ORIGINAL ARTICLE

Signatures of tumour immunity distinguish Asian and non-Asian gastric adenocarcinomas

Suling J Lin,1 Johann A Gagnon-Bartsch,2 Iain Beehuat Tan,3 Sophie Earle,4 Louise Ruff,4 Katherine Pettinger,4 Bauke Ylstra,5 Nicole van Grieken,5 Sun Young Rha,6 Hyun Cheol Chung,6 Ju-Seog Lee,7 Jae Ho Cheong,8 Sung Hoon Noh,8 Toru Aoyama,9 Yohei Miyagi,10 Akira Tsuburaya,11 Takaki Yoshikawa,9 Jaffer A Ajani,12 Alex Boussioutas,13,14 Khay Guan Yeoh,15,16 Wei Peng Yong,17 Jimmy So,18 Jeeyun Lee,19 Won Ki Kang,19 Sung Kim,20 Yoichi Kameda,21 Tomio Arai,22 Axel zur Hausen,23,24 Terence P Speed,2,25,26 Heike I Grabsch,4,23,24 Patrick Tan1,27,28,29

ABSTRACT

Objective Differences in gastric cancer (GC) clinical outcomes between patients in Asian and non-Asian countries has been historically attributed to variability in clinical management. However, recent international phase III trials suggest that even with standardised treatments, GC outcomes differ by geography. Here, we investigated gene expression differences between Asian and non-Asian GCs, and if these molecular differences might influence clinical outcome.

Design We compared gene expression profiles of 1016 GCs from six Asian and three non-Asian GC cohorts, using a two-stage meta-analysis design and a novel biostatistical method (RUV-4) to adjust for technical variation between cohorts. We further validated our findings by computerised immunohistochemical analysis on two independent tissue microarray (TMA) cohorts from Asian and non-Asian localities (n=665).

Results Gene signatures differentially expressed between Asians and non-Asian GCs were related to immune function and inflammation. Non-Asian GCs were significantly enriched in signatures related to T-cell biology, including CTLA-4 signalling. Similarly, in the TMA cohorts, non-Asian GCs showed significantly higher expression of T-cell markers (CD3, CD45R0, CD8) and lower expression of the immunosuppressive T-regulatory cell marker FOXP3 compared to Asian GCs (p<0.05). Inflammatory cell markers CD66b and CD68 also exhibited significant cohort differences (p<0.05). Exploratory analysis revealed a significant relationship between tumour immunity factors, geographic locality-specific prognosis, and postchemotherapy outcomes.

Conclusions Analyses of >1600 GCs suggest that Asian and non-Asian GCs exhibit distinct tumour immunity signatures related to T-cell function. These differences may influence geographical differences in clinical outcome, and the design of future trials particularly in immuno-oncology.

INTRODUCTION

Gastric cancer (GC) is a leading cause of global cancer mortality, with high incidence rates in Asia and parts of Latin America.1 Survival outcomes differ across geographical regions, with overall 5-year survival rates being 10%–15% in North America2 and 45%–50% in East Asia.3,4 These differences cannot be simply explained by improved early diagnosis in Asian countries, as they persist even after stratifying for disease stage.5 It has been suggested that these

Significance of this study

What is already known on this subject?

▸ Gastric cancer patients from different geographic localities are known to show different clinical outcomes.
▸ Conventionally, these clinical outcome differences have been attributed to differences in clinical management and disease stage.
▸ However, recent phase III clinical trials with standardised treatment arms still demonstrated differences in gastric cancer outcomes by geographic locality.

What are the new findings?

▸ Our investigation of >1600 gastric cancers revealed that tumour immunity signatures differ significantly between Asian and non-Asian gastric cancers.
▸ Non-Asian gastric cancers were associated with enrichment of tumour infiltrating T-cells as well as T-cell gene expression signatures, including CTLA-4 signalling.
▸ Exploratory analysis suggests that these tumour immunity differences may contribute to geographical differences in clinical outcome.

How might it impact on clinical practice in the foreseeable future?

▸ The design of future gastric cancer trials, particularly in immuno-oncology, should consider tumour immunity differences in patients from different geographic localities, as they may impact on treatment response and clinical outcomes.
differences may reflect geographic variability in clinical practice. However, Asian patients treated in Western countries still exhibit superior outcomes compared with Caucasians, albeit worse than patients from Asian registries in home countries. Recently, a potentially deeper relationship between geographic locality and GC clinical outcome has emerged from international Phase III randomised trials, where the study populations and treatments are standardised across multiple countries. In the Avastin in Gastric Cancer (AVAGAST) trial evaluating bevacizumab in advanced GC, subgroup analysis revealed a survival benefit in non-Asians but not in Asians. Conversely, in the Lapatinib Optimization Study in ErbB2 (HER2) Positive Gastric Cancer (LOGiC) trial benefit from lapatinib was observed in Asians but not non-Asians. Importantly, both studies were ultimately deemed ‘negative’ in the overall study population. It is thus important to investigate whether Asian and non-Asian GCs represent distinct molecular entities. This information may prove vital in selecting appropriate study populations for future trials.

Asians and non-Asian populations may differ in their GC risk factors. For example, Asian Helicobacter pylori strains are enriched in cagPAI pathogenicity gene variants compared with non-Asian strains, which may modulate carcinogenicity. Germline polymorphisms (eg, IL-1B, IL-1RN) associated with GC risk also appear to differ across different populations. Regional variability in histopathological subtypes and tumour site locations have been reported, with intestinal subtype GC being particularly common in endemic countries while the incidence of diffuse subtype GC is less variable. However, somatic gene mutation and gene amplification rates in major cancer oncogenes such as KRAS, HER2 and FGFR2 appear to be similar in Asian and non-Asian GCs, suggesting that oncogenic pathways are broadly conserved between different populations. In summary, conflicting and overall very little data exist regarding whether Asian and non-Asian GCs are molecularly distinct.

Gene expression profiling technologies can provide ‘molecular snapshots’ of global transcriptional programmes expressed in tumours, not only from cancer cells but also from other tumour-associated cell types, such as fibroblasts, blood vessels and immune cells. To date, only one report (involving only 11 patients) has directly compared gene expression profiles of 1016 primary GCs from nine studies comprising Asian and non-Asian localities. We discovered and validated immune signature differences at the gene expression level between Asian and non-Asian GCs. The discovered gene expression differences were corroborated by subsequent identification of quantitative differences in intratumoural immune cells in two independent Asian and non-Asian GC cohorts. Exploratory survival analyses revealed that these tumour immunity factors significantly influenced geographic locality-specific survival.

MATERIALS AND METHODS

GC patient cohorts
Nine GC expression cohorts were analysed, involving 1016 patients (890 Asians and 126 non-Asians; online supplementary table S1). Studies were approved by Local Research Ethics Committees.

Data preprocessing
Platform-specific preprocessing steps were applied for Affymetrix and cDNA microarrays (detailed in online supplementary materials and methods). Two datasets (YGC and SamsungMC) were received in normalised form from collaborators. For these latter datasets, only summarisation of multiple probes mapping to the same gene was performed. All preprocessing steps were conducted within the statistical computing language R environment version 2.15 (http://www.r-project.org).

Gene expression microarray analysis
Due to technical variation across different microarray cohorts, gene expression measurements are often influenced not only by biological factors of interest (eg, geographic locality), but also by technical factors—including known/observed unwanted factors (eg, batch of hybridisation), and unknown/unknown unwanted factors (eg, RNA degradation). We used RUV-4 to reduce the impact of these unwanted factors (see online supplementary materials and methods). Briefly, RUV-4 uses ‘negative control genes’ (genes assumed to be uninfluenced by the biological factor of interest) to help identify the unwanted factors, which are then incorporated into a standard linear regression. In this study, housekeeping genes were used as controls (see online supplementary table S2). However, RUV-4 is distinct from standard normalisation methods using housekeeping genes, where gene expression values are adjusted using weights derived from differences in housekeeping gene levels. Specifically, to identify differentially expressed genes, RUV-4 fits a linear model on the expression data, accounting for known and unknown technical factors identified. In the process, two-group comparisons using empirical Bayes moderated t-statistics are made (eg, between ‘Asian’ and ‘non-Asian’ cohorts). Multiple hypothesis testing corrections were achieved using the Benjamini and Hochberg algorithm. Genes with adjusted p<0.05 were considered to be differentially expressed. Ingenuity Pathway Analysis (IPA) and Gene-Set Enrichment Analysis (GSEA) was performed (online supplementary materials and methods) to identify expression modules and pathways distinct to Asian and non-Asian GCs. All top-ranked genesets were manually curated to confirm their accurate functional and pathway categorisations, and are presented in online supplementary table S3. For this study, GSEA was used largely for exploratory analyses, and instead of applying a stringent false discovery rate cut-off, we chose to analyse the top 10 genesets (ranked by normalised enrichment score) that were (1) associated with each of the localities and also (2) associated with low nominal p values (see online supplementary materials and methods).

Immunohistochemistry and other tissue analyses
Tissue microarrays (TMA) were constructed from an Asian GC cohort (Japanese; n=219) and a non-Asian cohort (Caucasian; n=446). Slides from resection specimens were reviewed by four senior gastrointestinal pathology histopathologists (HG, TA, YM and YK). A block with the highest tumour cell density was selected and marked for TMA construction. TMA sections were stained for immune cell markers CD66b (neutrophils), CD68 (pan-macrophages), CD45 (pan-leucocytes), CD3 (T-cells), CD8 (cytotoxic T-cells), CD45R0 (memory/activated T-cells) and FOXP3 (T-regulatory cells). TMA slides were subjected to automated image analysis as described previously. Wilcoxon Mann–Whitney tests were used to test whether particular markers were different between Japanese and Caucasian GCs, using tumour content-adjusted median immunohistochemical (IHC) values (significance p<0.05). Epstein-Barr virus (EBV) RNA in situ hybridisation and assessment of the mismatch...
Statistical analyses

Clinicopathological factors between Asian and non-Asian cohorts were compared using a two-tailed Welch’s t-test. Survival analyses between Asian and non-Asian GCs were performed using the Kaplan–Meier method, with 5-year overall survival as the outcome metric, and log-rank tests to assess significance. Univariate and multivariate analyses were performed using Cox regression and Wald tests used to assess significance. Year of surgery was added to the multivariate model in order to account for potential influence of tissue block age on antigen recovery. For analyses involving CD68/CD3 ratios, samples were categorised into ‘high’ (3rd/4th quantiles) and ‘low’ (1st/2nd quantiles) groups, and a generalised linear model, using a gamma distribution, was fitted with CD68/CD3 ratios and significant covariates as the predictor variables and 5-year overall survival time as the response variable. One-tailed Welch’s t-test was employed to test associations between CD68/CD3 ratios and geographic locality (significance threshold $p<0.05$). CD68 and CD3 profiles from expression microarray datasets were also grouped into quantiles and CD68/CD3 ratios computed for all samples. Samples with a CD68/CD3 ratio greater than two were classified as ‘high’ and the remaining as ‘low’ CD68/CD3 ratio. Datasets with approximately 10% or a greater number of ‘high’ CD68/CD3 ratio samples (6 datasets) were included in further survival analyses. Survival curves and hazards were found to cross in all datasets analysed, hence supremum version of log-rank test (Renyi-type test) was used to assess significance of difference between curves. Due to non-proportional hazards, restricted mean survival times were computed at 1-year, 5-year and 10-year intervals.

RESULTS

Baseline characteristics of Asian and non-Asian GC datasets

We assembled nine independent GC microarray cohorts comprising 1016 tumour gene expression profiles—six from Asian localities (ie, HK_cDNA, Korea_SY, SamsungMC, SGet1, SGet2, YGC; $n=890$), and three from non-Asian localities (ie, AMS_cDNA, AU_affy, LEEDS; $n=126$) (see online supplementary table S1). Most clinicopathologic parameters (eg, age, gender, stage etc.) were not significantly different between the Asian and non-Asian cohorts ($p>0.05$; table 1), with the exception of tumour location, where non-Asian cohorts had significantly more cases with tumours in the upper third of the stomach compared with Asian cohorts ($p=0.04$). However, in agreement with the published literature, 5-year survival outcomes were significantly better in the Asian cohorts (figure 1).

Two-stage meta-analysis

Prior to combining the cohorts (gene expression data) for comparative analyses, it was necessary to reduce study-specific technical variation confounded with geographic locality (see online supplementary figure S3A). This could not be achieved via conventional methods (eg, dummy regression analysis). Hence, we used a novel algorithm, RUV-4, to reduce study-specific effects in the gene expression data, while retaining geographic locality-specific differences (see online supplementary figure S3B). However, despite RUV-4 being able to reduce study-specific effects in the gene expression data, it cannot eliminate them entirely, and errors at the individual gene level can still persist even after correction. The potential presence of residual confounding prompted us to compare the Asian and non-Asian GCs at the gene signatures level (multiple genes) rather than single genes. We also employed a two-stage meta-analysis design to further reduce platform-specific differences (figure 2). In Stage 1, the analysis was restricted to ‘Affymetrix’ datasets ($n=299$) since examination of the mean profile pairs-plot for each dataset suggests that ‘Affymetrix’ datasets were most comparable (see online supplementary figure S4). In Stage 2, the assessment of differential gene signatures between Asian and non-Asian GCs identified in Stage 1 also showed similar differential enrichment in the noisier ‘non-Affymetrix’ datasets ($n=717$).

Distinct expression signatures in Asian and non-Asian GCs (Stage 1)

In Stage 1, we identified 2102 significantly differentially expressed genes between Asian and non-Asian GCs (corrected $p<0.05$, online supplementary table S4). We conducted gene signature analyses using two different methodologies. First, using Ingenuity Pathway Analysis (IPA), we observed a general theme of inflammatory and immunological disease among the top enriched ‘Diseases and Disorders’ ($p<0.05$) in tumours from both geographic localities (figure 3A, left). Within the major category ‘inflammatory disease’, Asian and non-Asian GCs shared several subdisease signatures (see online supplementary table S5). However, none of the constituent genes in the subdisease signatures were common between these GCs (see online supplementary table S5 and figure S5) suggesting that GCs from both localities have distinct inflammatory mechanisms. Supporting this idea, IPA analysis using ‘canonical signalling pathways’ revealed that only non-Asian GCs were associated with multiple signalling pathways related to T-cell biology (Fisher’s exact $p<0.05$; figure 3A, right).

Second, using Gene Set Enrichment Analysis (GSEA), we performed pathway analysis on ranked differentially expressed genes between Asian and non-Asian GCs, specifically querying genesets in the C2 (curated) and C7 (immunological) sections of the Molecular Signatures Database (MSigDB). Notably, because MSigDB genesets are based on their original data sources, their names may sometimes obscure their actual gene content (eg, ‘Rickman_Head_and_neck_Cancer_D’ and ‘Shedden_Lung_Cancer_Good_Survival_A12’ are immune-related genesets containing genes largely associated with immune function). Therefore, all enriched MSigDB genesets were manually curated to accurately reflect their specific biological processes (see online supplementary table S3).

Interrogation of C2 genesets revealed a majority (4/4) enrichment of immune-related genesets in non-Asian GCs (figure 3B; online supplementary table S6). Similarly, GSEA using C7 immunology-related genesets revealed a majority (6/8) of T-cell-related genesets enriched in non-Asian GCs, including signatures related to CD4, CD8 and memory T-cells (figure 3B; online supplementary table S7).

Enrichment of T-cell signatures in non-Asian GCs (Stage 2)

To confirm Stage 1, we tested whether non-Asian GCs profiled on non-Affymetrix arrays (see online supplementary table S4) were also enriched in T-cell gene signatures (Stage 2). Performing a similar analysis against the C2 genesets, immune signatures were again enriched in the non-Asian GCs (figure 3C, top). Likewise, enrichment of C7 T-cell-related genesets identified in Stage 1 (see online supplementary table S8) was also observed in Stage 2 non-Asian GCs (figure 3C, bottom). These results, derived from a larger series profiled on multiple


Stomach
Table 1  Clinical characteristics of expression microarray studies

<table>
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<th>AMS_cDNA (n=34)</th>
<th>AU_affy (n=68)</th>
<th>HK_cDNA (n=90)</th>
<th>Korea_SY (n=96)</th>
<th>SamsungMC (n=432)</th>
<th>SGset1 (n=152)</th>
<th>SGset2 (n=55)</th>
<th>YGC (n=65)</th>
<th>LEEDS (n=24)</th>
<th>p Value* (t test)</th>
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<td>280 (64.8)</td>
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<td>–</td>
<td>43 (44.8)</td>
<td>186 (43.1)</td>
<td>90 (59.2)</td>
<td>29 (52.7)</td>
<td>6 (9.23)</td>
<td>10 (41.7)</td>
<td>0.24</td>
</tr>
<tr>
<td>Others§</td>
<td>–</td>
<td>–</td>
<td>8 (8.33)</td>
<td>127 (29.4)</td>
<td>–</td>
<td>–</td>
<td>20 (30.8)</td>
<td>–</td>
<td>N.A.</td>
<td></td>
</tr>
<tr>
<td>Missing</td>
<td>1 (2.94)</td>
<td>–</td>
<td>1 (0.23)</td>
<td>3 (1.97)</td>
<td>2 (3.64)</td>
<td>4 (1.65)</td>
<td>–</td>
<td>–</td>
<td>–</td>
<td></td>
</tr>
</tbody>
</table>

*p-value from hypothesis testing between Asian and non-Asian cohorts.
†American Joint Committee on Cancer (AJCC) staging 6th edition.
‡Include whole stomach, anastomosis, multisite, etc.
§Other mixed signet-ring, mucinous, tubular, hepatoid etc. non-conventional classifications as provided by the original study investigators. These have not been reviewed, thus may not be 100% comparable between studies due to differing local practice.
different microarray platforms, suggest a global enrichment of T-cell expression signatures in non-Asian GCs.

**Immunohistochemistry supports T-cell enrichment in non-Asian GCs**

To validate the immune-related gene expression differences between Asian and non-Asian GCs, we performed IHC on TMA panels comprising Asian and non-Asian GCs. Using automated image analysis to investigate quantitative and qualitative differences in intratumoural inflammatory cells and T-cells, we analysed an independent cohort of 219 Japanese GCs, and a separate cohort of 446 Caucasian GCs. Among these patients, a total of 133 were treated with chemotherapy after surgery (127 were treated with a 5-fluorouracil (5FU)-related chemotherapy) (see online supplementary table S9).

Neutrophils and macrophages are common cells associated with inflammation. We observed that Japanese GCs had a significantly higher number of cells positive for neutrophil marker CD66b ($p=2 \times 10^{-05}$; figure 4A; online supplementary table S10), and the pan-leukocyte marker CD45 ($p=2 \times 10^{-03}$; online supplementary figure S6 and table S10). Conversely, Caucasian GCs were significantly enriched for the macrophage marker CD68 ($p=4 \times 10^{-04}$; figure 4A; online supplementary table S10).

Supporting the gene expression microarray results, the majority of T-cell markers were significantly enriched in Caucasian compared with Japanese GCs (T-cell markers: CD3: $p=4 \times 10^{-06}$; CD45R0: $p=9 \times 10^{-15}$; CD8: $p=6 \times 10^{-07}$; figure 4B; see online supplementary table S10). The exception was the immunosuppressive T-regulatory (Treg) cell marker FOXP3 ($p=6 \times 10^{-08}$; figure 4B; online supplementary table S10). These results confirm the two-stage gene expression analysis that Asian and non-Asian GCs are likely to have distinct immune-related components, especially a higher abundance of T-cell infiltration in non-Asian GCs.

**Tumour immunity may influence location-specific survival differences**

Continuing to focus on the TMA cohorts, we then explored whether geographic location-specific survival differences might be influenced by tumour immunity factors. We performed a univariate Cox regression analysis of 5-year overall survival with geographic locality. Subsequently, we fitted the univariate model with other factors (eg, cancer stage, age, CD3 status etc.), in turn, to assess whether these factors might influence the adjusted HR for locality effects. We found that besides ‘age’, only the immune cell markers (ie, CD3, CD68, CD66b, CD8, CD45R0 and FOXP3) could individually reduce the significance of geographic locality-specific prognosis (figure 5A). This suggests that, apart from ‘age’, differences in tumour-infiltrating immune cells may influence locality-specific survival in the cohorts. Additionally, adjustment for chemotherapy status also found locality-specific prognosis to be affected only by tumour immunity factors (see online supplementary figure S7).

A multivariate Cox regression model incorporating all significant factors (univariate Cox regression $p$ value <0.05; see online supplementary figure S8A) and year of surgery, revealed that CD68 and CD3 were the only immune markers with...
independent associations with survival, however, their relationship with prognosis is complex and non-linear (see online supplementary figure S8B). To explain this relationship, we extrapolated previous findings from other cancer types where tumours with very high intratumour CD68/CD3 ratios are usually highly invasive, with patients bearing such tumours typically exhibiting poor prognosis.25 26 Indeed, in our TMA cohorts, a regression analysis of CD68/CD3 ratios with 5-year overall survival revealed that increased CD68/CD3 ratios were significantly and independently associated with worse survival outcomes ($\beta = -0.141$; p value=$9 \times 10^{-03}$; online supplementary table S11).

We then focused on the ‘high-risk’ group of patients with high CD68 and low CD3 levels (n=55) within the TMA cohorts and found that non-Asians tended to have significantly higher CD68/CD3 ratios compared with Asians (Welch’s t-test p for trend=0.04; figure S8B). This observation may explain at least, in part, why non-Asians with GC generally have worse survival rates compared with Asians with GC.

To provide additional support for the influence of CD68/CD3 ratios in GC clinical outcome, we revisited the gene expression cohorts. Due to residual confounding in the RUV-4 combined data, it was not possible to directly compare the pooled Asian to the non-Asian GCs at the single-gene level. We could, however, perform similar analyses within each dataset separately. In agreement with our results from the TMA cohorts, assessment of CD68/CD3 ratios in six of the microarray expression cohorts revealed that five cohorts also exhibited a significant trend for worse survival times with higher CD68/CD3 ratios (p for trend <0.05; online supplementary table S12 and figure S9).
Figure 4  Immunohistochemical (IHC) assessments of immune and inflammatory cell populations in Caucasian and Japanese gastric cancer (GC) cohorts. Panel (A) depicts IHC assessment of tumour infiltration by macrophages (CD68) and neutrophils (CD66b) in Caucasian (blue) versus Japanese (red) cohorts. There is significantly more CD68 staining (top diagram) in Caucasian compared with Japanese GCs, and significantly more CD66b staining (bottom diagram) in Japanese compared with Caucasian GCs. These results suggest the presence of inflammation in both cohorts, but different cellular recruitment. Panel (B) illustrates results after quantitative IHC for T-cell-related immune markers, that is, CD3 (general T-cell marker; top diagram), CD8 (cytotoxic T-cell marker; bottom left diagram), CD45R0 (marker for memory T-cells; bottom middle diagram) and FOXP3 (marker for regulatory T-cells; bottom right diagram), in Caucasian (blue) and Japanese (red) cohorts. With the exception of FOXP3, all markers are significantly enriched in Caucasian GCs. For FOXP3, there is significantly more staining in the Japanese GC. For all cases, *** indicates a significant (Wilcoxon Mann–Whitney test p<0.05) difference in the extent of staining. Panels also show corresponding IHC images from Caucasian and Japanese cohorts.
GC patient outcome after chemotherapy may be influenced by tumour immunity

Accumulating evidence suggests that antitumour activity of chemotherapy may involve the immune system. In an exploratory analysis, we tested whether the Asian versus non-Asian immune differences might impact on outcomes of patients treated with surgery with/without chemotherapy. In our retrospective series, patients receiving only surgery had significantly better survival than those given surgery with chemotherapy—this is likely due to chemotherapy-treated patients in our series tending to have a higher Tumour Node Metastasis (TNM) stage at the time of surgery. Given this caveat, a univariate Cox regression model was fitted. The model was then individually adjusted for various factors (e.g., cancer stage, age, immune cell marker etc.). We observed that, besides cancer staging, adjusting for CD3, CD68 and CD45R0 tended to negate the survival difference between patients with or without chemotherapy. Conversely, adjusting for CD8 and FOXP3 made the survival difference more pronounced (figure 5C). These results suggest that in GC, chemotherapy outcomes and immune effects may be interdependent.

Association of immune markers with other clinicopathologic factors

Finally, we assessed whether the immune markers (including CD68/CD3 ratios) might be associated with any other clinicopathologic factors (e.g., pathological Tumour Node Metastasis (pTNM) stage, EBV status, MMR status etc). We found that pTNM stage was positively correlated with CD68 and CD68/CD3 ratios, while EBV positivity was correlated with higher CD3, CD8 and CD45R0 levels but negatively correlated with CD68/CD3 ratios (see online supplementary table S13). Additionally, MMR deficiency was correlated with lower CD3, CD68, FOXP3 levels, while higher tumour differentiation grade was positively correlated with all markers except CD66b, FOXP3 and CD68/CD3 ratio (see online supplementary figure S13). Higher CD3, CD8 and CD68 levels were significantly correlated with diffuse-type GC (see online supplementary figure S13). Higher FOXP3 was the only marker significantly correlated with tumour location and correlated with tumours located in the middle third of the stomach.

Figure 5 Assessment of effect of immune factors on geographic locality-based and chemotherapy-based survival. Panel (A) illustrates a univariate Cox model fitted with only the geographic locality factor (orange box) and fitted, in turn, with each of the respective factors (e.g., chemotherapy, TNM staging, age etc). Adjusted HRs and their corresponding p values for locality are presented. All factors causing a change in significance of locality-specific prognosis are coloured in red text. Panel (B); scatterplot of 5-year overall survival (months) against CD68/CD3 ratios in the tissue microarray patient group (n=55) with high CD68 and low CD3 levels. There are more non-Asians (blue) than Asians (red) with higher CD68/CD3 ratios and low 5-year overall survival (light blue box). While those with lower CD68/CD3 ratios and high 5-year overall survival (light yellow box) tend to be Asians. Panel (C) illustrates a similar diagram to Panel (A), except a univariate Cox model is first fitted with the chemotherapy factor (orange box). Adjusted HRs and their corresponding p-values for chemotherapy are presented. All factors causing a change in significance of chemotherapy outcomes are coloured in red text.

1728


Stomach

Gut: first published as 10.1136/gutjnl-2014-308252 on 10 November 2014. Downloaded from http://gut.bmj.com/ on February 7, 2021 by guest. Protected by copyright.
DISCUSSION

Our analysis of >1600 GCs provides evidence that Asian and non-Asian GCs may exhibit distinct gene signatures related to inflammation and immunity, with T-cell pathways preferentially associated with non-Asian GC. Our study should be regarded as hypothesis-generating requiring further validation. Due to the retrospective nature of meta-analysis studies, there is no control over the initial study design of each cohort, and confounding variables present a major issue. Our study addressed this by using a novel statistical algorithm, RUV-4, to reduce technical noise; a two-stage analysis plan for reducing platform-specific artefacts; and a gene signature-based analysis (c.f. individual genes). Moreover, although residual confounding after RUV-4 adjustment clearly remains in the gene expression data, our immunohistochemistry validation studies in independent cohorts suggest that, at least on a disease-pathway level, we have captured true biological signals.

At the gene-expression level, non-Asian GCs exhibited enrichment in multiple signatures related to T-cell biology, including CD28 and CTLA-4 signalling (figure 3). Both CD28 and CTLA-4 (cytotoxic T-lymphocyte-associated protein 4) are glycoproteins expressed by T-cells. Quantitative IHC assessment of intratumoural T-cells in independent GC samples further confirmed T-cell enrichment in Caucasian compared to Japanese GCs, such as activated (memory) CD45R0-positive T-cell populations. We also observed lower levels of FOXP3, a T reg marker, in non-Asian GCs. This is consistent with our gene expression results as T reg cells are known to have immunosuppressive functions and are required for maintaining immune tolerance. Presently, it remains unclear why we see these immune-related differences between the cohorts. Rates of EBV infection and MMR status, both associated with GC immune cell infiltration, were similar between the cohorts, although a larger sample size may be required for a definitive conclusion.

A limitation of our study is that H pylori status information was unavailable for several cohorts, precluding a correlative analysis between the immune differences and H pylori exposure. H pylori status is not routinely tracked in clinical management, and assessment of active exposure is also challenging since H pylori exposure often occurs early in life (serology studies) and these organisms are usually absent in GC resection specimens. Notably, a recent study comparing US and Japanese GC patients reported that genetic variants in tumour immune checkpoint genes (eg, CTLA-4, FOXP3) may impact on clinical outcome in an ethnic-specific manner. It is thus possible that population differences in germline polymorphisms of immunity genes may contribute to the immune differences between GCs from different regions.

We explored whether these observed immune differences could contribute, at least in part, to differences in survival between GC patients from different regions. Our results suggest that in GC, differences in tumour immunity may indeed affect geographic locality-specific survival, similarly to what has been previously reported for other tumour types. For example, in bladder and breast cancer, high CD68/CD3 or CD3/CD20 ratios have been associated with worse prognosis. Assessment of CD68/CD3 ratios within our TMA and expression microarray cohorts yielded a similar conclusion. High tumour stage and EBV positivity, factors known to be associated with worse and better prognosis in GC patients, respectively, have also been found to be positively and negatively correlated with CD68/CD3, respectively. Notably, non-Asians in the high-risk CD68 high/CD3 low group tended to have significantly higher CD68/CD3 ratios compared with Asians, consistent with non-Asian GC patients having a generally worse prognosis. In the case of chemotherapy, previous studies have shown that 5FU can activate immune regulatory cells to stimulate pro-tumourigenic cytokines which can impact therapeutic efficacy. For patients in our cohorts given mainly 5FU-related-therapy (eg, S-1 or UFT), we found that tumour immunity can affect prognosis after chemotherapy.

The observation of immune-related differences in Asian and non-Asian GCs may potentially explain part of the geographical differences seen in recent clinical trials (eg, AVAGAST, RAINBOW (A Study of Paclitaxel With or Without Ramucirumab (IMC-1211B) in Metastatic Gastric Adenocarcinoma)), LOGiC). Two of these trials used drugs targeting angiogenesis (ie, AVAGAST-VEGF; RAINBOW-VEGFR2), and immune cells (eg, CD68 macrophages, CD4 T-cells) have been shown to have pro-angiogenic effects, while antigen-presenting cells can inhibit angiogenic activity through cytokines (eg, IFN (interferon), IL-12 and IL-27). It will be interesting to assess whether infiltrating T-cells represent an ‘immune biomarker’ for the efficacy of anti-angiogenic agents.

Our results may also have relevance for future trials investigating immunotherapy such as antibodies targeted at T-cell regulatory molecules (eg, CTLA-4, PD-1 and PD-L1). Our study identified enriched CTLA-4 signalling pathway in non-Asian GCs. CTLA-4 is the target of the anti-CTLA-4 antibody ipilimumab, and in GC, a Phase II clinical trial evaluating the role of CTLA-4 antibodies has recently been initiated (NCT01585987). For such therapies, geographical differences in the underlying tumour-immune biologies may affect treatment outcomes and hence the ultimate success of global trials.

In conclusion, the current study was performed on the largest collection of retrospectively sourced GC expression data, and revealed immune signature differences between Asian and non-Asian GCs. Due to the retrospective nature of this study, our results need to be further validated, perhaps through a prospective study designed to study locality effects. If confirmed, our study highlights the importance of identifying and validating novel targets and biomarkers in a geographically specific manner, to personalise interventions given to specific patient groups.

Author affiliations

1Department of Cancer Therapeutics and Stratified Oncology, Genome Institute of Singapore, Singapore, Singapore
2Department of Statistics, University of California at Berkeley, Berkeley, California, USA
3Department of Medical Oncology, National Cancer Centre, Singapore, Singapore
4Department of Pathology, Free University Medical Center Amsterdam, Amsterdam, The Netherlands
5Department of Internal Medicine, Yonsei Cancer Center, Seoul, South Korea
6Department of Pathology, Yonsei University College of Medicine, Seoul, South Korea
7Division of Cancer Medicine, Department of Systems Biology, MD Anderson Cancer Center, Houston, Texas, USA
8Department of Surgery, Yonsei University College of Medicine, Seoul, South Korea
9Department of Gastrointestinal Surgery, Kanagawa Cancer Center Hospital, Yokohama, Japan
10Department of Gastrointestinal Medical Oncology, MD Anderson Cancer Center, Houston, USA
11Department of Internal Medicine, Yonsei Cancer Center, Seoul, South Korea
12Gastroenterological Center, Yokohama City University Medical Center, Yokohama, Japan
13Gut Genomics and Biochemistry Laboratory, Peter MacCallum Cancer Centre, East Melbourne, Victoria, Australia

REFERENCES
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