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Article

Comprehensive Longitudinal Linear Mixed Modelling of CTCs Illuminates the Role of Trop2, EpCAM, and CD45 in CTC Clustering and Metastasis

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Simple Summary

Circulating tumor cells (CTCs) are the primary mediators of metastatic disease. After leaving the tumor, they can travel through the blood alone or with cluster partners to aid them on their way to distant metastatic sites. While CTCs are critical diagnostic tools for monitoring cancer progression, the biology underlying their behavior is obscured by their rarity and heterogeneity. To unravel the complexity of their actions, we undertook a large-scale longitudinal study to determine the factors that affect their presence, their clustering behaviors, and their impacts on disease progression.

Abstract

Background/Objectives: Breast cancer is the most commonly diagnosed cancer worldwide, with high rates of distant metastasis. While circulating tumor cells (CTCs) are the disseminatory units of metastasis and are indicative of a poor prognosis, CTC heterogeneity within individual patients, between breast cancer subtypes, and between primary and metastatic tumors within a patient obscures the relationship between CTCs and disease progression. EpCAM, its homologue Trop2, and a pan-Cytokeratin marker were evaluated to determine their contributions to CTC presence and clustering behavior over the study period. We conducted a systematic longitudinal analysis of 51 breast cancer patients during the course of their treatment to deepen our understanding of CTC contributions to breast cancer progression. **Methods:** 305 total blood samples from 51 metastatic breast cancer (mBC) patients were included in the study. Patients received diverse treatment schedules based on discretion of the practicing oncologist. Patients were monitored from July 2020 to March 2023, with blood samples collected at scheduled care appointments. Nucleated cells were isolated, imaged, and analyzed using Rarecyte® technology, and statistical analysis was performed in R using the lmerTest and lme4 packages, as well as in Graphpad Prism version 10.4.1. **Results:** Both classical CTCs (DAPI+, EpCAM+, CK+, CD45-) and Trop2+ CTCs were detected in the blood of breast cancer patients. A high degree of correlation was found between CTC biomarkers, and CTC expression of EpCAM, Trop2, and the presence of CD45+ cells all predicted cluster size, while Pan-CK did not. Furthermore, while analyses of biomarkers by receptor status revealed no significant differences between HR+, HER2+, and TNBC patients, longitudinal analysis found evidence for discrete trajectories of EpCAM, Trop2, and clustering behaviors between HR+ and HER2+ cancers after diagnosed metastasis. **Conclusions:** Correlation and longitudinal analysis revealed that EpCAM, Trop2, and CD45+ cells were predictive of CTC cluster presence and size, and highlighted distinct trajectories of biomarker change over time between HR+ and HER2+ cancers following metastatic diagnosis.

Keywords: EpCAM; Trop2; CTC; CTC cluster; breast cancer; metastasis; liquid biopsy Cytokeratin; CD45

1. Introduction

Breast cancer is the most frequently diagnosed cancer in the world, and results in an estimated 685,000 deaths every year, with this burden projected to increase substantially over the next two decades [1,2]. While great strides have been made with increasingly potent and specific chemotherapeutics such as antibody-drug conjugates [3,4], treatment remains challenging because breast cancer is a heterogeneous disease spanning diverse subtypes, each with differing biomarker expression, prognoses, and metastatic capacities [5–7]. However, across breast cancer and all cancer types, 90% of deaths are due to distant metastasis from the primary tumor [8,9].

Breast cancers are classified based on expression of the estrogen receptor (ER), the progesterone receptor (PR), and human epidermal growth factor receptor2 (HER2). In particular, tumors are often grouped into three categories: hormone receptor positive in the absence of HER2 expression (HR+), HER2-expressing in the presence or absence of hormone receptor expression (HER2+), or negative for all three biomarkers (triple negative, or TNBC). HR+ cancers carry the best prognosis and are targeted by many well-established therapeutics [10,11]. While HER2+ cancers are more aggressive and carry a higher metastatic load, anti-HER2 therapies are successfully applied in the clinic [12,13]. TNBC is the most aggressive and least-treatable subtype, with a 5-year mortality rate of 40%, a median survival time of 13.3 months after metastatic diagnosis, and recurrence rate as high as 25% [3,14,15].

Regardless of subtype, metastasis spreads via circulating tumor cells (CTCs), which disseminate from the primary tumor through the vasculature to secondary sites where they enter dormancy and/or trigger distant metastasis. Breast cancer CTCs are classically defined as CD45-, EpCAM+, and Cytokeratin 8/18/19+ (classical CTCs or cCTCs), and this definition has yielded important insights into how CTC shedding predicts survivorship [16]. EpCAM is a transmembrane glycoprotein with homeostatic roles in the maintenance of epithelial barrier integrity, while elevated EpCAM expression within solid tumors is associated with poor prognosis [17]. EpCAM is a central regulator in many critical tumorigenic processes, including cell proliferation, adhesion, and migration [12,17,18], and has been explored as a druggable target in many solid tumors. For example, efficacy of CAR-T cell immunotherapy targeting EpCAM has recently been demonstrated [19].

Cytokeratins (CK) are intermediate filament proteins known to regulate structural integrity of epithelial cells, and in addition to being markers for classical CTCs, are also widely overexpressed within solid breast tumors and associated with poor survivorship [20,21]. Cytokeratins 8 and 18 (CK8 and CK18) are co-expressed in healthy mammary glands, and while their expression in tumor tissue negatively correlates with rates of recurrence, ER status, and tumor grade, CK8/18 are expressed by breast cancer CTCs [21]. CK19 has roles in cell-adhesion, motility, maintenance of epithelial morphology, is linked to recurrence, and is also expressed by breast cancer CTCs [20–23]. Curiously, despite their shared presence on breast cancer CTCs, overexpression of CK18 and CK19 in tumor tissues correlate in opposite directions with disease progression, though this may differ by subtype [24]. Like EpCAM, CK18 and CK19 are both reliable diagnostic markers in addition to being promising therapeutic targets [19–21].

Despite advances made using the classical CTC paradigm, it is also clear that many metastasis-competent CTCs do not fit the classical definition. For example, downregulation of EpCAM and Cytokeratins (CK) in CTCs is known to be associated with a mesenchymal subtype, as is upregulation of Vimentin [25,26]. While the epithelial-to-mesenchymal transition (EMT) is typically seen as a key step in metastatic progression, it is not a requirement for breast cancer lung metastasis, and within-tumor heterogeneity in EMT status may confer complementary benefits as non-EMT cells are metastatically competent and EMT cells can facilitate chemoresistance and recurrence at metastatic sites [27].

Furthermore, certain tumor-specific markers such as HER2 or EGFR may not be present in CTCs due to tumor heterogeneity, early dissemination, epigenetic changes, and/or drug-induced selection events [5,25,28]. Indeed, biomarker discordance between CTCs and the tumors they shed from have frequently been documented [25,28–30]. For these reasons, expansion of biomarker-based classifications, biomarker-agnostic approaches such as microfluidics based on size/deformability/density, and negative depletion of CD45+ immune cells are employed to cast a wider net in characterizing neoplastic cells outside of classical CTC classification ([25,31–34]. To this aim, recent research from our lab and others have highlighted EpCAM's only homologue Trop2 [7,35,36], (also known as EpCAM2) as a promising CTC biomarker and target in aggressive and otherwise untreatable HER2+ and TNBC diseases [3,4]. Trop2 shares many functions with EpCAM, including stabilization of tight junction proteins and oncogenic roles in proliferation, adhesion, and migration [37,38], however they also undergo distinct post-translational modifications, show different patterns of expression in healthy and cancerous tissues [39,40], and while both allow for cell contractility, they may play discrete roles in finetuning adhesion and migration [41].

EpCAM and Trop2 themselves promote homophilic cell adhesion in addition to their role in stability tight-junction Claudins of healthy epithelia and neoplastic cells [37,38], and due to the potential relevance of these mechanisms in CTC motility, further longitudinal studies utilizing large datasets are necessary to unravel the underlying complexity of EpCAM's role in CTC dissemination [17,18,40,42]. Indeed, some evidence suggests that despite EpCAM's known roles as an adhesion molecule, surface expression can also impair adhesion [18].

CTCs in blood travel as single cells, clusters of homotypic CTCs, or in close association with CD45+ immune cells, platelets, or cancer-associated fibroblasts. Evidence suggests diverse benefits of clustering with stromal and immune cells, including immune evasion, anoikis resistance, resistance to shear- and oxidative stress, and 20-100x improved metastatic competency relative to single CTCs [6,43–45]. Additionally, non-tumor cells associated with CTCs may be a potent source of growth factors and cytokines which eventually aid in the establishment of the distant pre-metastatic niche [43,44,46]. While EpCAM and Trop2 are relatively weak cell adhesion molecules compared to the classical junction proteins such as E-Cadherin [17], this may be an advantage for forming clusters with tumor-resident immune cells prior to invasion of the surrounding connective tissue. EpCAM and/or Trop2 expression may allow for transient but stable cell-cell contacts to be maintained in the absence of contact inhibition and polarization, allowing increased motility and migratory capacity to co-occur with protection from stressors conferred by cluster formation [6,43–45]. This intermediate state with both epithelial and metastatic characteristics is a known trait of CTC clusters and may undergird successful survival in the blood stream and eventual dissemination [27,34,47–49].

Over a 32-month period, we collected and enumerated Trop2+ (T2CTCs) and cCTCs in an unbiased way, notably from the blood of breast cancer patients of all subtypes, covering diverse treatment regimens and metastatic stages. 205 blood samples from 51 patients were included in the cCTC dataset, with 100 blood samples from 26 patients in the T2CTC dataset, totaling over 6,000 output images for scoring. We undertook correlative analysis and longitudinal analysis via linear mixed effects modelling in order to illuminate the roles of CK8/18/19, the EpCAM family, and CD45+ cells in CTC clustering. We hypothesized that EpCAM, Trop2, and CK would all significantly predict CTC cluster presence and size, as well as diversity of available sites for distant metastasis.

2. Materials and Methods

2.1. Study Design and Participants

Patients with metastatic breast cancer provided informed consent in accordance with IRB protocols, and all patient identities were anonymized prior to receipt by our technicians. Blood was collected into sodium-EDTA tubes and all blood samples began processing within 4 hours of retrieval. Patient parameters for the classical and Trop2 CTC datasets are shown in Tables 1 and 2.

Table 1. Clinical characteristics of patients included in the classical CTC dataset

Clinical Characteristic	Category	Full Cohort	HR+/HER2-	HR-/HER2+	TNBC	Fisher exact test, p-value
Age at 1 st blood collection	<65	30 (60.0)	17 (54.8)	11 (68.8)	2 (66.7)	0.792
	65+	20 (40.0)	14 (45.2)	5 (31.2)	1 (33.3)	
	Total	50 (100)	31 (100)	16 (100)	3 (100)	
Number of Metastatic sites	1	4 (8.0)	4 (12.9)	0 (0.0)	0 (0.0)	0.417
	2	15 (30.0)	7 (22.6)	7 (43.8)	1 (33.0)	
	3+	31 (62.0)	20 (64.5)	9 (56.2)	2 (66.7)	
	Total	50 (100)	31 (100)	16 (100)	3 (100)	
Lung metastasis	N	22 (43.1)	14 (43.8)	7 (43.8)	1 (33.0)	1.000
	Y	29 (56.9)	18 (56.2)	9 (56.2)	2 (66.7)	
	Total	51 (100)	32 (100)	16 (100)	3 (100)	
Bone metastasis	N	14 (27.5)	6 (18.8)	7 (43.8)	1 (33.0)	0.141
	Y	37 (72.5)	26 (81.2)	9 (56.2)	2 (66.7)	
	Total	51 (100)	32 (100)	16 (100)	3 (100)	
Liver metastasis	N	25 (50.0)	16 (51.6)	7 (43.8)	2 (66.7)	0.816
	Y	25 (50.0)	15 (48.4)	9 (56.2)	1 (33.3)	
	Total	50 (100)	31 (100)	16 (100)	3 (100)	
Brain metastasis	N	34 (68.0)	24 (77.4)	9 (56.2)	1 (33.3)	0.105
	Y	16 (32.0)	7 (22.6)	7 (43.8)	2 (66.7)	
	Total	50 (100)	31 (100)	16 (100)	3 (100)	

Of the 51 patients in the cCTC dataset, 1 was missing age, 1 did not have full metastatic site data.

Table 2. Clinical characteristics of patients included Trop2 CTC dataset.

Clinical Characteristic	Category	Full Cohort	HR+/HER2-	HR-/HER2+	TNBC	Fisher exact test, p-value
Age at 1 st blood collection	<65	16 (64.0)	7 (50.0)	7 (87.5)	2 (66.7)	0.246
	65+	9 (36.0)	7 (50.0)	1 (12.5)	1 (33.3)	
	Total	25 (100)	14 (100)	8 (100)	3 (100)	
Number of Metastatic sites	1	2 (7.7)	2 (13.3)	0 (0.0)	0 (0.0)	0.834
	2	9 (34.6)	4 (26.7)	4 (50.0)	1 (33.0)	
	3+	15 (57.7)	9 (60.0)	4 (50.0)	2 (66.7)	
	Total	26 (100)	15 (100))	8 (100)	3 (100)	
Lung metastasis	N	13 (50.0)	8 (53.3)	4 (50.0)	1 (33.0)	1.000
	Y	13 (50.0)	7 (46.7)	4 (50.0)	2 (66.7)	
	Total	26 (100)	15 (100)	8 (100)	3 (100)	
Bone metastasis	N	7 (26.9)	2 (13.3)	4 (50.0)	1 (33.0)	0.139
	Y	19 (73.1)	13 (86.7)	4 (50.0)	2 (66.7)	
	Total	26 (100)	15 (100)	8 (100)	3 (100)	
Liver metastasis	N	11 (42.3)	6 (40.0)	3 (37.5)	2 (66.7)	0.724
	Y	15 (57.7)	9 (60.0)	5 (62.5)	1 (33.3)	
	Total	26 (100)	15 (100)	8 (100)	3 (100)	
Brain metastasis	N	18 (69.2)	11 (73.3)	6 (75.0)	1 (33.3)	0.418
	Y	8 (30.8)	4 (26.7)	2 (25.0)	2 (66.7)	
	Total	26 (100)	15 (100)	8 (100)	3 (100)	

*Of the 26 total patients included in the T2CTC dataset, 1 did not have a listed age.

2.2. RareCyte® Sample Processing, Scanning, and Analysis

Upon receipt of patient blood in EDTA tubes, blood was incubated, fractionated, and mounted per the manufacturer's instructions. In brief, a maximum of 7.5 mL of blood was transferred to an AccuCyte® Blood Collection Tube for a period of 24-48 hours prior to processing. Blood was then removed from the collection tubes and dispensed into the AccuCyte Separation Tube, centrifuged to remove red blood cells and enrich for nucleated cells. An additional round of centrifugation isolated the nuclear cell layer into cell isolation fluid (RareCyte, Seattle, WA, USA, 24-1090-002), which was applied as a monolayer to a microscope slide using the Cytospreader Slide Preparation Device®. Slide-mounted

Classical CTCs were stained according using the RareCyte Rareplex 0700-MA staining protocol, while Trop2+ CTCs were stained according to the Rareplex 1200-MA staining protocol, with a final concentration of 1:100 of anti-Trop2 antibody (ECM Biosciences, Aurora, CO, USA, TM0051). All slides were scanned using the Cytfinder II® imaging platform and analyzed with CyteHub® software. Images output by CytHub were then quality-checked by trained technicians to confirm classification as individual CTCs and/or clusters. Figures 1A and 1B show representative CyteHub output images from the 0700-MA and 1200-MA staining procedures from HR+ and HER2-expressing cancers, respectively. Figure 1C shows an image of an exceptionally large CTC cluster from a HER2-expressing cancer, originally captured by CyteHub® and subsequently imaged via confocal microscopy.

2.3. Confocal Imaging

High-resolution confocal imaging of the cluster in Figure 1C was performed on the Zeiss LSM 800 Airyscan using the 63x oil objective.

2.4. Endpoints and Assessments

Because of staggered periods of enrollment and differing frequency of sample availability from each patient, a clear clinical endpoint was not set. Rather, each patient was analyzed relative to their initial diagnosis date with mBC as well as diagnosis dates for each unique metastatic disease. Clinical parameters and diagnosis dates were provided by the practicing oncologist. Sample receipt and analysis spanned from July 2020 through March 2023.

2.5. Statistical Analysis

Due to non-normality and zero-skewing in the patient biomarker datasets, data were base 2 log-transformed after the addition of 1 to each original observation. Correlation coefficient (r) values are listed, and correlation analysis was performed alongside linear regressions for all biomarker pairs, as shown in Figures 2 and 3. The classical CTC (cCTC) and Trop2+ CTC (T2CTC) dataset each evaluated 8 biomarkers, however the cCTC dataset includes EpCAM and pan-CK on different fluorophores, the T2CTC dataset does not, because the 1200-MA staining requires that EpCAM and CK be measured with the same fluorophore. Longitudinal data analysis was performed based on a linear regression model with random coefficients according to the formula $Y = a + b x t + \epsilon$, where a represents the initial biomarker value, decomposed into a fixed effect a_0 , and a random effect a_1 , while b represents the slope, decomposed into b_0 and b_1 as above. Both datasets were fitted to the full model described above to determine if biomarker trajectories differed by receptor status (HR+ and HER2+), as there were not sufficient patients or collection events to include TNBC in the longitudinal analysis. In the full model, the intercept a_0 and slope b_0 differ between the two receptor statuses. Where evidence was insufficient to detect differences by receptor status, data was fitted to a reduced model to determine if the slope was significantly non-zero for each biomarker value over the analysis period. In the reduced model, the intercept a_0 but not the slope b_0 differ by receptor status. Longitudinal analysis was performed with the date of each patient's first diagnosis with metastatic breast cancer standardized as time 0. For longitudinal analysis by metastatic site, time 0 represents each patient's

diagnosis with metastasis to the indicated site. All statistical analysis was performed in R using the lmerTest and lme4 packages. Data visualization for Figure 4 was performed in Graphpad Prism version 10.4.1. We used the markdown-file tool implemented in R for coding and creating dynamic documentation in statistical analysis to ensure scientific rigor and reproducibility.

2.6. Data Availability

Data are available upon request but are not publicly available for protection of patient privacy.

3. Results

3.1. RareCyte Reveals Expression of Trop2 in Breast Cancer Patient CTCs and High Inter-Marker Correlation

Analysis of breast cancer patient blood by RareCyte technology revealed the clear presence of c-CTCs (EpCAM+, PanCK+, CD45-, Figure 1A) as well as the presence of T2CTCs (Figures 1B and 1C). Additionally, both classical and T2CTCs were found as singlets and in homotypic and heterotypic cluster configurations (Figure 1A-C).

As expected, correlation analysis revealed a high degree of correlation between a majority of the included CTC biomarkers. Figures 2 and 3 list biomarkers in the diagonals, with each box on the lower left representing scatter plots with the regression line of the two biomarkers in the same row and column, and each box in the upper right representing the correlation coefficient of the same two biomarkers in that same row and column, symmetrically about the diagonal. The tables included in both figures list all strong correlations (those with an $r \geq 0.7$). In the cCTC dataset (Figure 2), expression of EpCAM and the presence of CD45+ cells in cluster with CTCs were highly predictive of both cluster presence and size (highlighted in red text), with a correlation coefficient of 0.78 between EpCAM+ CTCs and the presence of CD45+ cells. No strong correlations were observed between pan-Cytokeratin expression and any of the other evaluated biomarkers.

Likewise in the T2CTC dataset, cluster presence and size was highly correlated with the presence of CD45+ cells, as well as by positivity for the CK/EpCAM channel jointly with Trop2 (Figure 3). Together this data strongly suggests that the EpCAM family and CD45+ cells are facilitative of CTC cluster formation and intravasation to the blood of breast cancer patients.

3.2. No Individual CTC Biomarker Condition Was Significantly Associated with Receptor Status

In order to determine whether patterns of individual biomarker expression were associated with a particular receptor status, we compared biomarker values across all patients in both the cCTC (Figure 4 A-D) and T2CTC datasets (Figure 4 E-F). In cases where patients had more than 1 sample analyzed, each patient is represented by the average value for each biomarker to avoid pseudo-replication. ANOVA revealed all associations to be non-significant. Results are shown in Table 1 (cCTC) and 2 (T2CTC).

3.3. Longitudinal Analysis Reveals Differences in Clustering Behavior Between HR+ and HER2+ Cancers

In order to determine whether biomarkers changed over the course of multiple measures, we performed longitudinal analysis on HR+ and HER2+ patients for both the cCTC (Figures 5 and 7) and T2CTC datasets (Figures 6 and 8) by linear mixed modelling. Figure 5 A-F show spaghetti plots [50] with trendlines for clustering behaviors in the cCTC dataset, with those for EpCAM in G and H. Full model results are shown in Table 3.

EpCAM and the number of clusters containing 2 or more cells (Clusters >2) showed statistically significant differences, ($p = 0.02183$ and $p = 0.00691$, respectively) in slope between HR+ and HER2+ cancers, indicating different biomarker rates of change over time between receptor subtypes. In particular, analysis indicates that these two parameters increased over time in HER2+ cancers during the analysis period relative to first metastatic diagnosis, while there is no significant change with time

in HR+ cancers. Clusters showed marginal significance ($p = 0.0675$) in the same direction, and all other comparisons were non-significant.

Figure 6 A-F show spaghetti plots with trendlines for clustering behaviors in the T2CTC dataset, with those for Trop2 in **G** and **H**. Full model results are shown in Table 4. All comparisons by receptor status were non-significant.

Table 3. Full longitudinal model results by receptor status for all biomarkers in the cCTC dataset. .

Biomarker (per mL)	Receptor Effect	Time Effect	Interaction
cCTCs	0.6591061	0.7461176	0.840692
CK	0.4685809	0.6062315	0.4074604
EpCAM	0.7158591	0.1999038	0.0218357
Clusters	0.9510659	0.028952	0.0674635⁺
Clusters of 2	0.6826174	0.0805827	0.1121893
cCTC in cluster	0.4323639	0.5820074	0.5686966
Clusters >2	0.5347161	0.0062573	0.0069147
CD45 in cluster	0.9598183	0.0401535	0.1091464

⁺ Indicates a marginally significant result

Table 4. Full longitudinal model results by receptor status for all biomarkers in the T2CTC dataset. .

Biomarker (per mL)	Receptor Effect	Time Effect	Interaction
CK/EpCAM/Trop2	0.4802839	0.8340946	0.8919177
CK/EpCAM	0.5649867	0.0955775	0.1789835
Trop2	0.5484958	0.8870074	0.490742
Clusters	0.966292	0.2493243	0.5708911
CTC in cluster	0.7295486	0.5810308	0.8777693
CD45 in cluster	0.9998403	0.2495356	0.5364823
Clusters of 2	0.2829963	0.6461081	0.4631998
Clusters >2	0.633026	0.2716701	0.4975725

To determine if biomarker trajectories over time differed by metastatic site, independent models were employed for each of four metastatic sites: brain, liver, bone, and lungs. The model for each metastatic site included analysis of all biomarkers, with clustering behaviors and EpCAM-family expression shown in Figure 7 for the cCTC dataset and in Figure 8 for the T2CTC dataset. Complete results for the full and reduced models for each metastatic site in the cCTC dataset are shown in Table S1 and T2CTCs in Table S2. In the cCTC dataset, the full model found no significant differences between HR+ and HER2+ cancers by metastatic site. The cCTC reduced model found that relative to first diagnosis of lung metastasis, Clusters >2 ($p = 0.07211$), CD45 in cluster ($p = 0.07763$) had marginally non-zero slopes. Relative to diagnosis with bone metastasis, slopes for Clusters >2 ($p = 0.06979$) and CD45 in cluster (0.09091) were marginally non-zero. Relative to diagnosis with brain metastasis, Clusters >2 (0.07983) were marginally non-zero. Likewise for liver metastasis, Clusters >2 per ml ($p = 0.08623$).

In the T2CTC dataset, a baseline effect of Trop2 expression was found for the bone metastatic condition ($p = 0.04255$), with a marginally significant difference in slope between those with bone metastasis and without ($p = 0.05163$). Clusters >2 was significant both at baseline ($p = 0.00882$), and with respect to slope ($p = 0.02084$), indicating an increase in larger clusters over time after diagnosis with brain metastasis. The reduced model found marginally non-zero slopes by lung metastasis for Clusters ($p = 0.08503$) and by brain meta-stasis for CD45 in Cluster ($p = 0.07605$). All other comparisons were non-significant across both datasets.

4. Discussion

Due to their roles as epithelial cell adhesion molecules, the EpCAM family has long been suspected to play a fundamental role in CTC invasion, intravasation, circulation, dissemination, and eventual metastasis, though their mechanisms of action during tumor progression have remained elusive. EpCAM is employed as the primary discriminating diagnostic marker in breast cancers, and decades of study have confirmed its salience in predicting overall survival and progression free survival [12,42]. With the rationale of discriminating CTC heterogeneity and the widely acknowledged need to expand biomarker criteria in CTC identification, we incorporated EpCAM homologue Trop2 in our analyses, and our results indicate that EpCAM and Trop2 have overlapping but distinct roles in CTC shedding, clustering, and homing to distant metastatic sites. In support of this hypothesis, both EpCAM+ CTCs and Clusters >2 increased after first metastasis in HER2+ cancers but not HR+, suggesting discrete evolutionary trajectories in CTC clustering behavior between the receptor statuses.

While the full cCTC model comparing trajectories of biomarkers between HR+ and HER2+ disease did not find any significant association between metastatic site and biomarker status, the full T2CTC model did find a significant baseline effect of Trop2 expression and a marginally significant difference in slope between receptor statuses after diagnosis with bone metastasis. While the smaller size of the T2CTC dataset relative to the cCTC dataset does warrant caution, the results suggest that shedding of Trop2+ CTCs may be a feature of bone metastatic disease. Additionally, larger clusters (those with greater than 2 cells) differed both at baseline and by slope between receptor subtypes after diagnosis with brain metastasis, suggesting that larger cluster size corresponds more to brain metastasis in HER2+ disease than in HR+. The reduced cCTC models by metastatic status revealed marginally non-zero slopes for Clusters >2 and clustering with CD45+ cells after diagnosis with lung bone metastasis, as well as for Clusters >2 for brain and liver metastasis. The reduced T2CTC model found marginally non-zero slopes for cluster present after diagnosis with lung metastasis and CD45+ immune cells in clusters after brain metastasis.

Taken together, modelling by metastatic site suggest that certain sites are predisposed to shedding larger heterotypic clusters containing CD45+ cells. Larger clusters are known to be associated with poor prognosis [34,51], and prior research from our group has demonstrated a transcriptomic signature underlying breast cancer brain metastasis [52]. Importantly, to our knowledge this is the first study using patient-derived CTCs to tie specific CTC cluster configurations with specific organ metastatic sites [53]. Although recent investigations on organotropism of breast cancer subtypes has been performed with cell lines using microfluidic chips, SKBR3, the Trop2-high, HER2-expressing cell line employed, did not show substantial bone organotropism [54].

Contrary to our hypothesis, while the pan-cytokeratin marker targeting CK8/9/19 was detected in many patients, it did not correlate strongly ($r > 0.7$) with any other CTC biomarker or clustering behavior. This is surprising given the role of CKs in maintenance of cell morphology and in particular the role of CK19 in cell-cell adhesion. However, this mechanism is mediated by E-Cadherin, which is frequently lost in the partial EMT state [20,21]. Because CK19-E-cadherin interactions result in stronger cell adhesion and cell polarization that may impede clustering and migration, the relationships between clustering behaviors and EpCAM and Trop2 may be due to their ability to mediate both homophilic adhesion and tight-junction mediated adhesion, allowing simultaneously for cytoskeletal flexibility, maintenance of a depolarized state, and preservation of cell-cell contacts. EpCAM+ CTCs expressing low levels of CK8/18/19 correspond with decreased overall survival [21,55]. Our results suggest that this could be a function of clustering behaviors in the hybrid EMT state. This hypothesized dual role in clustering behavior is consistent with the partial/hybrid EMT phenotype which may confer the plasticity found in the most metastatically competent CTC clusters [5,27,28,34,35,51,56], and further underscores the complexity of oncogenic patterns of differentiation and dedifferentiation. Indeed, while CK19-KO cells have improved motility, they are less able to form tumor mammospheres [22], suggesting a complex role whereby they are not integral to invasion and cluster formation, but may aid in guiding structured micrometastasis upon dissemination.

Our analysis of biomarker values by subtype did not reveal any significant associations, although all biomarker values for a given patient were averaged in this analysis to avoid pseudoreplication. Given high inter- and intra-patient heterogeneity in circulating EpCAM+ cells shown here (Figures 5 and 6) and elsewhere [42], use of averaged values may have obscured underlying patterns that differed by receptor subtype.

Conversely, our longitudinal analysis, which incorporates intra-patient heterogeneity in CTC biomarkers, revealed that biomarker trajectories did in fact differ significantly between subtypes (Figure 5 and Table 3). Furthermore, as previous investigations yielded conflicting reports as to how CTC number and presence differs between subtypes [25,28–30,42], we suggest that this disparity may in part be explained by high intra-patient heterogeneity, discrete biomarker trajectories between subtypes, and unaccounted-for EpCAM-negative CTCs.

Correlational analysis also revealed CD45+ cell presence in clusters also correlated with cluster size, independent of EpCAM and Trop2 expression. While CD45+ cells may be more likely to develop connections to larger clusters in a probabilistic manner, and the EpCAM family may have an indirect effect on CD45+ cells clustering with tumor cells mediated by greater cluster size, our analysis cannot establish a causal link between CD45+ cells in clusters with EpCAM+ and Trop2+ CTCs. However, EpCAM is capable of mediating homotypic interactions between the intestinal epithelia and epithelia-resident lymphocytes [18,57], which is highly suggestive of a possible role in cluster formation prior to dissemination.

There are limitations to this study. First, the smaller size of the T2CTC dataset and the marginal significance of the results by metastatic site merit caution, however given the consistency of the effect patterns despite the substantial confounds due to tumor heterogeneity and treatment status, we contend these effects reveal an underlying mechanism. Second, while our data suggest that EpCAM and Trop2 play roles in the presence and size of CTC clusters, there are other known biomarkers involved in cluster formation unaccounted for in our dataset, including plakoglobin [6], CD44 [56], and ICAM [56]. CK19's interactions with plakoglobin and E-Cadherin are complex and context-dependent, further complicating interpretation. Third, some known bidirectional molecular interactions such as those between HER2 and CK19 [21,58] are not analyzable in our datasets due to our use of the pan-CK marker which stains CK8, CK18, and CK19. Finally, CK19 may serve to impair or facilitate metastasis depending on cancer stage [22].

Altogether, our results illustrate the critical roles of EpCAM and Trop2 in CTC clustering behavior after metastatic diagnosis, giving clarity to a clinical literature largely confounded by inter- and intra-patient heterogeneity in CTC shedding and clustering behavior. Further studies are needed to interrogate the roles of EpCAM and Trop2, and in particular whether their actions in Claudin stabilization and homophilic adhesion differentially regulate CTC cluster formation and metastatic competency.

5. Conclusions

CTC shedding is a complex process, and given the rarity of CTCs and the further rarity of metastatically competent CTCs and CTC clusters, our work is an important first step in unravelling the mechanisms underlying successful metastasis. Correlation and longitudinal analyses revealed that EpCAM, Trop2, and CD45 expression were highly predictive of cluster presence and size, and highlighted distinct trajectories of biomarker change over time between HR+ and HER2+ cancers. By incorporating analysis of EpCAM, Trop2, and CK8/18/19, we were able to shed light on the roles of these CTC markers in clustering and metastatic trajectories, finding that EpCAM and Trop2, but not CK, significantly affected clustering behavior and distant metastasis. Future studies can build on our research to illuminate the mechanistic roles of EpCAM, Trop2 and the partial EMT in CTC clustering and organotropism, in turn yielding clinically actionable strategies to undermine the competency of these deadly metastatic seeds.

Supplementary Materials: The following supporting information can be downloaded at the website of this paper posted on Preprints.org.

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Data Availability Statement: Data are available upon request but are not publicly available for protection of patient privacy.

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Abbreviations

The following abbreviations are used in this manuscript:

CTC	Circulating Tumor Cell
cCTC	Classically defined Circulating Tumor Cell
T2CTC	Trop2-expressing Circulating Tumor Cell
mBC	Metastatic Breast Cancer
TCCP	Total Cancer Care Protocol
CK	Cytokeratin
EpCAM	Epithelial Cell Adhesion Molecule
HER2	Human epidermal growth factor receptor 2
EGFR	Epidermal growth factor receptor

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