Abstract

**Purpose:** Gastric cancer may be subdivided into 3 distinct subtypes—proximal, diffuse, and distal gastric cancer—based on histopathologic and anatomic criteria. Each subtype is associated with unique epidemiology. Our aim is to test the hypothesis that these distinct gastric cancer subtypes may also be distinguished by gene expression analysis.

**Experimental Design:** Patients with localized gastric adenocarcinoma being screened for a phase II preoperative clinical trial (National Cancer Institute, NCI #5917) underwent endoscopic biopsy for fresh tumor procurement. Four to 6 targeted biopsies of the primary tumor were obtained. Macrodissection was carried out to ensure more than 80% carcinoma in the sample. HG-U133A GeneChip (Affymetrix) was used for cDNA expression analysis, and all arrays were processed and analyzed using the Bioconductor R-package.

**Results:** Between November 2003 and January 2006, 57 patients were screened to identify 36 patients with localized gastric cancer who had adequate RNA for expression analysis. Using supervised analysis, we built a classifier to distinguish the 3 gastric cancer subtypes, successfully classifying each into tightly grouped clusters. Leave-one-out cross-validation error was 0.14, suggesting that more than 85% of samples were classified correctly. Gene set analysis with the false discovery rate set at 0.25 identified several pathways that were differentially regulated when comparing each gastric cancer subtype to adjacent normal stomach.

**Conclusions:** Subtypes of gastric cancer that have epidemiologic and histologic distinctions are also distinguished by gene expression data. These preliminary data suggest a new classification of gastric cancer with implications for improving our understanding of disease biology and identification of unique molecular drivers for each gastric cancer subtype. *Clin Cancer Res; 17(9); 2693–701. ©2011 AACR.*

Introduction

Gastric cancer is the second most common cause of cancer-related mortality worldwide with 700,349 deaths annually, and is the third most common malignancy worldwide with 974,000 new cases in the year 2000 (1). Gastric cancer has been considered a single heterogeneous disease with several epidemiologic and histopathologic characteristics; for the purposes of medical management, gastric cancer is treated in a uniform fashion, without regard to subtype. Pathologically, gastric adenocarcinoma may be distinguished according to the Lauren’s classification as intestinal, diffuse, or mixed subtypes (2). Epidemiologically, intestinal gastric cancer, particularly of the antrum, is strongly associated with chronic inflammation (i.e., atrophic gastritis; refs. 3, 4) often as a consequence of chronic infection with *Helicobacter pylori* (5, 6). Conversely, inflammation is characteristically absent in the development of Lauren’s diffuse type gastric cancer, particularly when as a result of a germline mutation in *CDH1* (7). Anatomically, proximal gastric cancer may be classified as a third type of gastric cancer, as tumors of the gastric cardia/gastroesophageal junction (GEJ), for which inflammation of a different type (i.e., chronic gastric acid/bile reflux) may be the driving force for carcinogenesis (8, 9). Proximal/GEJ tumors are also usually not diffuse in histology, similar to distal nondiffuse gastric cancer.

As noted above, at present, the histopathologic, anatomic, and epidemiologic distinctions that subdivide this disease are not taken into account in the clinical management of the disease, for either initial potentially curative treatment or in palliation of advanced disease. For patients with metastatic disease, the available cytotoxic agents are applied indiscriminately to all disease subtypes, and with only modest success (reviewed in ref. 10). In other
Translational Relevance

Considering gastric cancer as one disease has limited the rational development of new prevention and therapeutic strategies. Herein, we examine cDNA expression data from gastric adenocarcinoma samples from 36 individual primary tumors to examine a new classification of gastric cancer. There is sufficient epidemiologic, histopathologic, anatomic, and now molecular evidence to categorize gastric cancer into 3 diseases—proximal nondiffuse, diffuse, and distal nondiffuse gastric cancer. Differentially expressed genes distinguish each subtype of gastric cancer from each other and from adjacent normal gastric mucosa. Our genomic classifier tightly groups the defined gastric cancer subtypes into specific individual clusters with high discrimination, suggesting that more than 85% of samples were classified correctly. In addition, gene set analysis identifies several differentially regulated pathways between individual gastric cancer subtypes. The ramifications of this classification are significant, including improving our understanding of unique molecular drivers of each gastric cancer type, aiding in the identification of novel biomarkers and targets of each disease, and ultimately helping to develop new treatment paradigms for each gastric cancer type.

Materials and Methods

Study population

From May 2003 to January 2006, we screened patients with gastric or gastroesophageal adenocarcinoma by endoscopic ultrasound, laparoscopy, CT scan, and positron emission tomography (PET) scan for enrollment in a National Cancer Institute (NCI)-sponsored neoadjuvant clinical trial of irinotecan and cisplatin chemotherapy followed by surgical resection (19). This protocol was reviewed and approved by the Institutional Review Board of Memorial Sloan-Kettering Cancer Center and by the NCI (#5917, NCT00062374). Written informed consent was obtained from each patient. All patients without evidence of metastatic disease on CT scanning underwent preoperative evaluation including endoscopic evaluation with ultrasound during which an endoscopic biopsy was carried out for the procurement of fresh tumor tissue for RNA extraction and analysis. All endoscopic tumor biopsies were carried out prior to initiation of any treatment for the malignancy.

Endoscopy and endoscopic biopsy

All patients underwent standard video endoscopy using the Olympus gastroscope GIF-160 (Olympus America). Targeted biopsies of the gastric mass, ulcer edge, or thickened folds were obtained using the Bard Precisor EXL–coated disposable biopsy forceps (Bard International). Four to 6 biopsies were carried out for each patient, with each biopsy usually measuring approximately 2 to 3 mm in diameter. On receiving the biopsy tissue from the endoscopic biopsy forceps, a small sample was placed immediately into buffered formalin (for histopathologic evaluation) or saline (for immediate freezing) while still in the endoscopy suite. The specimens in saline were immediately transported to the Tumor Procurement Laboratory where they were individually placed in optimum cutting temperature (OCT) media and frozen at −80°C. The time from obtaining the biopsy to OCT was less than 15 minutes. The formalin fixed samples were submitted to the Pathology Department for routine processing.

Specimen analysis for RNA processing

OCT-embedded biopsy samples were maintained below −20°C during processing. Frozen section slides were made from OCT-embedded biopsy samples and were then stained with H&E and reviewed by reference pathologists (L.T., D.S.K.). The presence or absence of invasive adenocarcinoma and the extent of malignant and nonmalignant cell involvement of the sample was recorded. We defined an adequate biopsy specimen as one having a proportion of at least 80% carcinoma nuclei. Macrodisssection was carried out in a specimen dissection chamber maintained at −20°C, using the marked H&E (hematoxylin and eosin) slide as a guide enabling us to remove OCT and nonmalignant tissue from the carcinoma. Following...
in vitro amplification, and labeling of cRNA were accomplished using the Stranded cDNA Synthesis Kit (Invitrogen). Synthesis, linear (depending on availability) was used for cDNA synthesis samples meeting this standard, 1.2 to 2 ratio of 1.8 to 2.0 were considered suitable for labeling. For samples meeting this standard, 1.2 to 2 ratio of 1.8 to 2.0 were considered suitable for labeling.

The quality of RNA was verified before labeling by analyzing 20 to 50 ng of each sample using the RNA 6000 NanoAssay and a Bioanalyzer and processing only if they contained 100% cancer on the H&E reference slides. In cases where individual patients had several biopsy samples that were suitable for RNA processing, the samples were combined for RNA processing. Throughout these tissue handling procedures, care was taken to use RNAase-free gloves and laboratory equipment to minimize contamination.

**RNA isolation, probe preparation, and microarray hybridization**

Total RNA was isolated from tumor specimens using RNeasy columns (Qiagen), and all samples were treated on the column with RNase-free DNase. The quality of RNA was verified before labeling by analyzing 20 to 50 ng of each sample using the RNA 6000 NanoAssay and a Bioanalyzer 2100 (Agilent). Samples with a 28S/18S ribosomal peak ratio of 1.8 to 2.0 were considered suitable for labeling. For samples meeting this standard, 1.2 to 2 μg of total RNA (depending on availability) was used for cDNA synthesis using an oligo-dT-T7 primer and the SuperScript Double-Stranded cDNA Synthesis Kit (Invitrogen). Synthesis, linear amplification, and labeling of cRNA were accomplished by in vitro transcription using the MessageAmp aRNA Kit (Ambion) and biotinylated nucleotides (Enzo Diagnostics). Ten micrograms of labeled and fragmented cRNA were then hybridized to the Human Genome U133A GeneChip (Affymetrix), which contained 22,215 oligonucleotide-based probe sets, at 45°C for 16 hours. Posthybridization staining, washing were processed according to manufacturer (Affymetrix) guidelines. Finally, chips were scanned with a high-numerical aperture and flying objective (FOL) lens in the GS3000 scanner (Affymetrix). The image was quantified using MAS 5.1 (MicroArray Suite, Affymetrix) with the default parameters for the statistical algorithm and all probe set scaling with a target intensity of 500.

**Definition of gastric cancer subtypes**

Three individual types of gastric cancer are strongly suggested by clinical and epidemiologic data (20). Using these characteristics, we defined the subtypes of gastric cancer histopathologically as follows:

1. **Proximal nondiffuse gastric cancer**—The bulk of the tumor (>80%) is located in the gastric cardia which may extend up to the GEJ and small portion of the distal esophagus. On histopathology, there is evidence of precursor glandular dysplasia or in situ carcinoma in the setting of chronic inflammation usually without atrophy. Tumor differentiation may range from well to poorly differentiated, but the pattern of tumor infiltration should not be entirely diffuse.

2. **Diffuse gastric cancer**—Tumor location may be anywhere in the stomach. On histopathology, there is no apparent gastritis, neither severe chronic nor atrophic. The pattern of infiltration is entirely diffuse without excessive extracellular mucin pools (colloid carcinoma is not included). There should not be any component of gland-forming intestinal-type carcinoma. The tumor is poorly differentiated signet ring cell type either with or without intracellular mucin.

3. **Distal nondiffuse gastric cancer**—The bulk of the tumor is usually located in the distal stomach and may extend up to the mid body of the stomach or down to the pylorus. On histopathology, there is evidence of chronic gastritis with intestinal metaplasia and a spectrum of glandular dysplasia and in situ carcinoma. The dominant pattern is a moderately differentiated and intestinal type carcinoma without or with minor components of poorly differentiated or dedifferentiated carcinoma.

Patients were assigned to an individual subtype of gastric cancer based on the histopathologic and anatomic definitions above and the expression arrays were analyzed.

**Bioinformatics**

The affymetrix (HG-133A/B) data set that contains 38 gastric tumor samples and 31 normal samples taken from stomach tissues adjacent to cancerous tissues was downloaded from Gene Expression Omnibus (www.ncbi.nlm.nih.gov/geo/ GSE accession number GSE13911).
D’Errico and colleagues (21). The GEO and MSKCC data sets contain 120 arrays that were processed and analyzed using the Bioconductor suite of tools in R-statistical language (www.bioconductor.org). All data were normalized using standard GeneChip Robust Multi-array Average functions with default parameters. Normalized data was clustered using hierarchical clustering based on Euclidian distance (hclust R-function) to verify that there is no batch effect between the 2 data sets (Supplementary Fig. S1). Indeed, normal and tumor samples were mostly grouped together and not according to their batch. We next removed the tumor samples from the GEO data set as there is no information on their gastric cancer type, resulting in 73 samples for further analysis (37 normal stomach and 36 gastric cancers). Probes that are present in at least half of the normal samples or half of the tumor samples were retained for further analysis (23,828 probes). For analysis at the gene level, multiple probes corresponding to 1 gene were averaged, and probes without a gene symbol annotation were removed leaving 8,740 unique genes. Differential expression analysis was done using limma R-package. Probes/gene were declared differentially expressed with fold-change cutoff greater than 2, and false discovery rate (FDR) = 0.01 (i.e., up to 1% of the significant differences are expected to be false positives). For subtype classification only tumor samples from Memorial Sloan-Kettering Cancer Center (MSKCC) were considered. To build the features (genes) that separate different subtypes of gastric cancer, the data set was filtered and genes with fold changes greater than log 2(1.5) in any pair-wise comparison (‘proximal nondiffuse’ vs. ‘distal nondiffuse,’ ‘proximal nondiffuse’ vs. ‘proximal diffuse,’ ‘proximal nondiffuse’ vs. ‘distal nondiffuse’) were considered. An additional condition of (unadjusted) the value of $P \leq 0.005$ in any of 3 comparisons was also used. The resulting data set used for the learning classifier has data for 785 genes and 36 tumor samples. Results are similar if probes are selected based on other criteria. To build the classifier, we opted for a supervised classification algorithm that implements regularized regression with the optimal scoring algorithm (22, 23). This algorithm includes a procedure for finding gene signatures by ranking genes based on the fitted regression models. This methodology includes principal components, partial least squares, and ridge regression models. It has been applied to several microarray studies in cancer (22) and it is coded in R-package pdcmodels. We used ridge regression methodology for the classification. In addition, we used gene set analysis (GSA; ref. 24) for an exploratory analysis to determine if the members of a given gene set were concordantly up- or downregulated between gastric cancer subtypes and normal stomach. GSA was run with default parameters and number of permutations = 500 to compute $P$ values, with $FDR = 0.25$ as suggested in the GSEA manual (http://www.broadinstitute.org/gsea/doc/GSEAUserGuideFrame.html). The gene set enrichment analysis was carried out using the Molecular Signature Database (MSigDB) v2.5 released April 7, 2008. Functional analysis of differentially expressed genes was done using the DAVID tool (http://david.abcc. ncbi.nlm.nih.gov/home.jsp) using all human genes as a background set (25).

Results

Patient and tumor characteristics

Between November 2003 and January 2006, 57 patients with localized gastric adenocarcinoma based on CT scan imaging of the chest, abdomen, and pelvis underwent endoscopic biopsy and tissue biopsy. Of these 57 patients, tumor biopsy samples from 41 patients (72%) were adequate for RNA processing and analysis. Subsequent staging procedures [i.e., laparoscopy and [18F]-fluorodeoxyglucose (FDG)-PET scan] identified occult metastatic disease in 5 patients, leaving 36 patients with nonmetastatic gastric cancer in the final study population (see Table 1). The majority of patients had locally advanced, poorly differentiated tumors, with approximately 60% of the cases node positive on preoperative evaluation. Proximal nondiffuse gastric cancers ($n = 12$) were more commonly Lauren’s histological intestinal type. Diffuse gastric cancers ($n = 10$) were more commonly anatomically located in the body or distal stomach, were uniformly poorly differentiated and Lauren’s histology by definition. Distal nondiffuse gastric cancers ($n = 14$) were predominantly intestinal and mixed Lauren’s histology.

Gastric cancer subtypes and normal stomach

Our aim was to examine whether genomic signatures would significantly differentiate gastric cancer subtypes that were assigned solely based on anatomic and histopathologic knowledge. We first compared gastric cancer versus normal stomach. We examined expression data from 2 independent data sets (MSKCC data: 36 tumors, 10 adjacent normal stomach and D’Errico data: 38 tumors, 31 adjacent normal stomach; ref. 21). Tumor samples and normals from both data sets cluster primarily according to their malignancy status (normal, tumor), and not according to the data set (Supplementary Fig. S1). Then, when evaluating MSKCC gastric tumors (i.e., study population that was annotated according to gastric cancer subtype) versus normal adjacent stomach, we identified a large number of genes that were differentially expressed in each type of gastric cancer versus normal (limma analysis, FC cutoff = 2, $FDR = 0.01$; Fig. 1). We noted that although there is a large overlap in the genes differentially expressed between gastric cancer types and normal, a significant number of genes uniquely differentiate each subtype of gastric cancer from normal stomach. Direct comparison of tumor subtypes (full data set) yielded 3 genes (with $FDR = 0.05$) that are differentially expressed between proximal nondiffuse and diffuse gastric cancer subtypes, including PSCA (prostate stem cell antigen) and PGA3 (pepsinogen A) which were both downregulated more than 20-fold in proximal nondiffuse gastric cancer when compared with diffuse gastric cancer, and TRIM32
which was upregulated (>2-fold) in proximal nondiffuse gastric cancer.

We then carried out functional categories analysis on groups of differentially expressed genes using the DAVID tool. Genes that were upregulated in all 3 gastric cancer types versus normal are significantly enriched in many typical cancer-related categories including cell cycle, cell proliferation, cell adhesion, platelet-derived growth factor binding, and EGF domain, whereas genes that are downregulated in all gastric cancer types versus normal are enriched in digestion, disease mutation, and lipid metabolism. Similarly, genes that are upregulated in proximal and distal nondiffuse gastric cancer (but not diffuse gastric cancer) are enriched in numerous cell cycle and mitosis-related categories, as well as p53 signaling pathways, whereas genes downregulated in both nondiffuse gastric cancer subtypes are enriched in digestion, drug metabolism, and response to various stimuli (nutrient levels, hormone stimulus, organic substance).

### Table 1. Patient and tumor characteristics

<table>
<thead>
<tr>
<th>Patient characteristic</th>
<th>Total population = 36</th>
<th>Proximal nondiffuse (n = 12)</th>
<th>Diffuse (n = 10)</th>
<th>Distal nondiffuse (n = 14)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age</td>
<td>59 (25–74)</td>
<td>62 (33–74)</td>
<td>59 (25–70)</td>
<td>58 (52–71)</td>
</tr>
<tr>
<td>Gender</td>
<td></td>
<td>8 M/4 F</td>
<td>2 M/7 F</td>
<td>7 M/7 F</td>
</tr>
<tr>
<td>Histopathology</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Differentiation</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Mod-poor</td>
<td>12</td>
<td>6</td>
<td>0</td>
<td>6</td>
</tr>
<tr>
<td>Poorly</td>
<td>24</td>
<td>6</td>
<td>10</td>
<td>8</td>
</tr>
<tr>
<td>Lauren’s</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Diffuse</td>
<td>14</td>
<td>0</td>
<td>10</td>
<td>0</td>
</tr>
<tr>
<td>Intestinal</td>
<td>19</td>
<td>9</td>
<td>0</td>
<td>10</td>
</tr>
<tr>
<td>Mixed</td>
<td>8</td>
<td>3</td>
<td>0</td>
<td>4</td>
</tr>
<tr>
<td>Location</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>GEJ/proximal</td>
<td>15</td>
<td>12</td>
<td>3</td>
<td>0</td>
</tr>
<tr>
<td>Body/distal</td>
<td>21</td>
<td>0</td>
<td>7</td>
<td>14</td>
</tr>
<tr>
<td>Stage at EGD</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Node positive</td>
<td>23 (64%)</td>
<td>9 (69%)</td>
<td>5 (56%)</td>
<td>9 (64%)</td>
</tr>
<tr>
<td>Early (T1–T2, N0, M0)</td>
<td>7</td>
<td>0</td>
<td>2</td>
<td>5</td>
</tr>
<tr>
<td>Locally advanced (T3–T4 or N⁺)</td>
<td>29</td>
<td>12</td>
<td>8</td>
<td>9</td>
</tr>
</tbody>
</table>

![Discriminant Plot for predicted classes](image)

Figure 2. Discriminant plot for gene signatures showing that the RNA expression signatures of samples from different gastric cancers are quite well separated and very tightly grouped into 3 types of gastric cancer (Supplementary Table S1): proximal nondiffuse (type 1), diffuse (type 2), and distal nondiffuse (type 3). This sample classification was done using pdmclass R-function, using ridge regression. The leave-one-out cross-validation error is 0.14, which implies that more than 85% samples are classified correctly. The output from pdmclass cross-validation confusion matrix is provided to the right of the discriminant plot.

### Gene expression analysis and development of molecular classification

We next focused on identifying gene signatures that can be used to classify gastric cancer subtypes. We used ridge regression method on a smaller set of genes (785 genes; see Methods; refs. 22, 23).

The classifier built on these genes separates the 3 subtypes quite well and the samples are tightly grouped into 3 distinct groups (Fig. 2). The leave-one-out cross-validation error is 0.14 which implies that more than 85% samples are classified correctly. In our patient population of 36 patients, 29 tumors were locally advanced (T3 or greater, or N⁺). When limiting the analysis to these 29 cases, the leave-one-out cross-validation error is 0.13, implying that still more than 85% of locally advanced samples are classified correctly. The top genes that separate proximal nondiffuse and diffuse gastric cancer include PGA3, PSCA,
versus normal stomach. This pathway contains several translation of the KEGG HSA05222 (small lung cancer) pathway Conversely, each gastric tumor subtype shows downregu-

nondiffuse gastric cancers had a number of upregulated with FDR of underlying biological processes. Table 2 provides the list of subtype comparisons. Pathway analysis may be indicative of underly

gene expression. This was the case when examining all cases (both early and advanced) and also when limiting the cases to advanced disease. One gene of interest that significantly distinguished proximal nondif-
fuse gastric cancer from diffuse gastric cancer, PSCA, has already been implicated in gastric cancer. An intrinsic polymorphism (rs2294008) in PSCA, resulting in reduced PSCA expression, is significantly associated with increased risk for diffuse gastric cancer compared with an intestinal subtype (virtually all distant nondiffuse) in a Japanese population (28, 29). Expression of another gene, PLA2G2A, a secreted phospholipase, has prognostic sig-

nificant in gastric cancer. Specifically, tumors expressing high levels of PLA2G2A have improved survival compared with patients with low PLA2G2A expressing tumors (30) Recently, it was shown that PLA2G2A is a target of Wnt/β-catenin signaling in gastric cancer cells, and is associated with negative regulation of genes associated with invasion and metastasis (31). Using GSA, a number of pathways were either up- or downregulated when individual gastric cancer subtypes were compared with normal stomach. For example, the glycolysis pathway is upregulated in proximal and distal nondiffuse gastric cancers relative to diffuse gastric cancer. Glycolysis is the process of converting glucose into pyruvate and generating small amounts of ATP. It is a central pathway that produces important precursor metabolites, and may explain the increased glucose metabolism of many cancers [i.e., Warburg effect (32)], and is commonly linked with FDG avidity for PET scanning. Consistent with this finding, diffuse gastric cancers (i.e., those that did not show upregulation of glycolysis on GSA) are commonly FDG non-Avid, unlike their nondiffuse counterparts.

Oncogenic pathways have been used previously to iden-
tify pathway signatures in malignancy, and similar to this report, to identify pathway signatures in subtypes of malign-
nancy such as breast cancer (33, 34). Ooi and colleagues examined a subset of oncogenic and tumor suppressor pathways in gastric cancer, including the RAS pathway (35, 36) which was identified in our GSA as downregulated in proximal nondiffuse gastric cancer when compared with diffuse gastric cancer. These investigators suggested that combinations of several pathways may provide greater predictive value for patient outcomes than individual path-

ways themselves (35). Specifically, they noted 3 pathways were dysregulated in more than 70% of gastric cancers: proliferation/stem cell; NF-kB, and Wnt/β-catenin. They validated the pathways in gastric cancer–derived cell lines (35); however, the location of the primary tumor was not included in their analysis—that is, whether or not the

When comparing gastric cancer subtypes to adjacent normal, we identified a significant number of genes that uniquely distinguished individual gastric cancer subtypes from normal stomach. These data provide support for the hypothesis that gastric cancer subtypes may be distin-
guished molecularly.

Then, using supervised classification, we show a greater than 85% ability to successfully distinguish gastric cancer subtypes by gene expression. This was the case when


gastric cancer from distal nondiffuse gastric cancer include ABCA8 (≥4-fold), HMBOX1, COCH, S100A12, CYP2J2 (upregulated in diffuse vs. distal nondiffuse), and IFI44L (4-fold), HOXA9, MSLN, and ENPP4 (downregulated in diffuse vs. distal nondiffuse).

Gene set analysis

In addition, we applied the GSA tool to carry out exploratory pathway analysis by comparing each gastric cancer subtype to normal, as well as by carrying out direct subtype comparisons. Pathway analysis may be indicative of underlying biological processes. Table 2 provides the list of pathways that were either upregulated or downregulated with FDR = 0.25. We observed that proximal and distal nondiffuse gastric cancers had a number of upregulated pathways including glycolysis and gluconeogenesis when compared with normal stomach, whereas no pathways were identified to be upregulated in diffuse gastric tumors. Conversely, each gastric tumor subtype shows downregulation of the KEGG HSA05222 (small lung cancer) pathway versus normal stomach. This pathway contains several tumor suppressors, including p53, PTEN, RB, and FHIT.

Discussion

Gastric cancer is a heterogeneous disease with differences in epidemiology and histopathology that, when coupled with anatomic location, may be distinguished into at least 3 different cancers (20). We explored this hypothesis by examining the gene expression of individual gastric cancer subtypes in a training set, carrying out a comparison with the expression analysis of adjacent normal stomach as well as among individual gastric subtypes. We found that individual diseases arising from the stomach defined a priori, namely “proximal nondiffuse,” “diffuse,” and “distal nondiffuse” gastric cancer, have distinct gene expression. These findings from our training set are both clinically consistent and have significance in the context of current clinical management of gastric cancer. For example, 2 large global studies evaluating new cytotoxic and biologic therapies have failed to show a survival benefit over the standard of care (26, 27). The investigators suggested that this may be partially due to failure to appreciate biological differences in gastric cancer and how these differences may affect response to treatment.
different pathways were associated with proximal, diffuse, or distal gastric cancer subtypes. Similarly, in another gene expression analysis, investigators identified 3 subgroups of cancer as "tumorigenic," "reactive," and "gastric like" (37). In their analysis, there was no association between intestinal and diffuse Lauren classification, or between tumor sites. Notably, both studies suggest gastric cancer may be subdivided genomically, with different prognoses, independent of stage (35, 37). However, no study to date has incorporated epidemiological and histopathologic data together with anatomic location, as we have done, to define subtypes of gastric cancer. Based on epidemiology and

| Table 2. Pathways and a priori defined sets of genes that show statistically significant, concordant differences between each gastric cancer subtype and normal stomach (using GSA with FDR = 0.25) |
|---------------------------------|---------------------------------|
| **Significant pathways that differentiate "proximal nondiffuse" vs. normal stomach** | **Downregulated in "proximal nondiffuse"** |
| **Bile acid biosynthesis** | HSA05222_SMALL_CELL_lung cancer |
| HSA00624_1_AND_2_METHYLNAPHTHALENE_DEGRADATION | |
| HSA00641_3_CHLOROACRYLIC_ACID_DEGRADATION TYROSINE_METABOLISM GLUCONEOGENESIS GLYCEROLIPID_METABOL GLYCOLYSIS HSA00561_GLYCEROLIPID_METABOLISM HSA00980_METABOLISM_OF_XENOBOTICS_BY_CYTOCHROME_P450 HSA00010_GLYCOLYSIS__AND_GLUCONEOGENESIS | |
| **Molecular functions** | METALLOENZYMEEPTIDASE_ACTIVITY* |
| ATPASE_ACTIVITY__COUPLED_TO_MOVEMENT_OF_SUBSTANCES | RNA_DEPENDENT_ATPASE_ACTIVITY |
| HYDROLASE_ACTIVITY | ATP_DEPENDENT_ATPASE_ACTIVITY |
| PRIMARY_ACTIVE_TRANSMEMBRANE | |
| ACTIVE_TRANSMEMBRANE | |
| HORMONE_ACTIVITY | |
| **Significant pathways that differentiate "diffuse" vs. normal stomach** | **Upregulated in "diffuse"** |
| HSA04512_ECM_RECEPTOR_INTERACTION | |
| HSA05222_SMALL_CELL_lung cancer | |
| **Significant pathways that differentiate "distal nondiffuse" vs. normal stomach** | **Upregulated in "distal nondiffuse"** |
| GLYCEROLIPID_METABOLISM HSA00010_GLYCOLYSIS_AND_GLUCONEOGENESIS HSA00624_1_AND_2_MET HYLNAPHTHALENE_DEGRADATION BILE_ACID_BIOSYNTHESIS IS GLUCONEOGENESIS GLYCOLYSIS PROPANOATE_METABOLISM TYROSINE_METABOLISM VALINE_LEUCINE_AND_ISOLEUCINE_DEGRADATION HSA00280_VALINE_LEUCINE_AND_ISOLEUCINE_DEGRADATION HSA00640_PROPANOATE__METABOLISM HSA00640_3_CHLOROACRYLIC_ACID_DEGRADATION | |
| **Biologic processes** | REGULATION_OF_DNA_REPLICATION |
| COVALENT_CHROMATIN_MODIFICATION HISTONE_MODIFICATION | REGULATION_OF_MITOTIC_CELL_CYCLE |
| **Molecular functions** | |
| ATPASE_ACTIVITY__COUPLED_TO_MOVEMENT_OF_SUBSTANCES | |
| HYDROLASE_ACTIVITY__ACTING_ON_ACID_ANHYDRIDES__CATALYZING_TRANSMEMBRANE_MOVEMENT_OF_SUBSTANCES | |
| PRIMARY_ACTIVE_TRANSMEMBRANE_TRANSPORTER_ACTIVITY | |
| ATPASE_ACTIVITY__COUPLED_TO_TRANSMEMBRANE_TRANSPORTER_ACTIVITY | |
| PRIMARY_ACTIVE_TRANSMEMBRANE_MOVTION_OF_IONS COENZYME_BINDING STEROID_BINDING LIPID_BINDING | |
pathology, we proposed a division of gastric cancer into 3 distinct types of gastric cancer (20). Chronic inflammation (e.g., from H. pylori infection) is required for the development of distal gastric cancer, usually intestinal type (38), and a diet high in fruits or vegetables is protective for this type of gastric cancer. Proximal gastric cancer is mostly associated with obesity and gastroesophageal reflux disease (9), perhaps causing inflammation via different pathways than in distal nondiffuse gastric cancer, whereas diffuse gastric cancer does not currently have established environmental or clinical risk factors (20). Our genomic analysis data confirm a clear molecular distinction in these types of gastric cancer. The value of this classification may be shown even with currently defined biomarkers in gastric cancer, namely Her-2-neu. This gene is amplified or overexpressed in approximately 25% of gastric cancer cases, and overexpression confers sensitivity to Her2-targeted therapy, and importantly a significant survival advantage when patients with Her2 positive tumors are treated with trastuzumab and chemotherapy (39). Her2 amplification or overexpression is not uniform across gastric cancer subtypes, most prevalent in proximal or GEJ gastric cancer (~30% Her2 positivity rate) and least prevalent in diffuse type gastric cancer (~5% Her2 positivity rate). Assessment of Her2 positivity rates, therefore, depend entirely on the constituent population studied, and will be higher in areas were proximal gastric cancers prevail, and less frequent where diffuse gastric cancers prevail.

These data are parallel to emerging analyses in other malignancies. For example, recent approaches of molecular classification of breast cancer have identified 3 distinct subclasses of breast cancer with both biologic and prognostic significance. These subclasses are defined as estrogen receptor (ER) and/or progesterone receptor (PR) positive tumors, HER2-amplified tumors, and ER/PR/HER2 (triple) negative tumors. The 3 breast cancer subtypes have been reproducibly identified by gene expression profiling in multiple breast cancer cohorts and exhibit consistent prognostic significance (11, 12). Clinical implications of this subclassification of breast cancer include the development of therapeutic strategies, such as the use of PARP inhibition for triple negative disease (14, 40), as well as potentially significant racial and ethnic ramifications, such as the identification of triple negative breast tumors as more prevalent in premenopausal African American woman (39%; 41).

In summary, for the first time to our knowledge, we have shown that malignancies arising from the stomach that have epidemiologic and histologic distinction can also be distinguished by genomic/molecular analysis. These data have significant ramifications. Our analysis suggests that (i) different types of cancers arise from the stomach, (ii) there likely exist unique molecular drivers that may be identified among specific genetic pathways that distinguish each disease, and (iii) the presence of different biomarkers and therapeutic targets for each disease is also likely. We are conducting a separate validation study to confirm the classification error estimate of our classifier. However, these data provide corroborating molecular evidence of a new classification for gastric cancer. Ultimately, such distinction will allow us to begin to manage each of these diseases differently and uniquely. As we improve our understanding of gastric cancer heterogeneity and its clinical consequences, our hope is to improve patient outcomes with improved prevention, screening, and treatment options, using distinct biologic subtypes for improved application of targeted therapies.

Disclosure of Potential Conflicts of Interest

No potential conflicts of interest were disclosed.

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