Tumor-associated stroma shapes the spatial tumor immune microenvironment of primary Ewing sarcomas

Christopher Kuo^{1,2}, Krinio Giannikou¹, Nuoya Wang³, Mikako Warren⁴, Andrew Goodspeed^{5,6}, Nick Shillingford⁴, Masanori Hayashi⁷, Micha Sam Brickman Raredon^{3,8}, James F. Amatruda^{1,2}

¹Cancer and Blood Disease Institute, Division of Hematology-Oncology, Children's Hospital Los Angeles, ²Department of Pediatrics, Keck School of Medicine, University of Southern California, Los Angeles, CA; ³Department of Anesthesiology, Yale School of Medicine, New Haven, CT; ⁴Department of Pathology and Laboratory Medicine, Children's Hospital Los Angeles, Los Angeles, CA; ⁵University of Colorado Cancer Center, University of Colorado Anschutz Medical Campus, Aurora, CO; ⁶Department of Biomedical informatics, University of Colorado Anschutz Medical Campus, Aurora, CO; ⁷Department of Pediatrics, University of Colorado School of Medicine, Aurora, CO; ⁸Vascular Biology & Therapeutics, Yale School of Medicine, New Haven, CT

Corresponding Author:

James F Amatruda MD., PhD Children's Hospital Los Angeles 4650 Sunset Blvd MS #57, Los Angeles, CA 90027 E-mail:jamatruda@chla.usc.edu

Abstract

To date, few studies have detailed the tumor microenvironment (TME) of Ewing sarcoma (EwS). The TME has a vital role in cancer survival and progression with implications in drug resistance and immune escape. By performing spatially resolved transcriptomic analysis of primary treatment-naïve EwS samples, we discovered greater stromal enrichment in localized EwS tumors compared to metastatic EwS tumors. Through spatial ligand-receptor analysis, we show that the stromal enriched regions harbor unique extracellular matrix related cytokines, immune recruitment and proinflammatory microenvironmental signals, implying EwS stroma may play an anti-tumorigenic role by acting as an immune recruitment center. All EwS tumors expressed pro-tumorigenic MIF-CD74 immune signaling, suggesting a potential immune-evasive mechanism and immunotherapy target. Our findings provide insight into tumor cell/stromal cell interactions in EwS and serve as a valuable resource for further investigations in the tumor immune microenvironment of EwS.

Introduction

Ewing sarcoma (EwS) is a malignant bone and soft-tissue tumor that occurs most commonly in children, adolescents and young adults.^{1,2} Less than 30% of patients with initial metastases will survive for 5 years.^{3,4} New therapeutic approaches are needed given the vast majority of children with metastatic or relapsed EwS die despite intensive, multimodal therapy.⁵ Additionally, survivors face a lifelong risk of adverse health effects due to toxicity of therapy, including secondary cancers. EwS is caused by chromosomal translocations between a FET family member and an ETS-type gene, the most common of which is a t(11:22) that creates the EWSR1::FLI1 fusion oncogene.⁶ Despite the discovery of the EWSR1::FLI rearrangement more than 30 years ago, effective targeted therapies remain unavailable.⁷ Over the past decade, immunotherapy has revolutionized the field of cancer therapy. Instead of directly targeting cancer cells, immunotherapy harnesses the host's own immune system to eliminate cancer cells.⁸ Immunotherapy such as checkpoint inhibition through programmed death 1 (PD-1) blockade has recently achieved complete response in all patients with mismatch repairdeficient, locally advanced rectal cancer.^{9,10} Similar effectiveness with immunotherapy is seen with CAR-T therapies in leukemias.¹¹ However, despite advances in our understanding of immunobiology and cancer genomics, immunotherapy has not been successful in EwS.¹² While low tumor mutational burden and resulting scarcity of tumor infiltrating lymphocytes have been proposed as an explanation for the lack of immunogenicity of EwS, the potential immunosuppressive roles of the innate immune system and the tumor microenvironment (TME) in EwS have not been fully elucidated.⁶

Previous studies investigating the EwS TME have identified macrophages through conventional immunohistochemistry (IHC) using CD163 or CD68 as single-plex marker of macrophages.^{13,14} However, while informative, conventional IHC does not fully reflect the complexity of the TME, and although bulk tumor genomics has identified immunosuppressive M2 macrophages being the predominant immune cell type in EwS family of tumors.¹⁵ there is a paucity of literature on the spatial distribution of these immune cells within the TME. In recent years, single-cell RNA sequencing (scRNAseq) has allowed more granular dissection of the TME of many cancer types by profiling tumor and immune cells. Visser and colleagues performed scRNAseg of 18 primary EwS samples and reported dysfunctional antigen-presenting cells as well as immunosuppressive microenvironmental signals.¹⁶ Cillo et al. performed scRNAseg on immune cells from blood and tumors from patients with EwS, and identified distinct myeloid cells as well as CD8+ effector T cells and exhausted CD8+ cells.¹⁷ Although scRNA-seq analysis resolved the composition of cell types within the TME of EwS, detailed spatial transcriptomic analysis of the TME of EwS is lacking. The intricate interactions between the host's innate/adaptive immune system and the cancer cells lead to distinct spatial phenotypic and architectural changes within the tumor.¹⁸ Being able to visualize the spatial heterogeneity of cell types and cell states in intact tumor samples can provide insight into how the host's innate and adaptive

immune cells contribute to the TME. Additionally, this information could help identify novel treatment approaches to prevent tumor progression and/or invasion.¹⁸

Here we characterize the TME of EwS through spatial transcriptomics (ST) of primary biopsied human EwS tumors. We implemented a comprehensive pipeline for ST analysis of EwS, and noted significant enrichment of extracellular matrix-related (ECM-r) genes in primary tumors from patients with localized disease. These ECM-r genes colocalized with the tumor stroma. Gene composite analysis demonstrated that macrophage/monocyte and cancer-associated fibroblast (CAF) genes colocalized with the ECM-r genes. Fusion-driven sarcomas such as EwS have been noted to have low immune infiltrations compared to sarcomas with complex genomes.¹⁹ Despite this, our studies showed infiltrative T and B cells in EwS tumors. We found that within the tumor stroma there were significant ECM and immune related microenvironmental signals, indicating interactions between different types of immune cells and between immune cells and stromal cells. Tumors from patients with localized EwS (loc-EwS) showed enrichment of TGFB signaling in the stroma compared to tumors from patients with metastatic EwS (met-EwS). Tumor stroma harbored unique spatial microenvironmental connectivities related to tumor inhibition, immune activation and recruitment, which were predominantly observed loc-EwS tumors. Notably, we identified *MIF-CD74* immune signaling in all tumors that may contribute to a pro-tumoral and immune-evasive TME. We validated key TME features identified in our ST analysis via high-plex proteomics immunostaining at single-cell resolution. Collectively, our results provide novel insights into the spatial architecture and immune landscape of EwS tumors, suggesting possible antitumorigenic roles of stroma and potential mechanisms by which EwS manipulates the surrounding immune microenvironment to ultimately evade immune surveillance.

Results

A pipeline for spatial transcriptomic analysis of Ewing sarcoma

To define the spatial transcriptional landscape of Ewing sarcoma (EwS) tumors, we implemented a comprehensive pipeline for ST analysis of clinical tumor specimens (Figure 1). We selected formalin-fixed, paraffin-embedded biopsy tissue samples from 16 primary EwS tumors from patients who had either localized EwS or metastatic EwS at the time of diagnosis. All of the tumors were histopathologically confirmed by an expert pathologist to be Ewing sarcoma, and molecular diagnostics confirmed the presence of a *FET-ETS* fusion gene (Table 1). Quality control of the tumor specimens was performed and regions that were annotated as non-necrotic and tumor dense were then sectioned and placed on Visium spatial gene expression slides.²⁰ Following sequencing, data were aligned to the human reference genome (hg19) and spatially mapped. Expression data were then analyzed via dimensionality reduction (Uniform Manifold Approximation and Projection (UMAP)), clustering and visualization through STUtility and Seurat R packages.^{20,21} Expression data for each sample were

normalized using LogNormalize to optimize preservation of biological variation. After initial data processing, we assessed expression of gene markers specific to Ewing sarcoma (NKX2-2, CD99) to identify the spatial location of tumor cells.

We used a publicly available scRNA-seq EwS dataset as a reference to perform conditional autoregressive-based deconvolution analysis to enhance the ST resolution and infer the composition and abundance of all different cell types within each spot.^{22,23} We also profiled the immune cell abundance in all 16 tumors through CIBERSORTx.²⁴ To uncover the cellular microenvironment signals from ST data, we utilized Niches Interactions and Cellular Heterogeneity in Extracellular Signaling (NICHES).²⁵ Lastly, we validated the transcriptomic findings at the proteomics level by performing single cell spatial proteomics analysis using PhenoCycler.²⁶ In conclusion, our ST pipeline allowed us to understand the tumor microenvironment (TME) of EwS at the spatial level.

Integrated clustering reveals intertumoral heterogeneity of Ewing sarcoma

To further characterize the transcriptional heterogeneity between tumors, we merged and integrated ST data from all 16 tumors. When analyzed individually, each tumor had distinct clusters on UMAP plots, reflecting unique spatial transcriptional profiles (Figure 2A). Upon merging the datasets, tumors retained their distinct clustering patterns, occupying separate regions on the integrated UMAP (Figure 2B). For integration, we normalized and identified variable features for each tumor independently, and selected features that are repeatedly variable across all tumors. We applied a resolution of 0.4 to avoid overclustering of the data and to preserve biological heterogeneity. Clustering of all 16 integrated tumors produced 10 clusters, and integrated ST data from all 16 tumors overlapped on the UMAP (Figure 2C). When visualizing the integrated clusters represented (Figure 2D). These findings highlight the inherent intertumoral heterogeneity of Ewing sarcoma at transcriptional spatial level, underscoring the diverse molecular landscape across tumors.

Inference of tumor histology from gene expression

Our finding that transcriptional clustering revealed intertumoral heterogeneity of EwS led us to examine whether tumor histology could be inferred from ST gene expression data. To achieve this, we implement 'Estimation of Stromal and Immune cells in Malignant tumours using Expression data' (ESTIMATE) algorithm and using gene expression signatures we inferred the proportions of stromal and immune cells in tumor samples.²⁷ ESTIMATE performs single-sample gene set-enrichment analysis ²⁸ by leveraging curated 'stromal gene signatures' to capture stromal cells in tumor tissue and 'immune gene signature' to identify immune cells in tumor tissues.²⁷ A stromal and immune scores were then generated and combined to infer a tumor purity score. Prior to ST analysis, all EwS tumors were

subjected to hematoxylin and eosin (H&E) staining and digital images were reviewed by an expert anatomic pathologist. Tumor and stromal regions were annotated in blinded fashion. We then compared the annotated H&E-stained tumor tissues to gene-signature enriched annotated regions. All tumors demonstrated good agreement between histopathologic annotation and ST delineation of stromal and tumor regions (Figure 3 and Supplementary Figure S1A-D). Moreover, regions that were transcriptionally annotated as stroma overlapped with immune-rich with significant correlation (Figure S1E). Thus, these ST gene expression data recapitulated the histopathology of primary EwS tumors, reinforcing the utility of ST in capturing the complex spatial architecture of EwS.

Extracellular matrix-related genes are enriched in EwS tumors from patients with localized disease

To gain further insight in the transcriptional differences between primary tumors from patients with localized EwS (loc-EwS) and primary tumors from patients with metastatic EwS (met-EwS), we merged all 16 tumors together (Figure 2B) and applied Gene Set Enrichment Analysis (GSEA) to assess differential enrichment of molecular pathways using Hallmark gene sets between loc-EwS and met-EwS. Loc-EwS tumors were significantly enriched in the epithelial to mesenchymal transition (EMT) pathway (Figure 4A).²⁹ Several inflammatory molecular pathway signatures were also highly enriched in loc-EwS, such as tumor necrosis factor alpha (TNFa) signaling, interferon (IFN) alpha and IFN gamma response (Figure 4A). There was also enrichment of PI3K/AKT/mTOR signaling and transforming growth factor beta (TGF-b) signaling in loc-EwS (Figure 4A). In contrast, met-EwS exhibited enrichment in proliferative pathways such as E2F targets, MYC targets, G2M checkpoint (Figure 4A). Notably, met-EwS also exhibited downregulation of inflammatory response pathways compared to loc-EwS. Annexin A1 (ANXA1) was significantly overexpressed in loc-EwS compared to met-EwS. ANXA1 was recently reported to have prognostic significance with low expression level associated with lower overall survival in both primary and metastatic EwS.³⁰ We also noted upregulation of the six-transmembrane epithelial antigen of the prostate 1 (STEAP1) in loc-EwS tumors. High STEAP1 expression has been correlated with improved overall survival (OS) in patients with EwS.^{31,32}

To determine genes that are differentially expressed in loc-EwS and met-EwS, we performed FindAllMarkers (R Seurat) to assess for significantly differentially expressed genes within loc-EwS and met-EwS using Wilcoxon rank-sum test with an average log-fold change > 0.25 (Table 2). Loc-Ews tumors displayed higher expression of extracellular matrix (ECM) genes such as *DCN*, *COL6A3*, *COL1A2*, *COL1A*, *TNC*, *COL5A1*, *SPARC* (Figure 4B). In contrast, met-EwS, exhibited upregulation of genes associated with proliferation and metastasis, such as High-Mobility Group AT-Hook 1 (*HMGA1*), the cancer/testis antigen Preferentially Expressed Antigen in Melanoma (*PRAME*), Glutathione S-Transferase Mu 1, and T cell receptors related genes. Previously, gene expression analysis from EwS

patients that died identified several differentially expressed genes related to cell motility, cell migration and cell adhesion, and reported that *GSTM2*, a glutathione metabolism gene, was associated with worse outcome.³³ We found that met-EwS tumors exhibited enrichment of genes associated with glutathione metabolism (*GSTM1, GSTM4, SMS* and *ODC1*) and cell proliferation (*TWIST1, IGFBP2, IGF2BP1*) (Table 2), along with higher expression of High-Mobility Group AT-Hook 1 (*HMGA1*) and the cancer/testis antigen *PRAME. HMGA1* was recently reported to be overexpressed and associated with trabectedin resistance in soft tissue sarcomas.^{34,35} *PRAME* has been proposed as a potential multipediatric cancer target, and noted to have robust protein expression in patient derived xenografts of EwS.¹⁹ Glutathione S-transferases (GSTs) are isoenzymes that have been implicated in developing chemoresistance in cancer, and have been noted to be a promising candidate to overcome drug resistance against conventional chemotherapy in osteosarcoma, EwS and rhabdomyosacoma.^{36,37}

Based on the GSEA showing that loc-EwS tumors are enriched in genes in the EMT hallmark pathway, and the results from differential gene expression analysis indicating enrichment in ECM genes in loc-EwS, we next examined the overlap between the 267 differentially upregulated genes in loc-EwS and the 200 genes from the EMT hallmark pathway (Figure 4C). This analysis identified 46 overlapping genes, which we designated as 'ECM-related' (ECM-r) genes. To further explore their spatial distribution, we then performed AddModuleScore to calculate average expression levels of the ECM-r composite genes per spot across each ST tumor sample. Visualization of ECM-r gene enrichment revealed strong colocalization with stromal regions in both loc-EwS and met-EwS tumors (Figure 4D). However, loc-EwS had a higher ECM-r composite scores when compared to met-EwS using a Wilcoxon rank-sum test (Figure 4E). In summary, loc-EwS express ECM-r genes more highly compared to met-EwS and their spatial expression is concentrated within the tumor stroma, underscoring the distinct microenvironmental characteristics of localized EwS.

Gene composite analysis demonstrates unique spatial localization of immune cells and ECM in EwS

Our finding that loc-EwS have more ECM-r genes compared to met-EwS, the complex relationship between the ECM and the immune system, and the spatial overlap between stroma and immune signatures, prompted us to look further into the spatial mapping of EwS tumor cells, stroma and immune cells.³⁸ To map the spatial organization of these elements, we performed AddModuleScore analysis on EwS composite genes which consisted of *EWSR1::FLI1* target genes *NKX2-2, CD99, CAV1, CCND1, HES1, KDSR, PAPPA*,^{16,39-41} and found that the visualized spatial expression of this EwS composite score was consistent with the tumor distribution on H&E slides (Figure 5A,B, Supplementary Figure S2A-G). ECM-r genes had prominent expression in loc-EwS compared to met-EwS. As expected, ECM-r gene expression did not overlap with EwS composite genes.

Next, we assessed the presence of macrophage/monocytes using composite genes for macrophage/monocytes,⁴² M1 macrophages and M2 macrophages (Table 3).⁴³ In both loc-EwS and met-EwS tumors, the macrophage gene composite colocalized with the ECM gene composite, though ECM gene expression was much more prominent in loc-EWS. Expression of T cell gene composite markers⁴⁴ occurred predominantly within the EwS tumor spots compared to ECM-r spots (Figure 5 A,B). This spatial pattern was seen in both loc-Ews and met-EwS tumors. Due to the prominence of ECM-r regions, we assessed the presence of cancer-associated-fibroblasts (CAFs) specifically with myofibroblastic CAFs (myCAFs) and inflammatory CAFs (iCAFs) gene composites.⁴⁵ Both myCAFs and iCAFs colocalized in regions enriched in ECM-r genes and regions enriched in M1 and M2 macrophages in both loc-EwS and met-EwS (Figure 5A,B) (Figure S3). Taken together, we demonstrated that spatial gene annotations of EwS, ECM and immune cells exhibit distinct spatial localization patterns within the tumors, however, with limited resolution to discern differences between loc-EwS and met-EwS.

Spatial deconvolution reveals infiltrative T and B cells in Ewing sarcoma tumors

Recently Stahl and colleagues utilized CIBERSORT to deconvolute 197 microarray gene expression datasets of primary EwS tumor samples and reported that the most abundant cell types were predominantly immunosuppressive M2 macrophages, followed by T cells.^{46,47} Given these findings and the intriguing unique spatial localization of immune cells based on gene composite scores observed in our tumor cohort, we investigated immune infiltration in our cohort of EwS tumors, by performing CIBERSORTx.²⁴ We generated pseudobulk RNA counts of each of the 16 tumors and performed CIBERSORTx.²⁴ The predominant immune cell population were resting memory CD4 T cells (median: 27%), followed by M0 macrophage (median: 21%) and M2 macrophage (median:14%) (Figure 6A) (Table 4). However, CIBERSORTx profiled is restricted to 22 defined human immune cell types,⁴⁷ thus potentially missing other relevant immune cell types. Additionally, CIBERSORTx does not provide spatial location of these immune cell types.

To understand the spatial distribution of the cell types within each of the 16 tumors, we leveraged publicly available scRNAseq datasets of EwS tumors and performed deconvolution analysis using CARD (Figure 6B).^{23,22} Deconvolution with the reference EwS scRNAseq dataset provided insight in the proportions of various cell types within each tumor (Figure 6B). Significant intertumoral heterogeneity is noted among all tumors, with each tumor having different proportions of immune cells. We then visualized the deconvoluted proportions spatially. As expected, the transcriptomic inference of EwS cells' location recapitulated where EwS cells are localized within the H&E image in both loc-EwS and met-EwS tumors (Figure 6C, Supplementary Figure S3A-F). Similar to what we observed previously in gene composite analysis, we noted colocalization of CAFs at regions where ECM is dense, and T cells

scattered throughout the tumor-rich regions (Figure 6C, Supplementary Figure S3A-F). Spatial deconvolution further revealed notable intratumoral heterogeneity within each immune cell subset. In addition to T cells, we identified previously undescribed infiltrative B cell populations, and various T cell subsets, including T regulatory cells (Tregs), CD8+ effector T cells, CD4+IL7R+ T cells and memory T cells. We also noted that monocytes and macrophages were mostly congregated near or in ECM dense regions, and colocalized with CAFs (Figure 6). Contrasting to the CD14+ complement macrophages, other myeloid cells such as IL1B+ inflammatory monocytes, FCN1+ inflammatory monocytes and osteoclasts were dispersed throughout the tumor. While enrichment of Tregs colocalized with enrichment of EwS tumor cells, the signal was not as prominent as CD8+ effector T cells. In summary, deconvolution of ST data enhanced the resolution of the EwS TME revealing previously unreported immune cell infiltrates, and enabled spatial mapping of their distribution. This analysis also reinforced the intratumoral neterogeneity of EwS tumors, providing deeper insights into the immune landscape of EwS.

Loc-EwS tumors exhibit unique spatial ECM, and immune microenvironmental signals

Based on our previous findings that loc-EwS tumors are enriched in ECM related genes compared to met-EwS, and display unique spatial organization of immune cells through spatial deconvolution, we tested whether differences in TME contribute to biological differences between loc-EwS and met-EwS. To explore this, we first imputed the ST data using Adaptively-thresholded Low Rank Approximation (ALRA),⁴⁸ and then performed NICHES on individual tumors, exploring the ligand-receptor (LR) signaling connectivity in the neighborhood of each Visium spot using LR lists from Omnipath.^{25,49} We integrated the NICHES results and clustered the data with a resolution of 0.2, resulting in five NICHES microenvironmental clusters (Figure 7A). These microenvironmental clusters exhibited a distinct pattern compared to the original tumors' transcriptomic phenotypic clusters (Supplementary Figure S4). Microenvironmental cluster 1 primarily corresponded to EwS dense regions. To further characterize these clusters, we performed FindAllMarkers and compiled a microenvironmental signaling list using only significant LR markers (p-value <0.001) (Table 5).

Based on these observations, we specifically investigated the predominant microenvironment signaling connectivity unique to microenvironmental clusters 0 and 1, and examined the generated LR marker list (Table 5). NICHES analysis revealed unique microenvironmental connectivity signals in cluster 0 and cluster 1 (Supplementary Figure S5). In cluster 0, we observed significant ECM and immune related signals, such as collagen related LR microenvironmental signals (*COL4A1-CD93, COL4A1-ITB1, COL1A2-ITGB1*), TGF-beta related signals (*TGFB1-SDC2, TGFB1-TGFBR1-TGFBR2*), immune-related signals (*CD14-ITGB1, CD14-RIPK1, CXCL12-CD4, IL16-CD4 and MIF-CD74*) (Supplementary

Figure S5). In cluster 1, we observed predominant proliferative connectivity signals such as *WNT* related signals (*WNT5A-FZD1, WNT5A-LRP5, APCDD1-FZD1-LRP5*), Fibroblast Growth Factor Receptor (FGFR) signaling (*FGF13-FGFR4*), and immune signals (*CTF1-LIFR*) (Supplementary Figure S5).

Our finding that microenvironmental clusters hold unique microenvironmental connectivity signals led us to examine further what distinct signals exist within loc-EwS tumors compared to met-EwS tumors. We generated a loc-EwS LR marker list (Table 6) and a met-EwS marker list (Table 7). Loc-EwS tumors had prominent ECM related microenvironmental signals compared to met-EwS, such as collagen-related interactions and heparan sulfate proteoglycan-related interactions (Figure 7B). We also noted that *TGFB* signals were enriched in loc-EwS compared to met-EwS. We observed *WNT* signaling activation in both loc-EwS and met-EwS tumors, with an enrichment of microenvironmental connectivity associated with *WNT* inhibition signaling pathways in both loc-EwS and met-EwS tumors, such as *APCDD1-FZD4-LRP6* and *APCDD1-FZD2-LRP5*. Loc-EwS predominantly exhibited immune signaling pathways such as *C1QB-LRP1*, *CXCL12-CD4*, and *IL16-CD4*, in contrast to met-EwS (Figure 7B). In summary, loc-EwS tumors are enriched in ECM related microenvironmental signals in contrast to met-EwS tumors, and loc-EwS tumors are enriched in immune signaling compared to met-EwS tumors. These differences underscore the distinct microenvironmental landscapes of localized and metastatic EwS, highlighting potential targets for therapeutic intervention.

Ultrahigh-plex spatial phenotyping reveals B cells, lymphoid aggregates, and infiltrative M2 macrophages in EwS tumors

Given that the immune signal *MIF-CD74* was noted to be expressed in all tumors and may play a role in immunosuppression within the EwS TME, we investigated this signal further. To validate the unique spatial organization of immune cells, spatial enrichment of ECM and immune-related microenvironmental signaling of *MIF-CD74* reported in ST, we performed ultrahigh-plex PhenoCycler analysis on available EwS samples.²⁶ Sufficient tissue was available for spatial proteomics in 14 of the 16 EwS tumors on which we previously performed ST. We confirmed our deconvolution results and localized the infiltrative B cell populations in both loc-EwS and met-EwS tumors (Figure 8A,B). CD8+ T cells were detected in all tumors, with higher density in loc-EwS tumors compared to met-EwS tumors (Figure 8A and Supplementary Figures S6A,C,E,G,I,K). Consistent with prior reports, the predominant leukocyte population in the TME were tumor-associated macrophages (TAMs) (Figure 8A,B).⁴⁶ The majority of the TAMs surrounding CD8+ T cells expressed M2 markers (CD163+ and CD206+) rather than M1 (CD68+),^{50,51} and direct cell-cell contacts between M2 TAMs and CD8+ T cells were visualized in all EwS tumors (Supplementary Figures S6A,C,E,G,I,K). Furthermore, we discovered dense

aggregates of T- and B- cells with TAMs in unique spatial locations within both loc-EwS and met-EwS tumors (Supplementary Figure S7).

TAMs are infiltrative in solid tumors, and are highly dynamic and heterogeneous.⁵² TAMs have the ability to polarize to a pro-inflammatory (M1) state or a pro-tumoral (M2) state in response to environmental perturbations.⁵² We observed an abundance of infiltrative M2 TAMs (CD163+ and CD206+) in EwS tumors compared to M1 TAMs (CD68+). Consistent with our ST results, in addition to being infiltrative in tumors, TAMs co-localized with tumor-associated stroma (collagen IV) (Supplementary Figure S6B,D,F,H,J,L). Loc-EwS tumors exhibited greater stromal enrichment (Collagen IV) compared to met-EwS tumors (Figure 8A,B). Additionally, we confirmed the protein expression of TGF-B1 in both loc-EwS and met-EwS. Finally, we validated the immune-related signals PD1-PD-L1 and MIF-CD74 (Figure 9A,B). Based on the abundance of M2 TAMs, we investigated the phenotype of TAMs within the environment of tumor-associated MIF. MIF was highly expressed in all EwS tumors, with M2 TAMs expressing CD74 clustering in areas of high MIF microenvironments (Supplementary figure S6B,D,F,H,J,L). Interestingly, spatial regions with less tumor-associated MIF expression harbored more M1 TAMs (Supplementary figure S6B,D,F,H,J,L). Taken together, these results suggest that MIF-CD74 signaling between EwS and TAMs may drive or potentiate the immunosuppressive M2 predominant TME in EwS tumors.

Discussion

The ECM is an integral component of the TME that is highly diverse, acting as a reservoir for protumorigenic and anti-tumorigenic cytokines, serving as communication network between surrounding cells, and functioning as an anchor for tumor migration/invasion and as a mechanical barrier against tumor growth.⁵³⁻⁵⁶ Previous efforts to identify prognostic gene profiles in tumors from patients with localized EwS highlighted variation in stromal content, with stroma-rich tumors expressing integrin and chemokine gene sets linked to survival.³³ While tumor stroma was traditionally viewed as an immune barrier, it is now increasingly redefined as an immune rheostat, exerting context specific pro- and antitumor effects.⁵⁷

In this study, we discovered differentially expressed ECM-r genes in loc-EwS tumors that were spatially enriched primarily in the stromal regions including several collagen genes (*COL1A2*, *COL3A1*, *COL4A1*, *COL5A2*, *COL6A3*), matrisome genes (*SPARC*, *TNC*, *FBN1*), and immune genes (*HLA-A*, *HLA-E*). Individually these ECM-r genes were expressed at unique spatial locations of the tumors, suggesting site-specific tumor-immune-stroma interface. To examine potential effects of the ECM on immune cell infiltration in EwS tumors, we assessed the spatial distribution of tumor and immune cell genes, and enhanced the resolution through deconvolution, employing reference EwS scRNAseq

data.²³ Matrix and inflammatory CAFs colocalized with macrophages, specifically CD14+ complement macrophages at the stromal region. Multiple T cell subsets were enriched where EwS tumors were enriched as well, such as CD8+ T cells and tissue resident memory T cells, juxtaposing the stromal region. Additionally, there was a notable presence of B cells inferred in spots coinciding with EwS tumor cells, a previously undescribed finding in the TME of EwS. Across all 16 tumors, B cells and CD8+ T cells did not colocalize with CAFs. Interestingly, majority of the loc-EwS tumors (60%) had spot correlation of cytotoxic CD8+ T cells with EwS cells, in contrast to met-EwS tumors (33%) (Supplementary figure S8A-C), further reinforcing the unique tumor immune microenvironment (TIME) in the ECM enriched loc-EwS tumors.

To study the TME signaling connectivities underlying the stromal and tumor regions between loc-EwS and met-EwS tumors, we performed NICHES analysis. We discovered that loc-EwS tumors harbored unique spatial ECM, immune and proliferative microenvironmental signals. A key finding was the enrichment of TGF-β, a well-established potent regulator of the TME.^{58,59} *TGFB1-TGFBR1-TGFBR2* and TGFB3-TGFBR1-TGFBR2 interactions were noted to be expressed in majority of the loc-EwS predominantly enriched in the stromal regions (Supplementary figure S9). Furthermore, we noted TGFB1-SDC2 microenvironmental signal in majority of the loc-EwS tumors. Functionally, SDC2 acts as a co-receptor for TGFBR to mediate TGFB signaling, increasing ECM deposition and fibrosis, aligned with our ECM-enriched loc-EwS tumors.^{60,61} Several collagens-integrins interactions were also noted within ECM-enriched cluster, including COL4A1-ITGB1, COL1A2-ITGB1, COL6A1-ITGB1, suggestive of integrin related immune cell adhesion and immune response.^{62,63} Di Martino et al recently reported that tumor derived collagen III interacts with DDR1 (discoidin domain receptor tyrosine kinase) to maintain a wavy collagen architecture, thereby restricting cancer cell proliferation in a breast cancer model.⁶⁴ Interestingly, we discovered significant enrichment of COL3A1-DDR1 connectivity predominantly in loc-EwS tumors, suggestive of stroma restricting tumor growth. In loc-EwS tumors, we noted significant enrichment of immune cell recruitment signals (CXCL12-CD4 and IL16-CD4) within the ECM enriched cluster. CXCL12 is known to play a critical role in monocyte extravasation, macrophage differentiation and can stimulate antigen-specific T cell responses.⁶⁵ IL16 is a chemoattractant for CD4-immune cells such as T cells and monocytes/macrophages, and induces production of inflammatory cytokines on monocytes and antigen presenting cells.⁶⁶⁻⁶⁹ We also noted signals of macrophage migration (HSPG2-LRP1), and proinflammatory activation (CD14-RIPK1) within the ECM enriched regions.^{70,71} These findings highlight the anti-tumorigenic role of stroma and its involvement in immune cell recruitment and activation within the stroma of EwS tumors.

Despite evidence of immune cell recruitment and proinflammatory activation in our tumors, the TME of EwS is typically classified as immune-cold, with generally poor responses to immunotherapy.⁷² This raises the question of whether other mechanisms within EwS are counteracting the immune activation

signals within the stromal compartment, thereby rendering immune surveillance dysfunctional. To investigate this further, we observed MIF-CD74 immune signaling expressed in all tumors. MIF is a proinflammatory cytokine that is overexpressed in several cancers such as lung cancer, osteosarcoma, neuroblastoma and melanoma.⁷³ MIF is known to bind to CD74 with a role in immune homeostasis, cancer proliferation and inhibition of apoptosis.⁷⁴ Importantly, MIF is implicated in immune evasion, driving the polarization of M2 macrophages and inhibiting T cell activation.⁷⁵⁻⁷⁷ Additionally we also found an IL1RAP-PTPRF signaling pathway that was expressed in all tumors. IL1RAP is known to be induced by EwS and suppresses anoikis to metastasize.⁷⁸ These spatially unique immune related TIME signals highlight the complex and novel mechanisms that EwS may utilize to manipulate its TME, enabling tumor proliferation and immune evasion. Additionally, we noted PDL1-PD1 was only expressed in <25% of tumors, with signals predominantly in loc-EwS tumors. This may explain why, to date, PD-1 inhibitor treatment has not yielded clinical responses in Ewing sarcoma.⁷²

There is a dearth of literature investigating in the expression of CD8+ T cells in EwS tumors. A study by Machado and colleagues reported the lack of infiltrative CD8+ T cells in EwS tumors.⁷⁹ Consistent with this, our ultrahigh-plex spatial proteomic analysis revealed a scarcity of CD8+ T cells within the TIME of met-EwS compared to loc-EwS. Interestingly, EwS tumors had a striking abundance of M2 TAMs infiltrating both the stromal and tumor compartments in direct contact with the surrounding CD8+ T cells and EwS cells. Despite M2 TAMs being the predominant TAM subtype in EwS tumors, the detailed functional role of TAMs within the EwS TME remains poorly understood. In pancreatic ductal adenocarcinoma, TAMs exhibited characteristics of both M1 and M2 macrophages, holding both anti-and proinflammatory properties.⁸⁰ Thus, the M2 TAMs within the TME of EwS may still retain antigen presenting capabilities. In recent years, a deeper understanding of lymphoid aggregates and tertiary lymphoid structures (TLSs) have improved the response to immunotherapy in lung cancer.^{81,82} Currently, there are no reports of lymphoid aggregates in EwS tumors. The presence of dense lymphoid aggregates in our EwS tumors raises the question of the possible function and the prognostic association of lymphoid aggregates in the TIME of EwS.

This is the largest study to date of ST analysis on treatment naïve EwS tumors obtained at diagnosis composed of both loc-EwS and met-EwS tumors. Previous efforts in profiling the immune composition of EwS have been limited based on microarray gene expression datasets, multiplex IHC, and scRNA-seq.^{17,46} While this study provides valuable insights into the EwS TME, we acknowledge some limitations. First, spatial transcriptomics performed via Visium V1 (10X) lacks a single cell resolution which may obscure finer cellular interactions. However, we attempted to mitigate this challenge via use of high-resolution multiplex immunostaining. Additionally, we did not have matched scRNAseq data from the same patients thus limiting our ability to perform direct transcriptional deconvolution. We primarily relied on clinical biopsy samples, which may not fully capture the heterogeneity of entire

tumors. The small sample size of our cohort limited the statistical power to identify significant correlations between ST data and clinicopathological variables such as disease status/clinical outcome, age, or gender. We also acknowledge the need for external validation in a larger independent patient cohort to address the sample size limitation.

In conclusion, consistent with other studies, we found that EwS tumors exhibit significant intra- and inter-tumoral spatial heterogeneity.^{16,17,23,83} Our data revealed variable stromal representation, with greater stromal enrichment in loc-EwS compared to met-EwS. Additionally, we observed a scarcity of cytotoxic T cells in tumor compartment of met-EwS compared to loc-EwS. This suggests that the stromal component of EwS tumors may play an anti-tumorigenic role by acting as an immune recruitment center. The stromal compartment of EwS appears to harbor both innate and adaptive immune cells, promoting antigen presentation to re-direct T cells into the tumor. The reduced stromal component in met-EwS tumors results in the loss of this antigen presentation niche, specifically mediated by macrophages, thereby contributing to immune escape. Furthermore, our data also emphasize the multidimensional complexity of the TIME in EwS. EwS tumors employ various immune signals to evade immune surveillance and convert the surrounding TME into an pro-tumorigenic state, such as through MIF-CD74, and VISTA. These signals occur simultaneously in various spatial regions of EwS tumors, which may explain the limited success of mono-immunotherapy approach thus far. Additionally, the presence of CD8+ T cells within an M2 TAMs predominant TIME raises the question of whether EwS has a hot TME instead, with the CD8+ T cells' presence being neglected due to the complex immune evading mechanisms in the TME.⁸⁴ Most importantly, we report previously undescribed novel immune targets that may revert the pro-tumor TME to an anti-tumor TME. Future studies incorporating whole-tumor mapping and higher-resolution transcriptomic methods will be essential to expand on these findings. Further investigation of the immunomodulatory role of stroma in EwS as well as the mechanistic roles of MIF-CD74 signaling within EwS TME in vitro and in vivo may uncover novel therapeutic opportunities to counteract immune evasion and improve EwS patient outcomes.

Methods

Sample selection

A total of 16 formalin-fixed paraffin-embedded treatment naïve primary Ewing sarcoma tissue samples were obtained from patients with localized or metastatic Ewing sarcoma at diagnosis. Samples underwent quality check process, and all samples had DV200 scores >/= 30-50% prior to being selected for spatial transcriptomics (Table S1). The clinicopathologic characteristics of samples used in the study were summarized in Table 1. All samples and clinical data were obtained from the Children's Hospital Los Angeles (CHLA) Tumor Registry Study under protocols approved by the CHLA's

Institutional Review Board (CCI-06-00198) Written informed parental permission, and where possible, child assent, was obtained for all patients participating in the study.

Visium spatial transcriptomics

Spatial transcriptomics was performed according to the manufacturer's protocol (10x Genomics). In brief, a 5-µm-thick EwS tissue FFPE sections were placed onto a spatial transcriptomics expression slide (10x Genomics, #1000188). Each gene expression slide contains four 6.5 mm x 6.5 mm capture areas covering 4992 spatially barcoded spots (resolution 1-10 cells/spot). Tissue sections on Visium slides were first fixed in methanol (Millipore Sigma #34860) followed by an aqueous eosin-based hematoxylin and eosin (H&E) staining, and imaging protocol CG000160). Brightfield images were acquired with a DMI6000B microscope equipped with a 20x/0.7 HC PL APO lens and a DFC295 color camera (Leica Microsystems, Buffalo Grove, IL). Following imaging of H&E staining, permeabilization, cDNA synthesis and library generation was performed according to manufacturer's protocol for the Visium Spatial Gene Expression Slide and Reagent Kit (10x Genomics). Quality control of dual-indexed barcoded libraries was performed with Agilent Bioanalyzer. Paired-end dual indexed final libraries were sequenced with 1% Phix on the Novaseq platform with these parameters: Read 1-28 cycles, Index 1-10 cycles, Index 2-10 cycles, Read 2-90 cycles.

ST data preprocessing and normalization

Sequencing ST data were preprocessed using the Space Ranger pipelines (version 1.3.1; 10x Genomics). Base call files (BCL) data were demultiplexed and converted to FASTQ files. The spaceranger count pipeline was then used for read alignment to human reference genome GRCh38, unique molecular identifier (UMI) counting, and generation of feature-spot matrices corresponding to the H&E image within the fiducial frame of the gene expression slide. The raw UMI counts and spatial barcodes were imported to R program STUtility (version 1.1.1).²⁰ Gene expression count data were then normalized using the NormalizeData function from the R package Seurat (version 5.1.0).²¹

Histopathologic annotation

FFPE tissues from each block selected for study were stained with H&E and non-necrotic tumor dense regions were manually selected by an expert pathologist for spatial transcriptomics. Images captured from Visium slides were reviewed and annotated by an expert pathologist.

ESTIMATE analysis

ESTIMATE analysis was performed with the R package tidyestimate (version 1.1.1).²⁷

Gene set enrichment analysis

Gene set enrichment analysis (GSEA) was applied at the merged processed objects of all 16 tumors using the R packages msigdbr (version 7.5.1) and fgsea (1.30.0). We chose to perform GSEA on the

merged object instead of integrated object because too much biological variation was removed upon integration process.

Gene composite score annotation

Gene signatures were compiled to annotate EwS, Macrophages, M1 Macrophages, M2 Macrophages, T Cells, myCAFs and iCAFs (Table 2). The gene signatures were then used to generate a gene composite score per spot using the function AddModuleScore in Seurat (version 5.1.0).²¹

CIBERSORTx analysis

Immune cell abundance estimation was performed through CIBERTSORTx.²⁴ Normalized gene expression data was compiled to create a pseudobulk profile. This profile was then compiled as a mixture file and analyzed against the LM22 immune.⁴⁷

CARD Deconvolution analysis

Processed and normalized gene expression count data objects from STUtility were deconvolute through CARD using reference EwS scRNAseq dataset.^{20,22,23}

Ligand-receptor interaction analysis

Processed and normalized gene expression count data objects from STUtiity from all 16 tumors were merged, integrated and imputed with ALRA.⁴⁸ Spatial ligand-receptor neighborhood to spot connectivity was assessed with NICHES on R using LR from Omnipath.^{25,49} Integrated NICHES results were then clustered with a resolution of 0.2 and predominant microenvironmental clusters 0 and 1 were selected for further downstream analysis and ligand-receptor (LR) marker list generation. LR marker list was generated based on microenvironmental clusters 0 and 1 and thresholds of average pct1 >0.8, average metastatic score >5, and average localized score >8 were selected in order to capture highly significant LRs that is present in majority of the tumors. To find significant mechanisms differentiating loc-/met Ews samples, we ran a statistical test for each possible 2-sample pairing, one sample from localized samples and the other from metastatic samples, with total of 10x6 = 60 pairs. In each statistical test we used FindAllMarkers() function in Seurat package, setting min.pct = 0.05, logfc.threshold=0.1 and only.post=T. Then we concatenated the 60 marker lists and counted the frequency of LR mechanisms, defining this frequency as score for that mechanism. Similarly, we define number of times the mechanism appeared as a Localized marker as "avg Localized score", while number of times the mechanism appeared as a Metastatic marker as "avg Metastatic score". During the aggregation of this concatenated list, we also generated new p-values and logFC values. We use Fisher's method to combine old p-values. New logFC values are calculated by log₂(mean(2^abg log₂FC)). An additional ratio score was generated by dividing avg pct1/avg pct2 score and a power score was generated by avg pct1/avg pct2*avg log₂FC to further assess most expressed LR pairs. Additionally, a loc-EwS LR

marker list and a met-EwS marker list were generated with thresholds of average pct1 >0.8, average score >35 were selected for loc-EwS marker list, and thresholds of average pct1 >0.8, and average score >30 were selected for met-EwS marker list.

Phenocycler-Fusion FFPE staining

FFPE sections of EwS tumors (5-µm-thick) were stained and barcode-conjugated antibodies (Table S2) according to the protocol offered by AKOYA biosciences (PhenoImager Fusion user guide PD-000011 REV M, Date: 12.05.2023) with minor modifications. In brief, sections were baked at 60°C overnight, dewaxed by incubating for 10 min in xylene (Sigma-Aldrich, 185566-1L) twice, hydrated in solutions of decreasing EtOH concentration, and washed for 5 min in 18.2MΩ water twice. They were then incubated in a NxGen Decloaking chamber (Biocare Medical DC2012) with 1x antigen retrieval AR9 solution (AKOYA, AR9001KT) for 20 min at 110°C. Slides were then allowed to sit at RT for 30 minutes and were then washed with 18.2M Ω water, followed by two washes in hydration buffer (AKOYA staining kit 7000017). Samples were then incubated in staining buffer (AKOYA staining kit 7000017) for 30 min, followed by staining with the antibody cocktail for 3 hours at RT. Stained slides were then washed in staining buffer, fixed with 1.6% Paraformaldehyde (EMS 15710) prepared in storage buffer (AKOYA staining kit 7000017) for 10 min at RT, washed 3x in PBS, treated with ice-cold methanol (Sigma-Aldrich 34860-1L-R) for 5 min at 4°C, washed 3x in PBS, fixed with fixative reagent (AKOYA staining kit 7000017) for 20 min at RT, washed 3x in PBS and stored in storage buffer at 4°C. Within 5 days of staining, antibody-stained slides were imaged using the reciprocal fluorescently labelled reporters (Table S2) on a Phenocycler-Fusion 2.0 system.²⁶ Output images were segmented using the Steinbock pipeline.85

References

1. Morales E, Olson M, Iglesias F, Dahiya S, Luetkens T, Atanackovic D. Role of immunotherapy in Ewing sarcoma. *J Immunother Cancer*. 12 2020;8(2)doi:10.1136/jitc-2020-000653

2. Hawkins AG, Julian CM, Konzen S, Treichel S, Lawlor ER, Bailey KM. Microenvironmental Factors Drive Tenascin C and Src Cooperation to Promote Invadopodia Formation in Ewing Sarcoma. *Neoplasia*. 10 2019;21(10):1063-1072. doi:10.1016/j.neo.2019.08.007

3. Riggi N, Suva ML, Stamenkovic I. Ewing's Sarcoma. *N Engl J Med*. Jan 14 2021;384(2):154-164. doi:10.1056/NEJMra2028910

4. Ewing J. The Classic: Diffuse endothelioma of bone. Proceedings of the New York Pathological Society. 1921;12:17. *Clin Orthop Relat Res.* Sep 2006;450:25-7.

doi:10.1097/01.blo.0000229311.36007.c7

5. Crompton BD, Stewart C, Taylor-Weiner A, et al. The genomic landscape of pediatric Ewing sarcoma. *Cancer Discov*. Nov 2014;4(11):1326-41. doi:10.1158/2159-8290.CD-13-1037

6. Spurny C, Kailayangiri S, Jamitzky S, et al. Programmed cell death ligand 1 (PD-L1) expression is not a predominant feature in Ewing sarcomas. *Pediatr Blood Cancer*. Jan 2018;65(1)doi:10.1002/pbc.26719

7. Delattre O, Zucman J, Plougastel B, et al. Gene fusion with an ETS DNA-binding domain caused by chromosome translocation in human tumours. *Nature*. Sep 10 1992;359(6391):162-5. doi:10.1038/359162a0

8. Billan S, Kaidar-Person O, Gil Z. Treatment after progression in the era of immunotherapy. *Lancet Oncol.* Oct 2020;21(10):e463-e476. doi:10.1016/S1470-2045(20)30328-4

9. Cercek A, Lumish M, Sinopoli J, et al. PD-1 Blockade in Mismatch Repair-Deficient, Locally Advanced Rectal Cancer. *N Engl J Med.* Jun 23 2022;386(25):2363-2376. doi:10.1056/NEJMoa2201445

10. Kholodenko IV, Kalinovsky DV, Doronin, II, Deyev SM, Kholodenko RV. Neuroblastoma Origin and Therapeutic Targets for Immunotherapy. *J Immunol Res.* 2018;2018:7394268. doi:10.1155/2018/7394268

11. Wan Z, Sun R, Moharil P, et al. Research advances in nanomedicine, immunotherapy, and combination therapy for leukemia. *J Leukoc Biol*. Feb 2021;109(2):425-436. doi:10.1002/JLB.5MR0620-063RR

12. Guven DC, Kilickap S, Yildirim HC, Ceylan F, Bas O, Dizdar O. Chemoimmunotherapy for the salvage treatment of Ewing sarcoma: A case report. *J Oncol Pharm Pract*. Jul 2021;27(5):1281-1283. doi:10.1177/1078155220965677

13. Handl M, Hermanova M, Hotarkova S, et al. Clinicopathological correlation of tumor-associated macrophages in Ewing sarcoma. *Biomed Pap Med Fac Univ Palacky Olomouc Czech Repub*. Mar 2018;162(1):54-60. doi:10.5507/bp.2017.049

14. Fujiwara T, Fukushi J, Yamamoto S, et al. Macrophage infiltration predicts a poor prognosis for human ewing sarcoma. *Am J Pathol*. Sep 2011;179(3):1157-70. doi:10.1016/j.ajpath.2011.05.034

15. Stahl D, Gentles AJ, Thiele R, Gütgemann I. Prognostic profiling of the immune cell microenvironment in Ewing's Sarcoma Family of Tumors. *Oncoimmunology*. 2019 2019;8(12):e1674113. doi:10.1080/2162402X.2019.1674113

16. Visser LL, Bleijs M, Margaritis T, van de Wetering M, Holstege FCP, Clevers H. Ewing Sarcoma Single-cell Transcriptome Analysis Reveals Functionally Impaired Antigen-presenting Cells. *Cancer Res Commun.* Oct 24 2023;3(10):2158-2169. doi:10.1158/2767-9764.CRC-23-0027

17. Cillo AR, Mukherjee E, Bailey NG, et al. Ewing Sarcoma and Osteosarcoma Have Distinct Immune Signatures and Intercellular Communication Networks. *Clin Cancer Res.* Nov 14 2022;28(22):4968-4982. doi:10.1158/1078-0432.CCR-22-1471

18. Maniatis S, Petrescu J, Phatnani H. Spatially resolved transcriptomics and its applications in cancer. *Curr Opin Genet Dev.* 02 2021;66:70-77. doi:10.1016/j.gde.2020.12.002

19. Brohl AS, Sindiri S, Wei JS, et al. Immuno-transcriptomic profiling of extracranial pediatric solid malignancies. *Cell Rep.* Nov 23 2021;37(8):110047. doi:10.1016/j.celrep.2021.110047

20. Bergenstråhle J, Larsson L, Lundeberg J. Seamless integration of image and molecular analysis for spatial transcriptomics workflows. *BMC Genomics*. Jul 14 2020;21(1):482. doi:10.1186/s12864-020-06832-3

21. Hao Y, Hao S, Andersen-Nissen E, et al. Integrated analysis of multimodal single-cell data. *Cell*. Jun 24 2021;184(13):3573-3587.e29. doi:10.1016/j.cell.2021.04.048

22. Ma Y, Zhou X. Spatially informed cell-type deconvolution for spatial transcriptomics. *Nat Biotechnol.* Sep 2022;40(9):1349-1359. doi:10.1038/s41587-022-01273-7

23. Goodspeed A, Bodlak A, Duffy AB, et al. Characterization of transcriptional heterogeneity and novel therapeutic targets using single cell RNA-sequencing of primary and circulating Ewing sarcoma cells. *bioRxiv*. Jun 20 2024;doi:10.1101/2024.01.18.576251

24. Steen CB, Liu CL, Alizadeh AA, Newman AM. Profiling Cell Type Abundance and Expression in Bulk Tissues with CIBERSORTx. *Methods Mol Biol*. 2020;2117:135-157. doi:10.1007/978-1-0716-0301-7_7

25. Raredon MSB, Yang J, Kothapalli N, et al. Comprehensive visualization of cell-cell interactions in single-cell and spatial transcriptomics with NICHES. *Bioinformatics*. Jan 01 2023;39(1)doi:10.1093/bioinformatics/btac775

26. Black S, Phillips D, Hickey JW, et al. CODEX multiplexed tissue imaging with DNA-conjugated antibodies. *Nat Protoc*. Aug 2021;16(8):3802-3835. doi:10.1038/s41596-021-00556-8

 Yoshihara K, Shahmoradgoli M, Martínez E, et al. Inferring tumour purity and stromal and immune cell admixture from expression data. *Nat Commun*. 2013;4:2612. doi:10.1038/ncomms3612
Ren EH, Deng YJ, Yuan WH, Wu ZL, Zhang GZ, Xie QQ. An immune-related gene signature for determining Ewing sarcoma prognosis based on machine learning. *J Cancer Res Clin Oncol*. Jan 2021;147(1):153-165. doi:10.1007/s00432-020-03396-3

29. Ehnman M, Chaabane W, Haglund F, Tsagkozis P. The Tumor Microenvironment of Pediatric Sarcoma: Mesenchymal Mechanisms Regulating Cell Migration and Metastasis. *Curr Oncol Rep.* Aug 15 2019;21(10):90. doi:10.1007/s11912-019-0839-6

30. Shi X, Wu Y, Tang L, Ni H, Xu Y. Downregulated annexin A1 expression correlates with poor prognosis, metastasis, and immunosuppressive microenvironment in Ewing's sarcoma. *Aging (Albany NY)*. Mar 28 2023;15(6):2321-2346. doi:10.18632/aging.204615

31. Markey FB, Romero B, Parashar V, Batish M. Identification of a New Transcriptional Co-Regulator of STEAP1 in Ewing's Sarcoma. *Cells*. May 24 2021;10(6)doi:10.3390/cells10061300 32. Grunewald TGP, Ranft A, Esposito I, et al. High STEAP1 expression is associated with improved outcome of Ewing's sarcoma patients. *Ann Oncol*. Aug 2012;23(8):2185-2190. doi:10.1093/annonc/mdr605

33. Volchenboum SL, Andrade J, Huang L, et al. Gene Expression Profiling of Ewing Sarcoma Tumors Reveals the Prognostic Importance of Tumor-Stromal Interactions: A Report from the Children's Oncology Group. *J Pathol Clin Res.* Apr 2015;1(2):83-94. doi:10.1002/cjp2.9

34. Hui P, Li N, Johnson C, et al. HMGA proteins in malignant peripheral nerve sheath tumor and synovial sarcoma: preferential expression of HMGA2 in malignant peripheral nerve sheath tumor. *Mod Pathol.* Nov 2005;18(11):1519-26. doi:10.1038/modpathol.3800464

35. Moura DS, Mondaza-Hernandez JL, Sanchez-Bustos P, et al. HMGA1 regulates trabectedin sensitivity in advanced soft-tissue sarcoma (STS): A Spanish Group for Research on Sarcomas (GEIS) study. *Cell Mol Life Sci.* May 17 2024;81(1):219. doi:10.1007/s00018-024-05250-y

36. Townsend DM, Tew KD. The role of glutathione-S-transferase in anti-cancer drug resistance. *Oncogene*. Oct 20 2003;22(47):7369-75. doi:10.1038/sj.onc.1206940

37. Pasello M, Manara MC, Michelacci F, et al. Targeting glutathione-S transferase enzymes in musculoskeletal sarcomas: a promising therapeutic strategy. *Anal Cell Pathol (Amst)*. 2011;34(3):131-45. doi:10.3233/ACP-2011-012

38. Sutherland TE, Dyer DP, Allen JE. The extracellular matrix and the immune system: A mutually dependent relationship. *Science*. Feb 17 2023;379(6633):eabp8964. doi:10.1126/science.abp8964 39. Town J. Pais H. Harrison S. et al. Exploring the surfaceome of Ewing sarcoma identifies a new

39. Town J, Pais H, Harrison S, et al. Exploring the surfaceome of Ewing sarcoma identifies a new and unique therapeutic target. *Proc Natl Acad Sci U S A*. Mar 29 2016;113(13):3603-8. doi:10.1073/pnas.1521251113

40. Aynaud MM, Mirabeau O, Gruel N, et al. Transcriptional Programs Define Intratumoral Heterogeneity of Ewing Sarcoma at Single-Cell Resolution. *Cell Rep.* Feb 11 2020;30(6):1767-1779.e6. doi:10.1016/j.celrep.2020.01.049

41. Heitzeneder S, Sotillo E, Shern JF, et al. Pregnancy-Associated Plasma Protein-A (PAPP-A) in Ewing Sarcoma: Role in Tumor Growth and Immune Evasion. *J Natl Cancer Inst*. Sep 01 2019;111(9):970-982. doi:10.1093/jnci/djy209

42. Schelker M, Feau S, Du J, et al. Estimation of immune cell content in tumour tissue using single-cell RNA-seq data. *Nat Commun*. Dec 11 2017;8(1):2032. doi:10.1038/s41467-017-02289-3 43. Cheng S, Li Z, Gao R, et al. A pan-cancer single-cell transcriptional atlas of tumor infiltrating myeloid cells. *Cell*. Feb 04 2021;184(3):792-809.e23. doi:10.1016/j.cell.2021.01.010

44. Yu X, Chen YA, Conejo-Garcia JR, Chung CH, Wang X. Estimation of immune cell content in tumor using single-cell RNA-seq reference data. *BMC Cancer*. Jul 19 2019;19(1):715. doi:10.1186/s12885-019-5927-3

45. Elyada E, Bolisetty M, Laise P, et al. Cross-Species Single-Cell Analysis of Pancreatic Ductal Adenocarcinoma Reveals Antigen-Presenting Cancer-Associated Fibroblasts. *Cancer Discov*. Aug 2019;9(8):1102-1123. doi:10.1158/2159-8290.CD-19-0094

46. Stahl D, Gentles AJ, Thiele R, Gutgemann I. Prognostic profiling of the immune cell microenvironment in Ewing s Sarcoma Family of Tumors. *Oncoimmunology*. 2019;8(12):e1674113. doi:10.1080/2162402X.2019.1674113

47. Newman AM, Liu CL, Green MR, et al. Robust enumeration of cell subsets from tissue expression profiles. *Nat Methods*. May 2015;12(5):453-7. doi:10.1038/nmeth.3337

48. Linderman GC, Zhao J, Roulis M, et al. Zero-preserving imputation of single-cell RNA-seq data. *Nat Commun.* Jan 11 2022;13(1):192. doi:10.1038/s41467-021-27729-z

49. Türei D, Valdeolivas A, Gul L, et al. Integrated intra- and intercellular signaling knowledge for multicellular omics analysis. *Mol Syst Biol*. Mar 2021;17(3):e9923. doi:10.15252/msb.20209923 50. Xu ZJ, Gu Y, Wang CZ, et al. The M2 macrophage marker. *Oncoimmunology*. 2020;9(1):1683347. doi:10.1080/2162402X.2019.1683347

51. Fang J, Ou Q, Wu B, et al. TcpC Inhibits M1 but Promotes M2 Macrophage Polarization via Regulation of the MAPK/NF-κB and Akt/STAT6 Pathways in Urinary Tract Infection. *Cells*. Aug 28 2022;11(17)doi:10.3390/cells11172674

52. Vitale I, Manic G, Coussens LM, Kroemer G, Galluzzi L. Macrophages and Metabolism in the Tumor Microenvironment. *Cell Metab.* Jul 02 2019;30(1):36-50. doi:10.1016/j.cmet.2019.06.001 53. de Visser KE, Joyce JA. The evolving tumor microenvironment: From cancer initiation to

metastatic outgrowth. *Cancer Cell*. Mar 13 2023;41(3):374-403. doi:10.1016/j.ccell.2023.02.016 54. Mun JY, Leem SH, Lee JH, Kim HS. Dual Relationship Between Stromal Cells and Immune Cells in the Tumor Microenvironment. *Front Immunol*. 2022;13:864739. doi:10.3389/fimmu.2022.864739

55. Sherman MH, Beatty GL. Tumor Microenvironment in Pancreatic Cancer Pathogenesis and Therapeutic Resistance. *Annu Rev Pathol.* Jan 24 2023;18:123-148. doi:10.1146/annurev-pathmechdis-031621-024600

56. Herrera-Quintana L, Vázquez-Lorente H, Plaza-Diaz J. Breast Cancer: Extracellular Matrix and Microbiome Interactions. *Int J Mol Sci*. Jun 30 2024;25(13)doi:10.3390/ijms25137226

57. Papadas A, Huang Y, Cicala A, et al. Emerging roles for tumor stroma in antigen presentation and anti-cancer immunity. *Biochem Soc Trans*. Dec 20 2023;51(6):2017-2028. doi:10.1042/BST20221083

58. Yang L, Pang Y, Moses HL. TGF-beta and immune cells: an important regulatory axis in the tumor microenvironment and progression. *Trends Immunol*. Jun 2010;31(6):220-7. doi:10.1016/j.it.2010.04.002

59. Ge R, Huang GM. Targeting transforming growth factor beta signaling in metastatic osteosarcoma. *J Bone Oncol*. Dec 2023;43:100513. doi:10.1016/j.jbo.2023.100513

60. Mytilinaiou M, Nikitovic D, Berdiaki A, et al. Emerging roles of syndecan 2 in epithelial and mesenchymal cancer progression. *IUBMB Life*. Nov 2017;69(11):824-833. doi:10.1002/iub.1678

61. Chen L, Klass C, Woods A. Syndecan-2 regulates transforming growth factor-beta signaling. *J Biol Chem*. Apr 16 2004;279(16):15715-8. doi:10.1074/jbc.C300430200

62. de Fougerolles AR, Sprague AG, Nickerson-Nutter CL, et al. Regulation of inflammation by collagen-binding integrins alpha1beta1 and alpha2beta1 in models of hypersensitivity and arthritis. *J Clin Invest*. Mar 2000;105(6):721-9. doi:10.1172/JCI7911

63. Mezu-Ndubuisi OJ, Maheshwari A. The role of integrins in inflammation and angiogenesis. *Pediatr Res.* May 2021;89(7):1619-1626. doi:10.1038/s41390-020-01177-9

64. Di Martino JS, Nobre AR, Mondal C, et al. A tumor-derived type III collagen-rich ECM niche regulates tumor cell dormancy. *Nat Cancer*. Jan 2022;3(1):90-107. doi:10.1038/s43018-021-00291-9 65. Sánchez-Martín L, Estecha A, Samaniego R, Sánchez-Ramón S, Vega M, Sánchez-Mateos P.

The chemokine CXCL12 regulates monocyte-macrophage differentiation and RUNX3 expression. *Blood*. Jan 06 2011;117(1):88-97. doi:10.1182/blood-2009-12-258186

66. Berman JS, Cruikshank WW, Center DM, Theodore AC, Beer DJ. Chemoattractant lymphokines specific for the helper/inducer T-lymphocyte subset. *Cell Immunol*. Oct 01 1985;95(1):105-12. doi:10.1016/0008-8749(85)90299-0

67. Cruikshank WW, Berman JS, Theodore AC, Bernardo J, Center DM. Lymphokine activation of T4+ T lymphocytes and monocytes. *J Immunol*. Jun 01 1987;138(11):3817-23.

68. Mathy NL, Scheuer W, Lanzendörfer M, et al. Interleukin-16 stimulates the expression and production of pro-inflammatory cytokines by human monocytes. *Immunology*. May 2000;100(1):63-9. doi:10.1046/j.1365-2567.2000.00997.x

69. Smith AJ, Humphries SE. Cytokine and cytokine receptor gene polymorphisms and their functionality. *Cytokine Growth Factor Rev*. Feb 2009;20(1):43-59. doi:10.1016/j.cytogfr.2008.11.006

70. Buchrieser J, Oliva-Martin MJ, Moore MD, et al. RIPK1 is a critical modulator of both tonic and TLR-responsive inflammatory and cell death pathways in human macrophage differentiation. *Cell Death Dis*. Sep 24 2018;9(10):973. doi:10.1038/s41419-018-1053-4

71. He Z, Wang G, Wu J, Tang Z, Luo M. The molecular mechanism of LRP1 in physiological vascular homeostasis and signal transduction pathways. *Biomed Pharmacother*. Jul 2021;139:111667. doi:10.1016/j.biopha.2021.111667

72. Tawbi HA, Burgess M, Bolejack V, et al. Pembrolizumab in advanced soft-tissue sarcoma and bone sarcoma (SARC028): a multicentre, two-cohort, single-arm, open-label, phase 2 trial. *Lancet Oncol.* Nov 2017;18(11):1493-1501. doi:10.1016/S1470-2045(17)30624-1

73. Cavalli E, Ciurleo R, Petralia MC, et al. Emerging Role of the Macrophage Migration Inhibitory Factor Family of Cytokines in Neuroblastoma. Pathogenic Effectors and Novel Therapeutic Targets? *Molecules*. Mar 06 2020;25(5)doi:10.3390/molecules25051194

74. Fey RM, Nichols RA, Tran TT, Vandenbark AA, Kulkarni RP. MIF and CD74 as Emerging Biomarkers for Immune Checkpoint Blockade Therapy. *Cancers (Basel)*. May 04 2024;16(9)doi:10.3390/cancers16091773

75. Noe JT, Mitchell RA. MIF-Dependent Control of Tumor Immunity. *Front Immunol.* 2020;11:609948. doi:10.3389/fimmu.2020.609948

76. Yaddanapudi K, Putty K, Rendon BE, et al. Control of tumor-associated macrophage alternative activation by macrophage migration inhibitory factor. *J Immunol*. Mar 15 2013;190(6):2984-93. doi:10.4049/jimmunol.1201650

77. Barbosa de Souza Rizzo M, Brasilino de Carvalho M, Kim EJ, et al. Oral squamous carcinoma cells promote macrophage polarization in an MIF-dependent manner. *QJM*. Nov 01 2018;111(11):769-778. doi:10.1093/qjmed/hcy163

78. Zhang HF, Hughes CS, Li W, et al. Proteomic Screens for Suppressors of Anoikis Identify IL1RAP as a Promising Surface Target in Ewing Sarcoma. *Cancer Discov*. Nov 2021;11(11):2884-2903. doi:10.1158/2159-8290.CD-20-1690

79. Machado I, López-Guerrero JA, Scotlandi K, Picci P, Llombart-Bosch A. Immunohistochemical analysis and prognostic significance of PD-L1, PD-1, and CD8+ tumor-infiltrating lymphocytes in Ewing's sarcoma family of tumors (ESFT). *Virchows Arch*. May 2018;472(5):815-824. doi:10.1007/s00428-018-2316-2

80. Helm O, Held-Feindt J, Grage-Griebenow E, et al. Tumor-associated macrophages exhibit proand anti-inflammatory properties by which they impact on pancreatic tumorigenesis. *Int J Cancer*. Aug 15 2014;135(4):843-61. doi:10.1002/ijc.28736 81. Rodriguez AB, Engelhard VH. Insights into Tumor-Associated Tertiary Lymphoid Structures: Novel Targets for Antitumor Immunity and Cancer Immunotherapy. *Cancer Immunol Res.* Nov 2020;8(11):1338-1345. doi:10.1158/2326-6066.CIR-20-0432

82. Weng Y, Yuan J, Cui X, et al. The impact of tertiary lymphoid structures on tumor prognosis and the immune microenvironment in non-small cell lung cancer. *Sci Rep.* Jul 15 2024;14(1):16246. doi:10.1038/s41598-024-64980-y

83. Wrenn ED, Apfelbaum AA, Rudzinski ER, et al. Cancer-Associated Fibroblast-Like Tumor Cells Remodel the Ewing Sarcoma Tumor Microenvironment. *Clin Cancer Res.* Dec 15 2023;29(24):5140-5154. doi:10.1158/1078-0432.CCR-23-1111

84. Lanitis E, Dangaj D, Irving M, Coukos G. Mechanisms regulating T-cell infiltration and activity in solid tumors. *Ann Oncol.* Dec 01 2017;28(suppl_12):xii18-xii32. doi:10.1093/annonc/mdx238

85. Windhager J, Zanotelli VRT, Schulz D, et al. An end-to-end workflow for multiplexed image processing and analysis. *Nat Protoc*. Nov 2023;18(11):3565-3613. doi:10.1038/s41596-023-00881-0

Acknowledgements

We thank the Children's Hospital Los Angeles Pathology (Julie Galindo, William Gordo, and Monica Mendez), Spatial Biology and Genomics (Long Hung, Michael Hadjidaniel, Xiangming Ding, and Shahab Asgharzadeh), and Cellular Imaging (G. Esteban Fernandez) Cores. CK is supported by the St. Baldricks Foundation as the Shohet Family Fund St. Baldrick's Fellow, the A.P. Giannini Foundation, and the Coco Johnson Fellowship in Pediatric Sarcoma. JFA is supported by grants from the National Cancer Institute (U54CA268072 and P30CA014089) and by the Alfred E. Mann Family Foundation Chair in Cancer Research. MSBR is supported by grant T32GM086287 from the National Institute of General Medical Sciences (NIGMS), and startup funds from the Yale School of Medicine. This study was partly supported by the National Institutes of Health P30CA046934 by utilizing the Bioinformatics and Biostatistics Shared Resource. The opinions expressed are those of the authors and do not necessarily represent the thoughts or opinions of NIGMS, NIH, or the United States government. This work is dedicated to Noah Shohet, in loving memory of Noah Shohet and Stanley Kuo.

Author contributions

Christopher Kuo, Conceptualization, Funding, Data curation, Formal analysis, Investigation, Methodology, Validation, Visualization, Writing – original draft, Writing – review and editing; Krinio Giannikou, Formal analysis, Writing – Review and editing; Nuoya Wang, Formal analysis, Investigation, Methodology, Validation, Writing – review and editing; Mikako Warren, Formal analysis, Investigation, Validation, Writing – review and editing; Andrew Goodspeed, Data curation, Writing – review and editing; Nick Shillingford, Formal analysis, Writing – review and editing; Masanori Hayashi, Data curation, Writing – review and editing; Micha Sam Brickman Raredon, Formal analysis, Investigation, Methodology, Validation, Writing – review and editing; James F Amatruda, Conceptualization, Funding, Investigation, Methodology, Project administration, Supervision, Validation, Writing – review and editing.



Spatial Transcriptomics data

Figure 1: Overview of comprehensive Spatial Transcriptomic analysis. (A) Overview of experimental method, beginning with selection of biopsied formalin-fixed, paraffin-embedded primary tumor specimens from patients with localized Ewing sarcoma (EwS) and metastatic EwS. Tumor dense region was selected by an expert pathologist followed by fixation of tumor to Visium spatial gene expression slide followed by sequencing and data analysis. (B). Downstream analyses of ST datasets to further elucidate immune cell populations and spatial interactions.



Figure 2: Integrated analysis of all EwS tumors reveals intertumoral transcriptional

heterogeneity. (A) Representative Uniform Manifold Approximation and Projections (UMAPs) from two tumors. (B) UMAP of all 16 tumors pre-data integration. (C) UMAP of integrated 16 tumors. UMAP split by unique clustering of integrated dataset, disease status (localized vs. metastatic at diagnosis), and individual tumor samples. (D) Spatial mapping of clusters in primary tumors from select patients with metastatic or localized disease at diagnosis.



Figure 3: Transcriptomic annotation correlates with tumor histology. (A) Representative H&Estained image from tumor EWS4550 (from a patient with localized disease) with tumor region and fibrous stroma region as noted by an expert pathologist. Tumor purity and stromal scores generated by ESTIMATE transcriptional analysis correlate with H&E stained images. The ESTIMATE immune score is also shown. (B) Representative H&E-stained image from tumor EWS4521 (from patient with metastatic disease) with tumor, skeletal muscle and fibrous stroma region noted by an expert pathologist. ESTIMATE analysis of EWS4521 recapitulating H&E-stained image.



Figure 4: EwS tumors from patients with localized disease are enriched in extracellular matrix genes. (A) Gene set enrichment analysis (GSEA) for pseudo-bulk expression datasets of tumors from patients with localized and metastatic disease against Hallmark collection gene sets. (B) Heatmap of the most differentially expressed genes between pseudo-bulk expression EwS datasets from patients with localized and metastatic disease. (C) A set of 46 differentially expressed genes in tumors from patients with localized disease that overlap Hallmark epithelial mesenchymal transition gene sets. (D) Spatial mapping of spots enriched in the overlapping 46 extracellular matrix genes (ECM) on representative tumors (localized: EWS4550, EWS4539, and metastatic: EWS4064, EWS4521). (E) Distribution of the average calculated enrichment score per spot between tumors from patients with localized and metastatic disease.



Figure 5: Unique spatial gene signatures reveal colocalization of macrophages with stromal regions of EwS tumors. (A) Representative spatial gene enrichment mapping of localized tumor EWS4550. Spatial mapping of spots enriched in composite genes that annotate EwS, ECM, and immune cells (macrophage, T cells, cancer-associated-fibroblasts) in EWS4550. (B) Representative spatial gene enrichment mapping of metastatic tumor EWS4521 with composite genes that annotate EwS, ECM, and EwS, ECM, and immune cells in EWS4521.



Figure 6: Deconvolution analysis demonstrates heterogeneous immune cell populations and distinct spatial localization of immune cells. (A) Estimation of immune cell abundances in all 16 tumor samples through CIBERTSORTx. (B) Deconvolution CARD analysis of all 16 tumor samples using a publicly available EwS scRNAseq dataset as a reference with individual boxplots. Each cell population is depicted in a different color. (C) Spatial visualization of deconvoluted populations from two representative tumors (EWS4550 (localized), and EWS4521 (metastatic)).



Figure 7: Spatial ligand-receptor microenvironmental transcriptomic signals in localized and metastatic EwS Tumors. (A). NICHES ligand-receptor neighborhood to spatial spot microenvironmental clustering identify 4 unique microenvironmental clusters. (B). NICHES analysis demonstrate enrichment of ECM-related microenvironmental connectivities in loc-EwS tumors.



Figure 8: Ultrahigh-plex single cell spatial proteomic analysis of immune cells in EwS TME. (A) Single-cell spatial proteomic analysis demonstrates presence of T-cells, B-cells, macrophages and enrichment of ECM in a representative Loc-EwS tumor (EWS4564). (B) Single-cell spatial proteomic analysis demonstrates presence of T-cells, B-cells, macrophages and enrichment of ECM in a representative for the term of term of



Figure 9: Ultrahigh-plex single cell spatial proteomic analysis of immune microenvironmental signals in EwS TME. (A) Single-cell resolution of TME modulating signals with TGFB1, PD1-PDL1, MIF-CD74 and VISTA in a representative Loc-EwS tumor (EWS4564). (B) Single-cell resolution of TME modulating signals with TGFB1, PD1-PDL1, MIF-CD74 and VISTA in representative Met-EwS tumor (EWS4578).