the components in Botox®, 2) known peripheral motor neuropathy disease ( for example: Amyotrophic Lateral Sclerosis, ALS), or subclinical or clinical deficiency of neuromuscular transmission (for example: Myasthenia Gravis or Eaton-Lambert`s Syndrome), 3) another cancer disease that is not under control, 4) another concomitant treatment for cancer, 5) serious mental illness and 6) performance status (ECOG) 3-4. One patient with TNM status T3N0M0 was rejected for surgery due to comorbidity following a short period with neoadjuvant chemotherapy. At the time of enrollment into this study, 4 out of 6 patients had metastatic disease, and 2 of these patients had extensive liver metastasis with short expected life expectancy. Extensive tumor masses in the stomach were present in 3 out of 6 patients. The patients were admitted to the hospital shortly after the baseline CT scan, and endoscopic BoNT-A injection was performed under sedation with midazolam. One hundred units with Botox® were diluted into 14 mL saline by the Department for Clinical Studies at St.Olav Hospital's Pharmacy. This amount was divided into 7 doses of 2 mL (14.3 U/dose) that were injected at 4 sites around the tumor and at 3 sites directly into the tumor. Some of the patients had advanced and extensive tumor masses in the stomach and for those patients, injections were concentrated to the area of the stomach with measurable tumor thickness or diameter, omitting the rest of the tumor masses in the stomach. After the endoscopic procedure, the patients were observed in the surgical ward and discharged from hospital the day after the procedure. Primary outcome measures were assessment of tumor size (volume density and/or thickness) in the stomach using standardized CT protocols after 2, 8 and 20 weeks. Two weeks after the injection, the patients had an outpatient clinical visit with complete physical assessment, specially emphasizing on detecting any adverse or toxic events related to the experimental treatment. At 8 and 20 weeks after the injection, another thoracic and abdominal CT scan was performed, together with a follow-up outpatient clinical examination. Secondary outcomes included toxicity (within 2- and 8-weeks post injection) and performance status (ECOG) after 2, 8 and 20 weeks. The safety evaluation was performed based on the CTC (Common Toxicity Criteria) criteria. The study was conducted in accordance with the guidelines for GCP (Good Clinical Practice) and it was approved by the Regional Committee for Medical and Health Research Ethics (2012/1031) and the Norwegian Medicines Agency (2012-002493- 31).

#### **Transcriptomics**

Total RNA was extracted from harvested stomachs of mice or surgical biopsies of patients. RNA quality and quantity were obtained using NanoDrop One (Thermo Scientific, Norway) and Agilent Bioanalyser. RNA sequencing of human GC samples was performed using Illumina platform as described earlier<sup>14</sup>, whereas RNA sequencing of mouse samples was performed using Illumina HiSeqNS500 instrument (NextSeq 500) at 75 bp with paired end (PE) reads using NS500H flowcells with 25 M reads/sample. Paired end forward read length (R1): 81, reverse read length (R2): 81. Illumina microarray data was analyzed using Lumi on the log2 scale and was analyzed using the empirical Bayesian method implemented in Limma. Gene expression was analyzed using a t-test between cancer and WT mice or between tumor and normal adjacent tissue in patients. Transcripts with a *p*-value of less than 0.05 were considered to be differentially expressed. Benjamini-Hochberg false discovery rates were included.

#### **Metabolomics**

Metabolomics was performed using a platform that incorporates two separate ultrahigh-performance liquid chromatography/tandem mass spectrometry (UHPLC/MS/MS2) injections and one gas chromatography/mass spectrometry (GC/MS) injection per sample by Metabolon (USA). Identification, relative quantification, data-reduction and quality-assurance components of the process were included in the analysis platform. 343 metabolites were identified (Data S3). The informatics system consisted of four major components, the Laboratory Information Management System (LIMS), the data extraction and peak-identification software, data processing tools for QC and compound identification, and a collection of information interpretation and visualization tools for use by data analysts. The hardware and software foundations for these informatics components were the LAN backbone, and a database server running Oracle 10.2.0.1 Enterprise Edition. For more details, see description of Metabolon QC samples in Table S15.

#### **Real-time PCR**

Total RNA was isolated and purified using an Ultra-Turrax rotating-knife homogenizer and the mirVana miRNA Isolation Kit (AM1560, Ambion) according to the manufacturer's instructions. Mouse WNT pathway RT2 profiler PCR array was used (StepOnePlusTM, Applied Biosystems), which targeted key genes involved in the canonical and non-canonical WNT pathway and endogenous genes for reaction control (89 genes and 7 controls, see Table S1). The reaction was performed according to the manufacturer's instructions (SABiosciences Corporation, QIAGEN Norway).

#### **Data visualization**

R/Bioconductor environment was used to process omics-data before differential expression analysis. Graphical data visualization and data analyses were carried out using GraphPad Prism software 6.0 (GraphPad Software, U.S), Excel 2016 (Microsoft), IPA (Qiagen, Aarhus, Denmark) and RStudio version 3.5.2 (2018-12- 20). Diagram plots in Figure 5 were created with JavaScript library D3.js v.4. SPSS v.23-25 was used to perform test statistics including *t*-tests and non-parametric tests, one-way ANOVA, and correlation/linear regression analyses. Heatmaps were encoded in RStudio using the heatmap.2 function. Single-cell data were processed using Seurat v3 (doi.org/10.1016/j.cell.2019.05.031) and visualized in a tSNE plot (Figure 3). IPA was used to cluster cell-specific marker genes to WNT/mTORglutamine-dependent gene markers in Figures 8C-D and Figures S6A-E.

## **Ingenuity Pathway Analysis (IPA)**

Transcriptomics and metabolomics datasets were analyzed using IPA (Qiagen, Hilden, Germany) which has sophisticated algorithms and criteria to calculate predicted functional activation/inhibition of canonical pathways, diseases and functions, transcription regulators and regulators based on their downstream molecule expressions and inc., the contract of the cont https://www.qiagenbioinformatics.com/products/ingenuitypathway-analysis). For human GC microarray, Illumina identifiers (ILMN) were uploaded together with log<sub>2</sub>fold change, *p*-values and *q-*values (false discovery rates). A total of 47,323 transcripts was assigned to analysis. A total of 37,489 transcripts were mapped/9,834 transcripts unmapped by IPA. For RNA sequencing, Ensembl identifiers were uploaded together with log<sub>2</sub>-fold change, *p*-values and q-values. A total of 54,460 transcripts was assigned to analysis. A total of 53,735 was mapped/725 unmapped by IPA. For mouse GC microarray, ILMN were aligned together with log<sub>2</sub>-FC and q-values before uploaded in IPA. A total of 12,519 transcripts was loaded, a total of 11,773 transcripts was mapped/746 unmapped in IPA. For metabolomics, HMDB and KEGG identities were aligned together with fold changes, expressed *p*-values and *q*-values. A total of 343 metabolites were uploaded for downstream analysis in IPA and 252 metabolites were mapped by IPAThe data was subjected to a metabolomics expression analysis using HMDB or KEGG as identifier type. One-way ANOVA was used between groups. Fold changes were inverted before IPA analyzes. Thus, a molecule with 0.5-fold change was negatively inverted (-1/0.5) to -2.0.

Regulatory z-scores for canonical pathways that overlapped with our experimental data were calculated using the formula described previously (Krämer et al., 2014). To generate the network of up- or down-regulated genes, custom-made molecular networks were developed based on information contained in the IPAs knowledge base. Networks of these genes were then algorithmically generated based on their interrelationships. Filtering of datasets included species, *p*-value cut-off and/or *q*-value cut-offs. Molecular networks and canonical pathways were algorithmically constructed based on known connectivity and relationships among metabolites and genes/proteins using IPAs knowledge base. The significance of the association between the dataset molecules and the canonical pathways was measured by Fischer's exact test that was used to calculate a *p*-value determining the probability that the association between the genes in the dataset and the canonical pathway by chance alone. Z-scores were calculated in IPA based on the dataset's correlation with the activated state. Negative z-scores indicate a decrease in activity, positive z-scores indicate an increase in activity. Canonical pathways were identified using statistical cut-offs at *p*<0.05 and/or *q*<0.05.

#### *In silico* **experiment**

Signaling pathways of WNT/β-catenin and mTOR were constructed based on the transcriptomic data of INS-GAS mice and were then entered into the "Pathway" module of the IPA to obtain the nodes in every corresponding signaling pathway. The expression data from INS-GAS *vs.* FVB mice (Mouse GC *vs.* WT) was compared to all genes in the pathways. Nodes were added as entries into the "My list"-function and all entries in the list were added to the "My pathway" in IPA. My pathway was used to produce a network of nodes/genes from the WNT and mTOR signaling pathways that matched with our experimental data from INS-GAS *vs.* FVB. The build-tool was used to connect nodes using edges, i.e. relationships including both direct and indirect interactions like chemical-protein interactions, ubiquitination, molecular cleavage, translocation, localization, phosphorylation, expression, protein-protein interactions, activation, regulation of binding, inhibition, membership, reaction, protein-DNA interactions, transcription and modification. The Canonical Pathway overlay-tool was used to arrange the entries into two clusters based on pathway. Next, the molecule activity predictor (MAP)-function was used to predict activation/inhibition between the nodes in the network. The *in silico* tool was employed to predict effects on the network after gene inhibition. Categorical values were set to each gene/node using a semiquantitative method to quantify the color-change resulting from *in silico* inhibition. Dark blue colored nodes were represented by -2, light blue as -1, white as 0, light orange as +1 and dark orange as +2. Values are represented of n=7-14 experiments per inhibition node/gene.

### **Upstream regulator analysis**

Ingenuity pathway analysis (IPA, QIAGEN) was used to perform upstream analysis of the transcriptomics datasets based on the literature and the Ingenuity Knowledge Base. The analysis examines how many known targets of the upstream regulators are present in the dataset. An overlap *p*-value is computed based on significant overlap between genes in the dataset and known targets regulated by the transcriptional regulator. The activation z-score algorithm is used to make predictions. In mouse GC, 144 regulators were found to be activated (z-score>2, p<0.05) based on the expression levels of target molecules in the datasets. The overlay-tool in the "My pathway" module was used to cluster the activated regulators into canonical pathways. Next, upstream regulators of interest were added into custom-made pathways in the Path Designer-tool and relationship-types between upstream regulator and target molecule were added.

## **tSNE plot of metabolic gene expression according to single-cell atlas**

Available data on a single-cell transcriptome network of gastric premalignant and early gastric cancer in patients was utilized (PMID: 31067475), including 13 biopsies from 9 patients: 3 mild superficial gastritis (NAG), 3 chronic atrophic gastritis (CAG), 6 intestinal metaplasia (IM), and 1 early gastric cancer (EGC). Single-cell data were processed using Seurat v3 (doi.org/10.1016/j.cell.2019.05.031) and normalized for each of the 13 samples independently. The functions FindIntegrationAnchors, IntegrateData, ScaleData and RunPCA with default parameters were used. Cells with number of expressed genes lower than 400 or larger than 7000 and 20% or more of UMIs mapped to mitochondrial or ribosomal genes were removed. 50 PCs were utilized to visualize single-cell atlas with a tSNE plot. The expression levels of marker genes in mouse GC *vs.* WT for each representative cell type were analyzed. Marker genes were identified by differential expression analysis with the threshold as fold change  $> 1.5$  and FDR  $< 0.01$ .

## **Statistics**

Values are expressed as means  $\pm$  SEM or SD and statistical methods are shown in the figure legends.

#### **Supplemental references**

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