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Genomic evidence supports a clonal diaspora model for metastases of esophageal adenocarcinoma

DOI: 10.1038/s41588-019-0551-3

Document Version

Accepted author manuscript

Link to publication record in Manchester Research Explorer

Citation for published version (APA): Noorani, A., Li, X., Goddard, M., Crawte, J., Alexandrov, L. B., Secrier, M., Eldridge, M. D., Bower, L., Weaver, J., Lao-Sirieix, P., Martincorena, I., Debiram-Beecham, I., Grehan, N., MacRae, S., Malhotra, S., Miremadi, A., Thomas, T., Galbraith, S., Petersen, L., ... Fitzgerald, R. C. (2020). Genomic evidence supports a clonal diaspora model for metastases of esophageal adenocarcinoma. *Nature Genetics*. https://doi.org/10.1038/s41588-019-0551-3

Published in: Nature Genetics

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1 1. Extended Data

Figuro #	Figure title	Filonamo	Figure Legend
Extended Data	One sentence only Flowchart	This should be the name the file is saved as when it is uploaded to our system. Please include the file extension. i.e.: <i>Smith_ED</i> <i>Fig1.jpg</i> Noorani et	If you are citing a reference for the first time in these legends, please include all new references in the Online Methods References section, and carry on the numbering from the main References section of the paper.
Fig. 1	describing key steps taken to construct	al_Extended Data1_201911 18.tif	reconstruction is further elaborated in Supplementary methods, Mutation clustering and phylogenetic tree
	trees		
Extended Data Fig. 2	Phylogenetic tree construction for example case S3	Noorani et al_Extended Data2_201911 18.tif	1) Battenberg algorithm to determine total copy number (purple line) and minor allele (blue line). Y-axis =number of chromosome copies, X-axis= chromosome and position. The average ploidy, aberrant cell fraction (cellularity) and goodness of fit to the model are shown for each sample, Primary E1, E2, Lymph node L1 and Distant metastasis D1. The goodness of fit is a measure of the amount of the genome with clonal, rather than subclonal copy number states. D1 has a subclonal mix of different copy number states resulting in noninteger total copy number, for example on chromosome 2, resulting in a goodness of fit below 100%. 2) Bayesian Dirichlet Process to cluster SNVs based on CCF in each sample. The density plots show the posterior probability of a mutational cluster, these are produced for every pair of samples and selected plots are shown High density at CCF of (0,0) indicates subclones that are not present in the pair of samples shown in a particular plot. 3) Clustering of results – Clusters are identified as local maxima in the posterior
			SNVs assigned to each cluster, and their associated CCFs. 4) Unscaled Tree construction using the sum rule and crossing rule as detailed in Supplementary Methods p25. 5) Final Tree -The tree is drawn as seen in Figures 2 and Extended Data2, branch lengths are proportional to

			the number of SNVs assigned to each
			subclone. Scales vary on a per case basis
			depending on the total number of SNVs, in
			order to fit cases on one figure. Trees are
			annotated with the gene names of known
			drivers, and the colour of each branch
			represents a trunk (pink), branch (purple)
			or leaf (yellow). The grey circles represent
			clones and subclones and their CCFs are
			shown in Supplementary Table5 and 6.
Extended Data	Phylogenetic	Noorani et	F=esophagus, D=distant organ, L=lymph
Fig. 3	trees of cases in	al Extended	node. B= Barrett's. For precise anatomical
8	cohort with only	Data3 201911	locations refer to Supplementary Table3
	nodal or distant	18 tif	and 4 MRCA-most recent common
	organ disease as	10.00	ancestor Pink-trunk (shared events)
	derived from H		Burnle-branch (shared by more than one
			sample) Vollow-loof (unique to one
	003		sample). Grow date at the end of the lines
			(truncal branches or loaves) represent
			(truncal, branches of leaves) represent
			subciones of ciones, whose CCFs are
			snown in Supplementary Tables5 and 6.
			Trees are annotated with key driver events
			as identified from the literature
			Black=point mutations, Red=copy number
			alterations, purple= structural variants. The
			adjacent scales are relative to the number
			of SNVs in that particular case and hence
			constructed on a case by case basis.
Extended Data	Structural	Noorani et	a. Similarity matrix based clustering for all
Fig. 4	variation of 18	al_Extended	SVs 122 genomes across 19 cases. SVs
	metastatic	Data4_201911	were deemed to refer to the same
	esophageal	18.tif	rearrangement event across cases if their
	adenocarcinoma		corresponding breakpoint locations fell
	cases		within a window of maximum 50 bp. The
			individual sample types
			are shown as a separate row on the x axis
			with the color key depicting the sample
			type. The purple scale indicates the
			number of shared SVs. (L=lymph nodes;
			M=metastasis; T=tumor). b. Histogram
			showing the percentage of rearranged
			genes that are concordant, unique to
			tumors and unique to metastases. Two-
			tailed Welch test P=0.2674 demonstrating
			no overall difference between total
			number of SVs in primary, local lymph
			nodes and distant metastases c. Stacked
			bar charts showing the composition of
			various SVs in each sample on a per patient
			basis INV= inversion, ME= mobile element,
			BND= translocation DEL=deletion,
			DUP=duplication, INS= insertion.

	1	1	
Extended Data	Random	Noorani et	The number of mutations detected
Fig. 5	simulation model	al_Extended	correlates strongly with the CCF of the
	for S-WGS	Data5_201911	cluster (Pearson r=0.992, n=100). Number
	cluster detection	18.tif	of mutations in each cluster =1000.
Extended Data	Correlation of	Noorani et	Pearson correlation coefficient is above
F1g. 6	fraction of	al_Extended	0.97 for clusters with 200 or more
	mutations	Data6_201911	mutations.
	detected with	18.tif	
	CCF as a function		
	of cluster size		
Extanded Data	S-WGS uald	Neeroniet	
Fig 7	domonstrating	Noorani et	
1 lg. /	the Pearson	Data7 201011	
	correlation	18 +if	
	coefficient of	10.0	
	VAF at 1xWGS		
	and High Depth		
	Resequencing		
	(n=33)		
Extended Data	Detection of	Noorani et	SNVs and indels from all cases (n=18) were
Fig. 8	Selection in	al_Extended	aggregated into 4 different subsets: clonal
	subsets of	Data8_201911	= variants found in the MRCA (n=378453);
	mutations	18.tif	subclonal = variants not found in the MRCA
			(n=516136); pre-diaspora = variants found
			above the diaspora founder clone in the
			phylogenetic tree (n=313545); post-
			diaspora = variants found in the diaspora
			founder or
			In clones below the founder in the
			subset dN/dS applycis was performed
			subset, div/us analysis was performed
			truncating variants. Bars show maximum
			likelihood estimates of dN/dS values with
			values greater than 1 (dashed line)
			indicating positive selection. Vertical lines
			= 95% confidence intervals, estimated
			using Wald test.
Extended Data	Percentage of	Noorani et	Stacked horizontal bar chart representing
Fig. 9	truncal and	al_Extended	the percentage of truncal and branch
	branch clusters	Data9_201911	clusters present in tissue from earlier time-
	in tissue from	18.tif	points on the x-axis and the Case ID on the
	earlier time-		y-axis. P1 diagnosis* is a frozen sample,
	points		while the rest are FFPE. Blue = truncal,
			maroon = branch, grey = not present. The
			number of clusters (n) is demonstrated for
Factor 1 - 1 D -		Negaratist	each case.
Extended Data	ctDNA analysis	Noorani et	Digital PCR traces of mutant allele fraction
FIG. IU	i irom historical	al_Extended	1 IOF 1P53 ON THE Y-AXIS AND DAYS FROM

plasma samples	Data10_20191	diagnosis on the X-axis, and grey areas
	118.tif	indicate periods of therapy. Where
		subclones and clones are seen at 1xWGS
		on plasma, they are highlighted on the 50x
		phylogenetic tree (coloured blue). The
		samples in which these subcloens and
		clones are present in are shown in
		Supplementary Table3. There was no TP53
		data for S3 as it was wild type for TP53
		mutations. Copy number traces for P1 are
		shown, with the arrow demonstrating an
		area of MET amplification.

2. Supplementary Information:

- A. Flat Files

Item	Present?	Filename This should be the name the file is saved as when it is uploaded to our system, and should include the file extension. The extension must be .pdf	A brief, numerical description of file contents. i.e.: Supplementary Figures 1-4, Supplementary Note, and Supplementary Tables 1-4.
Supplementary Information	Yes	Noorani et al_Supplementar yInformation_20 191118.pdf	Supplementary Figures 1-5, Supplementary Tables 1-16, Supplementary Note
Reporting Summary	Yes	Reportingsumma ry.pdf	

9 B. Additional Supplementary Files

	Number If there are multiple files of the same type this should be the	Filename This should be the name the file is saved as when it is uploaded to our system, and should include the file	Legend or Descriptive
	numerical indicator.	extension. i.e.: Smith_	Caption
	i.e. "1" for Video 1, "2"	Supplementary Video	Describe the contents
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		Noorani et	
		al_SupplementaryTabl	Excel Spreadsheets for
		es_Excel_20191118.xls	Supplementary Table
Supplementary Table	1	x	5, 6, 7, 8, 9, 11, 16

11 Genomic evidence supports a clonal diaspora model for metastases of esophageal

- 12 adenocarcinoma
- 13
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37 Abstract (95 words)

38 The poor outcomes in esophageal adenocarcinoma (EAC) prompted us to interrogate the 39 pattern and timing of metastatic spread. Whole genome sequencing and phylogenetic 40 analysis of 388 samples across 18 EAC cases demonstrated in 90% of cases that multiple 41 subclones from the primary tumor spread very rapidly from the primary site to form 42 multiple metastases, including lymph nodes and distant tissues, a mode of dissemination 43 that we term 'clonal diaspora'. Metastatic subclones at autopsy were present in tissue and 44 blood samples from earlier time-points. These findings have implications for our 45 understanding and clinical evaluation of EAC.

46

47 Introduction

48 Metastatic spread to distant sites accounts for the majority of cancer deaths¹. 49 Understanding the anatomical extent of disease is essential to determine the optimum 50 treatment strategy. This is challenging since cancer continually evolves at a microscopic 51 scale, often beyond the resolution of clinical imaging techniques. Furthermore, the patterns 52 of metastatic spread are often unpredictable in terms of time-course and anatomical 53 location. Treatments may therefore be unnecessarily toxic (e.g. radical lymphadenectomy 54 and chemotherapy) or insufficiently aggressive, leading to high recurrence rates²⁻⁴.

Esophageal cancer is the sixth most common cause of cancer-related death worldwide and the current median survival time is still <1 year⁵. Incidence rates for esophageal adenocarcinoma (EAC) have risen sharply and it is now the predominant subtype in developed countries. Prognosis is highly variable for EAC patients as shown by the wide range of 5-year survival (18-47% with lymph node involvement), making it difficult to advise patients when embarking on a long course of grueling treatment^{2,6}.

Theoretical and experimental studies attempt to understand how tumor cell populations respond to selective pressures over time⁷. A number of models of tumor evolution have been proposed, including linear, branching, neutral and punctuated evolution, but the extent to which these are specific to a given cancer type or co-occur is controversial^{8,9}. Genome sequencing studies have attempted to delineate different models of evolution¹⁰. However, many of these studies have focused solely on evolution within the primary site, and knowledge of how genetic diversity emerges during metastasis remains limited. The

68 lack of understanding is in part due to the practical challenge of collecting multiple samples69 over space and time from advanced stage cancer patients.

To better understand the evolution of EAC, we designed a prospective study with extensive sampling over time including samples from diagnosis, surgery and at warm autopsy (Figure 1). We used whole genome sequencing (WGS) at high depth (50x), to identify mutations, and at shallow (1x) coverage, to track known variants, to interrogate the clonal architecture across time and space.

75

76 Results

77 Genomic architecture of 18 cases

78 Eighteen cases were included in the study and the clinical demographics of these cases are 79 shown in Supplementary Table 1 and 2, with details of the individual samples given in 80 Supplementary Table 3 and 4. In the first part of the study (Figure 1a, Extended Data Fig. 81 1,2) we used 50x WGS to construct a phylogenetic tree for each case, to understand the 82 relationship between the primary and metastatic tumors (Figure 2, Extended Data Fig. 3, 83 Supplementary Figure 1, Supplementary Table 3, 4). Mutation clustering was performed, 84 and the fractions of tumor cells carrying each set of mutations (Cancer Cell Fraction, CCF) 85 within each sample were used to determine: 1) the clonal and sub-clonal architecture of 86 each tumor (subclonal CCF <95%, clonal CCF \geq 95%); 2) the hierarchy of events; and 3) the 87 distance of these sub-clonal or clonal clusters from the most recent common ancestor 88 (MRCA) (Figure 1a, Extended Data Fig. 1,2). The CCF and number of single nucleotide 89 variants (SNVs) associated with each clone and subclone are shown in Supplementary Table 5 and 6, as is the tumor purity of each sample using the Battenberg algorithm¹¹, in 90 91 Supplementary Table 7 and the confidence intervals of the clonal and subclonal CCFs in 92 Supplementary Table 8. Detailed information on experimental design is provided in the Life 93 Sciences Reporting Summary.

These analyses enabled us to construct phylogenetic trees (Methods). In all cases we observed a long trunk compared to the rest of the tree (median 19,034 SNVs, IQR 11,299-63,908), consistent with previous studies in EAC^{12,13}. The median size of clonal or subclonal clusters across all cases was 3,069 SNVs (IQR 1332-63908) and only 2/157 contained fewer than 200 SNVs (S1_3 and P5_11), Extended Data Fig. 3 and Supplementary Table 6.

The key driver events^{14,15} are depicted on each phylogenetic tree (Figure 2 and Extended 99 100 Data Fig. 3). The events identified as most frequent in previous studies occurred in the 101 trunks of the phylogenetic trees, consistent with their previous classification as drivers. TP53 102 was mutated in the trunk of 16 out of 18 cases, consistent with our knowledge of the disease^{14,16-19}. Amplifications (gene names in red) were often truncal, but also observed on 103 104 the branches of the phylogenetic tree, providing evidence of divergence during later 105 evolutionary stages (Figure 2, Extended Data Fig. 3). The majority of events in driver genes 106 were copy number alterations (CNAs) rather than SNVs or InDels (Figure 2, Extended Data Fig. 3) 14,19,20 . There was no significant difference in the overall number of structural variants 107 108 between primary and metastatic samples (p=0.41, generalized linear model; Extended Data 109 Fig. 4b). However, a larger proportion of structural variants in metastatic samples were 110 retro-transpositions of mobile elements than in the primary samples (p=0.045, Extended 111 Data Fig. 4c). This contrasts with pancreatic cancer, where deletions and fold-back 112 inversions are more common in metastases, and breast cancer where tandem duplications dominate²¹. Interestingly, the high rate of L1 transposon activity in EAC has recently been 113 associated with high activity in the germline²². Our results suggest a further increase in L1 114 115 activity in metastatic EAC. Furthermore, the proportion of structural variants found uniquely 116 in metastases or in primary sites was higher than that of SNVs (Figure 2, Extended Data 4a), 117 suggesting an increase in genomic instability in later stages of the disease. However, it 118 cannot be ruled out that some structural variants have not been identified in every sample 119 as a result of lower sensitivity in the detection of structural variants than SNVs.

Across the eighteen cases, 8 mutational signatures were observed, consistent with previous studies²³⁻²⁶ (Figure 3a), with varying prevalence across the cases. None of the signatures that we observe in patients in our cohort who had oncologic therapy have been associated with treatment with alkylating antineoplastic agents²⁷, platinum therapy²⁸ or radiation therapy²⁹.

124

125 Early seeding of oligometastases

Ten of eighteen patients (S3, S4, P1-4, P6, P8-10) had both nodal and solid organ
metastases, allowing a direct comparison of the genomic architecture between different
metastatic sites (Figure 2).

In four of these ten cases, an isolated clone or subclone confined to 1 or 2 distantmetastases, i.e. an oligometastasis, depicted as a dashed black node on the first branch of

131 the phylogenetic tree, shared the highest congruence to the MRCA, (P1, P4, P10, S3 in 132 Figure 2; Subclones P1_2, P4_3, P10_2, S3_2 in Supplementary Table 5). In P1, this clone 133 (P1 2) was observed only in the primary tumor and a pleural metastasis. In S3 and P4, the 134 clone involved in this isolated seeding was identified at a single distant site and not in the 135 primary tumor (S3_2: liver metastasis (D1), P4_3: para-aortic lymph node (L3)). In P10, the 136 early seeding clone (P10 2) was shared between a distant para-aortic node and a sub-clonal 137 metastasis in the right hemi-diaphragm. The subclones associated with these isolated 138 seeding events showed little divergence from the MRCA across these 4 cases (median 1,913 139 SNVs, range 832-8,591), suggesting early seeding to distant metastases. Notably, in P9 a 140 subclone (P9 10, Supplementary Table 5) was found in a premalignant area of Barrett's 141 esophagus and a pleural metastasis but not in any of four areas of the primary tumor 142 subject to 50x WGS. This subclone lineage shares no variants with the main lineage and 143 appears to be an independent second cancer (Figure 2).

144 A single clone gives rise to multiple metastatic sites

145 A striking observation was that 9/10 cases had a clone (outlined in red on the phylogenetic 146 tree in Figure 2) that was followed by dispersion of multiple subclones from the primary to 147 discrete metastatic sites, resulting in a model of metastasis that we term 'clonal diaspora'. 148 In most cases, this dispersion was visually stellate in nature, this being defined as a feature 149 of a phylogenetic tree involving 3 or more branches leading from a single founder clone (see 150 details in Discussion). The subclones forming diasporas were located in both primary and 151 metastatic tissue in eight cases (P1, P2, P3, S4, P4, P6, P8, P10) and in P9 were unique to 152 metastases (Figure 2). The only two cases lacking a stellate pattern on the phylogenetic tree 153 were P10 and S3. The latter is a non-autopsy case with limited tissue sampling and the early 154 distant seeding in this case is consistent with a pattern of parallel evolution (Figure 2).

155

156 Subclonal spread is not constrained by location or tissue

157 In the second step of the study we tracked the spread of metastases across a wider range of 158 lymph node and distant tissue sites by performing 1x WGS in a further 248 tissue samples 159 from 6 autopsy cases (Figure 1a,c). We did not call new mutations, as this would not be 160 possible at 1x sequencing, but used this method to detect the spread of clones and 161 subclones previously identified using 50x WGS (bioinformatic validation of methods in 162 Extended Data Fig. 5 and 6, Supplementary Note; wet lab validation in Extended Data Fig. 7, 163 Supplementary Table 9). The samples used in this part of the study are outlined in 164 Supplementary Table 10. The median size of subclonal and clonal clusters (identified 165 previously at 50x WGS) that we aimed to detect using 1x WGS was 3,784 (IQR 1,966-49,955). 166 Sample sites were grouped according to their similarity based on the presence of subclones 167 and clones previously detected with 50x WGS (Supplementary Note). The resulting groups 168 of samples are color coded and numbered, and each sample site, colored by group, is shown 169 on the adjacent body map (Figure 4, see also Supplementary Note). Notably, the samples 170 that grouped together based on shared clonal origins were widely dispersed anatomically.

Four out of six cases with extensive spatial sampling (Figure 4) had liver metastases evaluated and three of these contained samples that were more similar to local lymph node metastases than neighboring liver metastases (P4, P6, P8 but not P10). The high number of groups within the liver (up to four) suggested seeding by multiple subclones (seen in P4, P6, P8), whereas the single group in the liver of P10 (orange, group 3) indicated seeding by a common progenitor or a set of closely related cells.

177 A comparison of lymph node location and genomic contiguity showed no evidence of 178 tropism, i.e. genomically similar lymph nodes did not occupy nearby anatomical locations. 179 Lymph nodes above and below the diaphragm were frequently seeded from common 180 events (P2: groups 1, 3; P4: groups 5, 6; P6: group 5; P8: groups 2, 3,5, 6; P10: group 4), at 181 odds with a progression from local to distant nodes. Similarly, a comparison of lymph node 182 and solid organ metastases showed scant evidence for tropism, with the exception of P1 183 (Supplementary Note). This patient underwent surgical resection and subsequently had 184 metastatic disease recurrence. In this cancer, separate subclones seeded lymph node and 185 pleural metastases (Figure 2, 4). Notably, the distant metastasis (D1) was an early branching 186 oligometastasis whereas the lymph nodes (L1, L2) constituted the later diaspora event 187 (black and red circles, respectively, in Figure 2).

188 We further traced regions of the primary tumor at autopsy that had similar subclonal 189 compositions to each of the metastases, shown as adjacent tumor maps (Figure 4, bottom 190 left of each case). Subclones occupied spatially distinct areas in the primary tumor.

We also looked for driver amplifications post MRCA or post diaspora on a per case basis and identified selection in 6/10 cases. However, this is likely to be an under-estimate, since there may be non-copy number drivers present in additional cases. The ratio of non-synonymous

to synonymous SNVs (dN/dS) was analyzed across all cases in order to assess the presence
 or absence of positive selection³⁰. Results indicated positive selection in both clonal and
 subclonal genomes, albeit with lower levels of selection within subclones (Extended Data
 Fig. 8).

198

199 Metastatic spread is rapid in EAC

To examine the timing and speed of metastatic spread we analyzed base substitution mutational signatures, particularly the aging signature which features a predominance of C>T transition in the NpCpG trinucleotide context (Figure 1a, Figure 3).

203 Signature 1 arises from the spontaneous or enzymatic deamination of methylated cytosines, 204 which is an endogenous process that occurs continuously in both healthy and cancerous cells. This has been shown to act as a molecular clock^{27,31-35}, and was therefore used here as 205 206 a method to examine the temporal relationship between metastases. Using a previously described method for deconvolving mutational signatures³⁵, we observed that signature 1 207 208 was present in the trunk but absent in all subclones that constituted diaspora (following the 209 red parental clone in Figure 2) for P2, P4, P6, P9, P10, S4 and it was significantly reduced for 210 P1 (21% to 3%) and P3 (16% to 9%) (Wilcoxon signed rank test p=0.039, Figure 3c). To 211 account for the possibility that the number of signature 1 mutations in branch subclones 212 was below the resolution of our deconvolution methods, we also identified the number of 213 mutations with the characteristic feature of signature 1, i.e. C>T mutations in a CpG context. 214 To estimate the time of appearance of diaspora, we compared the number of these 215 characteristic mutations that occurred along the trunk to the parental red clone marking the 216 onset of diaspora with those that occurred on the longest branch leading from this point. 217 The median proportion of such mutations occurring prior to the onset of diaspora was 0.911 218 (Figure 3b). Thus, in the majority of cases one might deduce that little time has elapsed 219 between the appearance of the cell that is ancestral to disseminating cells and the individual 220 cells that seeded each of the metastases. With the exception of P8, the proportion of 221 mutations attributed to signature 1 was significantly lower after the parental (red) clone on the phylogenetic tree ($p < 9.1 \times 10^{-5}$, Chi-squared test across all cases; Figure 3c) suggesting 222 223 an increase in the activity of other processes in later evolutionary stages (Supplementary 224 Table 11). Of note, there was an increase in the proportion of signature 3 in subclonal SNVs

- compared to clonal SNVs (Wilcoxon signed rank test p=0.019, Figure 3b), suggesting failure
- 226 of DNA double strand break repair is predominantly a late-stage event in EAC.
- 227

228 Early detection from diagnostic samples

229 Next, we investigated eight cases (P1-4, P6, P8-10) for which the esophageal diagnostic FFPE 230 biopsy or surgical sample (primary tumor at resection for P1 and lymph node from surgery 231 for P9) were available, with a median time prior to autopsy of 12 months (range 5-30 months) (Figure 1). The diagnostic sample for P1 was snap frozen and sequenced to 50x 232 233 (Figure 2; highlighted with * in Extended Data Fig. 9), while 1x WGS was performed on the 234 remainder of the cases. Between 8% and 36% of the subclones and clones observed in 235 samples taken from autopsy were also present in the diagnostic samples (Supplementary 236 Note and Extended Data Fig. 9). In six cases, all subclones identified from the biopsy samples 237 were also found in the primary samples from autopsy. Two diagnostic endoscopic samples 238 from P4 also contained many of the mutations found in the lymph node L2 at autopsy, 239 which had not been previously identified in the primary tumor at autopsy (Figure 2, 240 subclone P4_17, Supplementary Table 5). Similarly, the biopsy sample from P10 contained a 241 substantial number of mutations from both the oligometastasis that seeded D2 and L4 242 (Supplementary Table 5, P10_2), and the lineage that later metastasized to multiple sites 243 (Figure 2). Notably, P4 and P10 had shorter survival times after diagnosis than the remaining 244 patients (5 and 4 months, respectively).

245

246 Plasma sample analysis at autopsy and earlier time-points

247 We assessed the clonal composition of circulating tumor DNA (ctDNA) at earlier time-points 248 in seven blood samples from five cases (Figure 1, Figure 5a,c; Extended Data Fig. 10, 249 Supplementary Table 12). Combined 1x WGS subclone/clone detection, copy number 250 aberrations and TP53 fraction using digital PCR data are displayed for two of these cases (P6 251 and P10) in Figure 5a. Notably, P6 was a patient being treated with curative intent and had 252 no radiological evidence of distant nodal or organ metastases at the time of clinical staging. 253 However, at the time of diagnosis mutations from the truncal cluster and three subclonal 254 clusters later found in the metastases were already present in the plasma (Figure 5a) along 255 with amplifications in MYC and GATA4. Case S4 is noteworthy as the brain metastases (D1, 256 D2 in Figure 2) appeared to have originated from a subclone shared between the primary and a local lymph node, both of which were removed at the time of surgery (Extended Data
Fig. 10c). However, mutations from the truncal cluster and four subclonal clusters were
already present in ctDNA prior to radiological recurrence.

260 In eight cases, plasma was available from rapid autopsy. One case (P3) failed wet lab SNV 261 validation and was hence removed from the SNV subclone analysis (Supplementary Note). 262 Analysis of ctDNA demonstrated that in all cases the truncal cluster from autopsy was also 263 represented in plasma (Figure 5c). In addition, mutations from between 0 and 7 subclonal 264 clusters were identified from plasma (Figure 5c). The ratio of mutations detected from each 265 subclone was very consistent between blood from earlier time points and autopsy (Pearson r range [0.851, 0.994], maximum P-value 8.9×10^{-4}) and in 2 of 5 cases the proportion of 266 267 mutations detected was higher in the earlier sample, suggesting an opportunity for earlier 268 detection of heterogeneous cancer cell populations. Further, subclonal proportions 269 estimated from exome sequencing of plasma samples were highly correlated with those 270 from 1x WGS (Supplementary Table 9).

271 The majority of driver CNAs identified in the MRCA of each tumor from 50x WGS of tissue 272 samples were also identified in plasma both at autopsy and at earlier time-points (Figure 273 5a,b). In addition, MET amplification, which was not present in the MRCA in P1 (Figure 2), 274 was identified in plasma both at autopsy and an earlier time point (Extended Data Fig. 10a), 275 suggesting opportunities for early detection of metastatic subclones. Notably, however, 276 amplifications found only in oligometastases or in post-diaspora subclones from 50x 277 sequencing were not identified in plasma, despite many of them being detected in 1x 278 sequencing of tissue samples (Figure 5b). A plausible explanation for this observation is that 279 each of the many metastasizing subclones contributed insufficient material to the sum of 280 detected ctDNA to enable confident detection of CNAs.

281

282 **Discussion**

We have gathered multiple lines of evidence which suggest that, for the majority of EACs, a complex mode of spread is operative. These lines of evidence can be summarized as follows (Figure 6). We observe multiple subclones, each seeding multiple metastatic sites. These subclones are frequently derived from a single parental clone, generally resulting in a stellate pattern on the phylogenetic tree. Metastases in solid organs can bypass nodal involvement and samples within solid organ sites frequently resemble distant metastases

289 more closely than neighboring metastases within the same organ, i.e. no tropism is 290 observed. All metastases appear to have spread directly from the primary site, with little or 291 no evidence of metastasis-to-metastasis seeding.

These features differ in some important respects from previously described models of metastasis and we propose that they may constitute a distinct, additional model of evolution. We suggest that this pattern be referred to as a 'diaspora', by extension of the anthropological term to cancer³⁶. Within this context, it is associated with the observation that multiple cell populations in metastatic sites are directly linked to the primary site of origin and that individual subclones seed multiple tissue types, analogous to a diaspora crossing multiple national boundaries.

A number of features were frequently associated with this phenomenon (Figure 6), with nine of the cases (all except S3) displaying at least two of the four following features: i) stellate pattern on the phylogenetic tree defined as three or more subclones emerging from the founder clones; ii) lack of signature 1 mutations post MRCA or post-diaspora; iii) spread of subclones to multiple organs of different type; iv) evidence for selection in post diaspora genotypes.

305 Until recently the genomic architectures of metastatic samples have not been defined with 306 enough resolution to discern temporal or spatial patterns of metastatic spread. Several 307 distinct patterns are now emerging which are not necessarily mutually exclusive or cancer-308 type specific. In pancreatic cancer, Yachida et al. demonstrated that distant organ seeding was a late event consistent with a linear progression model²⁴. In prostate cancer, linear 309 progression is often succeeded by multiple waves of seeding³⁷. The same study further 310 311 demonstrated widespread subclonal evolution in metastases and metastasis-to-metastasis 312 spread, in keeping with the relatively long longevity of prostate cancer. Strikingly, a stellate 313 pattern was not observed in any of the cases in that study, despite using a similar design to 314 that used here.

In Supplementary Table 13 we compare the features of our proposed Diaspora model to the previously posited linear³⁸ and parallel⁸ models. Whereas the linear model predicts that a single subclone seeding lymph node sites is followed by transmission to distant organs, the diaspora model posits simultaneous seeding of multiple sites directly from the primary. Unlike the parallel model, the diaspora model implies that metastasis formation occurs after the majority of evolution has occurred in the primary tumor, resulting in multiple subclones

321 found in common between primary and metastatic tumors. Lymphatic and distant 322 metastases in colon cancer have been shown to arise from independent subclones in the primary tumor with disparate evolutionary trajectories³⁹. In contrast, in EAC we find that 323 324 individual subclones frequently seed both lymph node and distant organs suggesting that 325 disparate trajectories for nodal and solid organ metastases do not exist for this disease 326 (Figure 2, 3). Of note we acknowledge that, despite the extensive and systematic sampling 327 across all autopsy cases, further sampling may add further branches to our phylogenetic 328 tree, although this is unlikely to affect the diaspora event itself.

In common with the Big Bang Model proposed for colorectal cancer⁴⁰, our model predicts 329 330 the occurrence of highly branching phylogenies. However, the Big Bang Model proposes 331 neutral dynamics, whereas we observe strong evidence for selection in subclonal 332 populations in the form of dN/dS ratios and the occurrence of subclonal driver 333 amplifications (Figure 2, Extended Data Figure 8, Supplementary Figure 2). Moreover, the 334 clonal maps of the primary tumor demonstrate subclones that occupy spatially discrete 335 areas of the primary tumor (Figure 4), in contrast to the intermixed subclones predicted by 336 the Big Bang Model⁴⁰.

337 The sequence of events in metastatic progression may have clinical implications that require 338 further study (Supplementary Table 13). Clonal architecture in EAC defies anatomical 339 location of lymph node stations and distant sites, which is the current basis for the TNM 340 staging and determines whether curative therapy is appropriate. It has been suggested that 341 the high recurrence rate, 52% within one year, results from seeding of distant metastases that are not detected at the time of diagnosis²⁶. This study provides molecular evidence for 342 343 this observation and highlights the need for different systemic approaches to disease 344 management, including consideration of more aggressive adjuvant therapy which is not currently the mainstay of treatment⁴¹⁻⁴⁴. With advances in the sensitivity of ctDNA assays, 345 346 metastatic subclones may be detectable in the blood, helping to determine when systemic therapy is required post-surgery and in detecting heterogeneity of acquired resistance⁴⁵. 347 Copy number variation in plasma may also be a future early detection strategy⁴⁶. 348

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The occurrence of metastasis is a pivotal event in the life history of a cancer. Understanding the drivers behind such an event would have potential relevance to patient stratification and predicting and preventing metastatic spread⁴⁷. While we have identified many drivers 353 on the trunks of the trees, prior to diaspora (Figure 2), we cannot be certain which event, if 354 any, was the immediate trigger of diaspora in individual cases. In a number of cases, 355 diaspora was coincident with an increase in the proportion of signature 3 mutations, 356 associated with failure of DNA double-strand break-repair by homologous recombination 357 (Figure 3b). Our findings are in keeping with the failure of DNA repair driving the 358 appearance of genomic heterogeneity. Whether the heterogeneity observed is itself the 359 driver of diaspora or merely a symptom is an important area for future study. Our 360 investigations of the potential drivers of diaspora were limited to genomic factors, and 361 further multi-platform studies looking at epigenetic and transcriptomic factors are other 362 important avenues of future research. We anticipate that analyses of single cells or small 363 clusters from primary sites, disseminated tumor cells and circulating tumor cells will also 364 yield finer resolution of the processes of dissemination and metastasis.

In cancer there are currently very few in-depth studies examining the spatial and temporal
 evolution of metastases⁴⁸. Further studies are required to ascertain the extent to which our
 diaspora theory pertains to other cancers.

368

369 Acknowledgements

370 Above all, we are indebted to the patients who donated tissue samples to this project and 371 thank them and their families who supported them through it. We would also like to thank 372 the following individuals for their help with study set-up, patient liaison and tissue 373 collection, Ben Smith, Nyrai Chinyama, Vijay Sujendran, Peter Safranek, Athanosios Xanthos, 374 Tara Nuckcheddy-Grant, Rachel de la Rue, Sebastian Zeki, Rachael Fels Elliott, Peter Collins, 375 Kitty Puttock, Sophie Rabey and staff at Arthur Rank Hospice and Luke A Wylie for scientific 376 discussion and contribution. We would like to thank the Oesophageal Cancer Clinical and 377 Molecular Stratification (OCCAMS) Consortium for providing the vehicle through which 378 funding for the International Cancer Genome Consortium (ICGC) was obtained. We are 379 grateful to Professor Simon Tavaré, FRS for his guidance and support for the esophageal 380 whole genome sequencing project as a part of the International Cancer Genome Consortium 381 (ICGC). We would like to thank Jo Westmoreland, LMB visual aids for her graphic art 382 expertise. Thanks also go to the Cancer Research UK Cambridge Institute Genomics Core for 383 their technical expertise. We thank the Human Research Tissue Bank, which is supported by 384 the National Institute for Health Research (NIHR) Cambridge Biomedical Research Centre,

from Addenbrooke's Hospital. Additional infrastructure support was provided from the CRUK funded Experimental Cancer Medicine Centre in Cambridge. Computation by DCW used the Oxford Biomedical Research Computing (BMRC) facility, a joint development between the Wellcome Centre for Human Genetics and the Big Data Institute supported by Health Data Research UK and the NIHR Oxford Biomedical Research Centre.

390 Ayesha Noorani was funded through an MRC Clinical Research Fellowship. The work was 391 funded through the above and an MRC core grant (RG84369) and an NIHR Research 392 Professorship (RG67258) to Rebecca Fitzgerald. Funding for sample sequencing (50x WGS) 393 was through the International Cancer Genome Consortium and was funded by a programme 394 grant from Cancer Research UK (RG66287). All OCCAMS samples which were part of the 395 surgical/endoscopy cohort were obtained from Cambridge patients. David Wedge is funded 396 by the Li Ka Shing foundation and the National Institute for Health Research (NIHR) Oxford 397 Biomedical Research Centre.

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399 Author Contributions

400 AN designed and implemented the rapid autopsy study, collected the samples, performed 401 the experiments, analyzed data and wrote the manuscript. MG and S.D.P contributed 402 expertise in pathology and sample collection for the rapid autopsy study. ID-B and NG 403 assisted in study implementation, and along with JC, assisted with sample collection at 404 autopsy. M.S performed the structural variant analysis. M.D.E performed genomic data 405 generation and QC. LB conducted data management. XL, PL-S and JW were involved with 406 autopsy sample collection, advice on experiments and data analysis, and XL contributed to 407 experiments, paper writing, and figure design. LA and IM assisted with data analysis. NG 408 assisted with study Implementation. SMac coordinated the sequencing of samples from the 409 OCCAMS project and contributed to paper writing. SM and AM provided pathology 410 data. TT, SG, LP and DG assisted in implementation and ethical conduct of the autopsy 411 study. R.H.H and AH were involved in surgical sample collection and providing surgical 412 expertise. M.R.S contributed to critical evaluation of the study data and manuscript. D.C.W 413 was responsible for data analysis, paper writing, and assuring integrity of data. The OCCAMS 414 consortium was the vehicle through which the infrastructure and funding was obtained to 415 support the study and the consortium contributed to discussions on the ICGC data and the

- 416 clinical ramifications. R.C.F provided grant funding and was responsible for study design,
- 417 supervision of the project, writing the paper and assuring integrity of the data.

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419 The authors declare no competing interests.

420	References			
421				
422	1.	Sporn, M.B. The war on cancer. <i>Lancet</i> 347 , 1377-81 (1996).		
423	2.	Waterman, T.A. <i>et al.</i> The prognostic importance of immunohistochemically detected		
424		node metastases in resected esophageal adenocarcinoma. Ann Thorac Surg 78, 1161-		
425		9; discussion 1161-9 (2004).		
426	3.	Matsuda, S., Takeuchi, H., Kawakubo, H. & Kitagawa, Y. Three-field lymph node		
427		dissection in esophageal cancer surgery. J Thorac Dis 9, S731-S740 (2017).		
428 429	4.	Lou, F. <i>et al.</i> Esophageal cancer recurrence patterns and implications for surveillance. <i>J Thorac Oncol</i> 8 , 1558-62 (2013).		
430	5.	Smyth, E.C. et al. Oesophageal cancer. Nat Rev Dis Primers 3, 17048 (2017).		
431	6.	Cunningham, D. et al. Capecitabine and oxaliplatin for advanced esophagogastric		
432		cancer. N Engl J Med 358 , 36-46 (2008).		
433	7.	Greaves, M. & Maley, C.C. Clonal evolution in cancer. <i>Nature</i> 481 , 306-13 (2012).		
434	8.	Klein, C.A. Parallel progression of primary tumours and metastases. <i>Nat Rev Cancer</i>		
435		9 , 302-12 (2009).		
436	9.	Davis, A., Gao, R. & Navin, N. Tumor evolution: Linear, branching, neutral or		
437		punctuated? <i>Biochim Biophys Acta</i> 1867 , 151-161 (2017).		
438	10.	Yates, L.R. & Campbell, P.J. Evolution of the cancer genome. <i>Nat Rev Genet</i> 13 , 795-		
439		806 (2012).		
440	11.	Nik-Zainal, S. <i>et al.</i> The life history of 21 breast cancers. <i>Cell</i> 149 , 994-1007 (2012).		
441	12.	Murugaesu, N. <i>et al.</i> Tracking the genomic evolution of esophageal adenocarcinoma		
442		through neoadiuvant chemotherapy. <i>Cancer Discov</i> 5 , 821-831 (2015).		
443	13.	Gerstung, M. <i>et al.</i> The evolutionary history of 2.658 cancers. <i>bioRxiv</i> (2017).		
444	14.	Secrier, M. <i>et al.</i> Mutational signatures in esophageal adenocarcinoma define		
445		etiologically distinct subgroups with therapeutic relevance. <i>Nat Genet</i> 48 , 1131-41		
446		(2016).		
447	15.	Dulak, A.M. <i>et al.</i> Gastrointestinal adenocarcinomas of the esophagus, stomach, and		
448		colon exhibit distinct patterns of genome instability and oncogenesis. <i>Cancer Res</i> 72.		
449		4383-93 (2012).		
450	16.	Weaver, J.M. <i>et al.</i> Ordering of mutations in preinvasive disease stages of esophageal		
451		carcinogenesis. <i>Nat Genet</i> 46 , 837-43 (2014).		
452	17.	Ross-Innes, C.S. et al. Whole-genome sequencing provides new insights into the		
453		clonal architecture of Barrett's esophagus and esophageal adenocarcinoma. <i>Nat Genet</i>		
454		47 , 1038-46 (2015).		
455	18.	Dulak, A.M. et al. Exome and whole-genome sequencing of esophageal		
456		adenocarcinoma identifies recurrent driver events and mutational complexity. Nat		
457		<i>Genet</i> 45 , 478-86 (2013).		
458	19.	Nones, K. et al. Genomic catastrophes frequently arise in esophageal adenocarcinoma		
459		and drive tumorigenesis. Nat Commun 5, 5224 (2014).		
460	20.	Frankell, A.M. <i>et al.</i> The landscape of selection in 551 Esophageal Adenocarcinomas		
461		defines genomic biomarkers for the clinic. <i>bioRxiv</i> (2018).		
462	21.	Yates, L.R. et al. Subclonal diversification of primary breast cancer revealed by		
463		multiregion sequencing. Nat Med 21, 751-9 (2015).		
464	22.	Rodriguez-Martin, B. <i>et al.</i> Pan-cancer analysis of whole genomes reveals driver		
465		rearrangements promoted by LINE-1 retrotransposition in human tumours. <i>bioRxiv</i> .		
466		179705 (2018).		
467	23.	Ajani, J.A. et al. Esophageal and esophagogastric junction cancers, version 1.2015. J		
468		Natl Compr Canc Netw 13, 194-227 (2015).		

469	24.	Yachida, S. et al. Distant metastasis occurs late during the genetic evolution of
470		pancreatic cancer. Nature 467, 1114-7 (2010).
471	25.	Sottoriva, A. et al. Intratumor heterogeneity in human glioblastoma reflects cancer
472		evolutionary dynamics. Proc Natl Acad Sci U S A 110, 4009-14 (2013).
473	26.	Mariette, C. et al. Pattern of recurrence following complete resection of esophageal
474		carcinoma and factors predictive of recurrent disease. Cancer 97, 1616-23 (2003).
475	27.	Alexandrov, L.B. et al. Signatures of mutational processes in human cancer. Nature
476		500 , 415-21 (2013).
477	28.	Liu, D. et al. Mutational patterns in chemotherapy resistant muscle-invasive bladder
478		cancer. Nat Commun 8, 2193 (2017).
479	29.	Behjati, S. et al. Mutational signatures of ionizing radiation in second malignancies.
480		Nat Commun 7, 12605 (2016).
481	30.	Dentro, S.C. et al. Portraits of genetic intra-tumour heterogeneity and subclonal
482		selection across cancer types. <i>bioRxiv</i> (2018).
483	31.	Lodato, M.A. et al. Aging and neurodegeneration are associated with increased
484		mutations in single human neurons. Science 359, 555-559 (2018).
485	32.	Gao, Z., Wyman, M.J., Sella, G. & Przeworski, M. Interpreting the Dependence of
486		Mutation Rates on Age and Time. PLoS Biol 14, e1002355 (2016).
487	33.	Letouze, E. et al. Mutational signatures reveal the dynamic interplay of risk factors
488		and cellular processes during liver tumorigenesis. Nat Commun 8, 1315 (2017).
489	34.	Blokzijl, F. et al. Tissue-specific mutation accumulation in human adult stem cells
490		during life. Nature 538, 260-264 (2016).
491	35.	Alexandrov, L.B. et al. Clock-like mutational processes in human somatic cells. Nat
492		<i>Genet</i> 47 , 1402-7 (2015).
493	36.	Pienta, K.J., Robertson, B.A., Coffey, D.S. & Taichman, R.S. The cancer diaspora:
494		Metastasis beyond the seed and soil hypothesis. Clin Cancer Res 19, 5849-55 (2013).
495	37.	Gundem, G. et al. The evolutionary history of lethal metastatic prostate cancer.
496		<i>Nature</i> 520 , 353-357 (2015).
497	38.	Foulds, L. The experimental study of tumor progression: a review. <i>Cancer Res</i> 14,
498		327-39 (1954).
499	39.	Naxerova, K. et al. Origins of lymphatic and distant metastases in human colorectal
500		cancer. Science 357 , 55-60 (2017).
501	40.	Sottoriva, A. et al. A Big Bang model of human colorectal tumor growth. Nat Genet
502		47 , 209-16 (2015).
503	41.	Sjoquist, K.M. <i>et al.</i> Survival after neoadjuvant chemotherapy or chemoradiotherapy
504		for resectable oesophageal carcinoma: an updated meta-analysis. <i>Lancet Oncol</i> 12,
505		681-92 (2011).
506	42.	Gabriel, E. et al. Novel Calculator to Estimate Overall Survival Benefit from
507		Neoadjuvant Chemoradiation in Patients with Esophageal Adenocarcinoma. JAm
508		<i>Coll Surg</i> 224 , 884-894 e1 (2017).
509	43.	Burt, B.M. <i>et al.</i> Utility of Adjuvant Chemotherapy After Neoadjuvant
510		Chemoradiation and Esophagectomy for Esophageal Cancer. Ann Surg 266, 297-304
511		(2017).
512	44.	Pasquali, S. et al. Survival After Neoadjuvant and Adjuvant Treatments Compared to
513		Surgery Alone for Resectable Esophageal Carcinoma: A Network Meta-analysis. Ann
514		Surg 265 , 481-491 (2017).
515	45.	Parikh, A.R. <i>et al.</i> Liquid versus tissue biopsy for detecting acquired resistance and
516	15	tumor heterogeneity in gastrointestinal cancers. <i>Nat Med</i> 25 , 1415-1421 (2019).
517	46.	van koy, N. <i>et al.</i> Shallow whole Genome Sequencing on Circulating Cell-Free
518		DNA Allows Reliable Noninvasive Copy-Number Profiling in Neuroblastoma
519		Patients. Clin Cancer Res 23, 6305-6314 (2017).
		20

- 47. Hu, Z. *et al.* Quantitative evidence for early metastatic seeding in colorectal cancer. *Nat Genet* 51, 1113-1122 (2019).
- 52248.Robinson, D.R. *et al.* Integrative clinical genomics of metastatic cancer. *Nature* 548,523297-303 (2017).

525 Figure Legends

526 Figure 1 Overall project strategy and study design

527 a. Overall Strategy to identify clonal evolution in metastatic EAC. There were three main 528 steps in this study which comprised: Clonal discovery at autopsy (see Supplementary Note 529 High Depth Whole Genome Sequencing (50x WGS), Mutation clustering and phylogenetic 530 tree construction, dN/dS analysis and Mutational Signature Analysis); Spatial tracking at 531 autopsy (see Supplementary Note Shallow Whole Genome Sequencing (1x WGS) and 532 Temporal tracking at earlier time-points (see Supplementary Note Shallow Whole Genome 533 Sequencing (1x) for Subclone identification, Supplementary Table 12 for precise samples for 534 plasma and Extended Data Fig. 9 for FFPE diagnostic samples). Colored circles depict clones 535 and subclones respectively. b. Sampling Strategy at Rapid Autopsy. Areas sampled for the 536 50x WGS part of the study are shown in blue and for 1x WGS are shown in orange. c. Study 537 Design and Sequencing Strategy. The flow chart demonstrates the study design and how this 538 relates to sequencing. Clonal Discovery is in blue and Clonal Tracking in orange. The sample 539 distribution for 50x WGS and 1x WGS are shown. 50x WGS = High depth WGS (50x), 1x WGS 540 = Shallow WGS (1x). n = number of cases, s = number of samples. +248 solid tissue 541 samples, and 8 ctDNA at autopsy. CNA, copy number alteration; SNV, single nucleotide 542 variant; MRCA, most recent common ancestor.

543 Figure 2 Phylogenetic Analysis of ten cases with nodal and distant metastases

544 Patient body maps (S=surgical case, P=rapid autopsy) are shown. Green circles denote 545 lymph node metastases and yellow circles distant metastases. The labels within each circle 546 describe the specific location (see Supplementary Table 3, 4). An organ is shown in color if 547 metastases were sequenced from that site. The adjacent wedged semi-circle depicts the 548 clinical timelines for each patient. Each wedge corresponds to one month; blue wedges 549 indicate the total lifetime of the patient and red wedges periods of therapy. Phylogenetic 550 trees for each patient are shown and methodology is in Supplementary Note and Extended 551 Data Fig. 1a-b; pink = truncal events shared by all samples, purple = branch events shared by 552 more than one sample, yellow = leaves, events unique to a sample. The circle at the end of 553 a trunk, branch or leaf represents a clone or subclone. Each clone or subclone is annotated 554 to show which samples it is present in. E1-E4 = primary esophageal tumor, L1-L4= lymph 555 nodes, D1-8 =distant metastases, B = Barrett's Esophagus. A subclone annotated with E1, L2

556 for example indicates that this subclone is seen only in samples E1 and L2. The CCF of each 557 subclone/clone (barring the MRCA) is in Supplementary Table 5 and 6. The length of the 558 branches of the tree are reflective of the number of SNVs in the subclone/clone. The scales 559 adjacent to each case are relative, given the variable number of SNVs per case. Trees are 560 annotated with potential driver events, black: missense variants, red: amplifications. Gray 561 dots outlined with a black dashed line denote the first subclone/clone to metastasize that 562 would be classified as non-curative based on anatomical location. Red dots mark the 563 stellate pattern on the phylogenetic tree.

564 Figure 3 Mutational Signatures

a. Contributions of mutational signature in 18 cases (n=122) across the cohort. The bar chart
displays samples on a per case basis (X-axis) and depicts the number of SNVs contributing to
each signature (Y-axis). b. Mutational signatures pre-and post- diaspora across all samples
(n=122) in 18 cases.

Mutations were separately assigned to signatures and the proportion of mutations within 569 570 each case assigned to each signature is shown. Dark lines = median, Boxes = 25th and 75th 571 quartiles, whiskers extend to the most extreme point within 1.5× interquartile range of the 572 box edge. Signatures 1 mutations have a significantly lower representation in post-diaspora 573 mutations, while signature 3 mutations have significantly high. c. Mutational signature 574 analysis of ageing signature (signature 1) pre-and post-diaspora in all cases (n=8) with local and distant spread ($p<1.18 \times 10^{-90}$ across all cases) Chi squared test was used to determine 575 576 the p value. Survival is shown in months from the point of diagnosis *=cases which 577 underwent surgery.

578

579 Figure 4 1x WGS and similarity matrix clustering of 248 further tissue samples from six

580 cases

1x WGS was performed at an average depth of 1x to track subclones and clones previously discovered using 50x WGS for further tissue samples (n=248). Pearson correlation similarity matrix clustering was performed on all samples for each case (plotted against each other) with red indicating sample similarity (r=1) and blue indicating dissimilarity (r=-1). Sample sites used in this part of the study are shown in Supplementary Table 9 and the entire organ is highlighted if solid organ sites were sequenced. For example, liver metastases were only seen in P4, P6, P8, P10. Similarly, P2 had lymph nodes only (only

588 colored dots are seen which represent lymph nodes, no solid organs are highlighted). 589 Clustering was performed based on the presence of subclones and clones already 590 detected using 50x WGS and distinct clusters were identified for each case as 591 demonstrated by the adjacent key per case (each group is both colored and numbered). 592 Samples are displayed on the adjoining body maps for which the color coding corresponds 593 to the genomic clustering in the adjacent heatmap. Sites with multiple samples are 594 magnified and the division of samples shown. Maps of the primary tumor with 595 representation of metastatic subclones are shown with each case, with the colors of the 596 subclones being the same as those in the matrix and body map. Areas shaded red in the 597 primary tumor represent subclones that were not detected in the metastatic samples that 598 underwent 1x WGS and were instead confined to areas of the primary tumor.

599

600 Figure 5 Temporal and spatial tracing of metastatic subclones in plasma

601 a. Plasma ctDNA 1x WGS and digital droplet PCR (ddPCR) analysis for TP53 mutant allele 602 fraction (MAF) for P10 and P6. The MAF of TP53 (%) is shown on the Y-axis and days from 603 diagnosis are shown on the X-axis. The shaded areas represent time periods of therapy. 1x 604 WGS at select time-points was performed and the clonal composition of these samples 605 are shown by the presence of colored clusters. The color of each corresponds to the color 606 of the corresponding node on the adjacent 50x phylogenetic tree with the presence of 607 colored clusters which correlate with the 50x tree. Moreover, copy number traces for 608 each time point are shown for select chromosomes. b. The presence or absence of 609 amplifications and deletions in plasma compared to tissue, detected from 1x WGS for 8 610 cases. Tissue refers to all samples collected at autopsy and at earlier time-points. c. 611 Stacked bar charts to demonstrate the presence or absence of clusters across all plasma 612 samples, including truncal and branch clusters using 1x WGS.

613

614 Figure 6 Diaspora model of metastatic spread and associated features

Panel a depicts clonal diaspora with colored circles representing clones and subclones. *= evidence of selection. Panel b explains the five features seen in diaspora (one is defining, and the other are associated with diaspora) and whether these are present (\checkmark) or absent (x) in each case. * \checkmark implies that the feature is present, and that the evidence was from 1x WGS.

620 Methods

621 Statistics

- 622 Unless otherwise stated, statistical analyses were performed using R, version 3.3.3.
- 623 Clustering of mutations was carried out using a previously published Bayesian Dirichlet
- 624 Process method, DPClust (https://github.com/Wedge-Oxford/dpclust), which calculates
- 625 CCFs of each SNV, taking into account tumor purity and copy number aberrations as
- 626 previously described⁴⁹. Analysis of structural variants used generalized linear models,
- 627 implemented with the R package MASS. Grouping of 1x WGS samples was performed with
- 628 the GENE-E package (https://software.broadinstitute.org/GENE-E/download.html).
- 629 Wilcoxon signed rank tests and Chi-squared tests were used as described in the main text.
- 630 Simulations were used to ascertain the robustness of DPClust to violations of the infinite
- 631 sites assumption and its sensitivity to detect small deviations from stellate patterns.
- 632 Simulations were also used to confirm the correlation between the number of mutations
- 633 detected from 1x WGS and CCF determined from 50x WGS, as described in Online Methods.
- 634 dN/dS analysis was performed using the previously published package dndscv⁵⁰
- 635 (<u>https://github.com/im3sanger/dndscv</u>).
- 636

637 **Patient recruitment and Sample collection**

638 EAC patients were recruited from Addenbrooke's Hospital, Cambridge University Hospitals 639 NHS Trust with the explicit aim to study the clonal evolution of metastases as a sub-study 640 within OCCAMS (Oesophageal Clinical And Molecular Stratification). When it was clear that 641 extensive sampling of metastases could not be achieved without multiple invasive 642 procedures, the PHOENIX autopsy study was set up (Phylogenetic of Oesophageal 643 Neoplasia – An Investigation of Clonal Expansion under REC 07/H0305/52, and REC 644 EE/0043) with a prospective study design. Due diligence was undertaken to ensure 645 compliance with ethical regulations at all times. Patients were eligible if they were at least 646 18 years of age and had received a confirmed diagnosis of EAC following central pathology 647 review. Patients were only approached for the PHOENIX study following a palliative 648 diagnosis of metastatic EAC, with the full involvement of the multidisciplinary team. 649 Samples from the PHOENIX autopsy study were obtained within 6 hours of death and all 650 post-mortems were carried out at Papworth Hospital NHS Trust, United Kingdom.

Samples from Cambridge OCCAMS patients were obtained during diagnostic oesophagogastroduodenoscopy (OGD), at endoscopic ultrasound (EUS) and/or from the surgical resection specimen. Where possible, multiple samples were taken from spatially distinct sites of the primary tumor or metastases. In two cases, brain metastases were sampled at a clinically indicated craniotomy. Blood or normal squamous esophageal samples, at least 5cm distant from the tumor, were used as a germline reference.

All tissue samples were snap-frozen in liquid nitrogen immediately after collection and stored at -80°C. Cancer samples were deemed suitable for DNA extraction only after consensus review of an H&E stained frozen section, from the same sample that would be sent for sequencing, by two expert pathologists who confirmed tumor cellularity at \geq 70%.

661 Samples with overall ≥70% cellularity underwent dissection of the whole surface area with 662 a scalpel, whereas marked areas of <70% underwent macrodissection or laser capture 663 micro- dissection aided by methylene blue staining visualized on the PALM-Zeiss 664 microscope (Zeiss, Oberkochen, Germany). An H&E stained slide was obtained before and 665 after extraction to confirm tumor cellularity of the microdissected section.

666 DNA was extracted from frozen tissues using the All PrepDNA/RNA Mini Kit (Qiagen, 667 Hilden, Germany) and from blood samples using the NucleonTM Genomic Extraction kit 668 (Gen-Probe, San Diego, USA) according to the manufacturer's instructions. Some samples 669 were preserved in paraffin blocks after initially being stored in formalin. DNA from these 670 samples was extracted using the QiAmp FFPE Kit (Qiagen). Plasma extraction (for ctDNA) 671 was performed using the QiASymphony platform (Qiagen) as per the manufacturer's 672 instructions. All samples were eluted in 60μ l of AE buffer and quantified using the High 673 Sensitivity Qubit (Thermo Fisher Scientific, MA, USA).

674 We included 388 samples, predominantly from PHOENIX, and some additional samples 675 from surgery and endoscopy (part of esophageal ICGC).

All samples were collected according to a strict SOP with quality control measures as already described. All demographic and clinical data was anonymized and stored on a central study database (OpenClinica and Labkey). The clinical characteristics of the patients are provided in Supplementary Table 1 and 2. In terms of specifics of sample collection at autopsy, the primary tumor was opened down the midline of the esophagus and the greater curve of the stomach to expose the lumen. The tumor was divided in 12 areas with sampling as shown.

682 The size of tumors varied per case, but the division of sampling was always kept identical to 683 preserve reproducibility. In terms of the strategy for genomic sequencing (as per Figure 1), 684 up to 3 lymph nodes were chosen for 50x WGS in the areas shown (cervical, regional and 685 para-aortic) and up to 24 lymph nodes in each case (8 further lymph nodes per cervical, regional and para- aortic areas (as per the Japanese Classification of nodal staging⁵¹) were 686 687 chosen for the 1x WGS part of the study. At least one metastasis per solid organ was chosen 688 for 50x WGS and for the 1x WGS part up to 8 samples were taken per organ for further 689 analysis. In addition, 8 samples from metastatic sites which had previously been sequenced 690 for 50x WGS were further sequenced for 1x WGS to assess the effects of metastatic 691 heterogeneity.

692

693 Whole genome sequencing and data analysis strategy

694 We used the Illumina HiSeq platform to perform WGS on multiple regions collected from 695 each primary tumor, lymph node and/or solid organ metastasis (Figure 1a,b, Supplementary 696 Table 3, 4). All DNA extractions and WGS conformed with ICGC quality control standards and 697 required \geq 70% cellularity and a matched germline sample. WGS was performed at high 698 depth (median coverage 66.3, IQR 56.1-87.2) to discover mutations in 122 samples from 18 699 patients (Supplementary Table 3, 4). In addition, low depth WGS (median coverage 1, IQR 1-700 5) was performed to track these mutations spatially in up to 48 solid tissue samples per 701 case, (total=248) and 8 ctDNA samples at autopsy. Temporal tracking was performed in 702 cases with archival biopsy material, and where historical bloods were available 703 (Supplementary Table 12, Figure 5, Extended Data Fig. 6). For each patient the number of 704 subclones and the cancer cell fraction within each subclone was inferred using an extension of a previously described Bayesian Dirichlet process¹¹ and we applied a set of previously 705 706 described rules to derive a phylogenetic tree (Additional Methods⁵²). All sequencing data 707 have been deposited in the European Genome-Phenome Archive under accession number 708 EGAD00001005434. TP53 analysis in cell free tumor DNA (ctDNA) was performed using 709 Digital PCR on the Bio-rad platform (Bio-rad, California) using validated TP53 assays 710 (Supplementary Table 14).

711

712 Mutation clustering and phylogenetic tree construction

713 The workflow used to perform mutation clustering and phylogenetic tree construction is 714 depicted in Extended Data Fig. 1a and illustrated with an example case, S3, in Extended 715 Data Fig. 1b. For each patient, we inferred the number of subclones and the fraction of 716 tumor cells within each subclone by using a previously described Bayesian Dirichlet process (BDP) to cluster mutations according to their mutation copy number⁴⁹. We extended this 717 718 process into n dimensions for patients with n related samples, where the number of 719 mutant reads obtained from multiple related samples were modelled as independent 720 binomial distributions. The BDP uses Markov chain Monte Carlo (MCMC) to sample the CCF 721 values of the subclones in each sample. MCMC is run for 1000 iterations and outputs, for 722 each iteration, the sampled position of each cluster, p_{i_h} and the weight of each cluster, V_{h_r} , 723 which is an estimate of the proportion of mutations assigned to that cluster. The first 200 724 iterations are considered as a 'burn-in' and are not used in subsequent steps. In order to 725 obtain the set of subclones present within a tumor and their CCF values, the following 726 procedure was followed:

727 Using the aforementioned MCMC sampling of CCF values from all n samples, for 728 every possible triplet of samples, obtain posterior density estimates of CCF using the function kde in the R package ks, with input parameters $x = pi_h$, bandwidth = 729 730 0.1, w = V_h . Set gridsize such that density estimates are obtained to a resolution of 731 0.02. Identify local peaks in the posterior mutation density as locations higher 732 than any other gridpoint within a range of 2 gridpoints. For each local peak, define 733 a region representing a 'basin of attraction', defined by a set of planes running 734 through the_point of minimum density between each pair of cluster positions. 735 Assign each mutation to the cluster in whose basin of attraction they are most 736 likely to fall, using CCF values from MCMC sampling.

 Across the set of all possible triplets, identify sets of mutations that are assigned to the same cluster in every triplet. Estimate the CCF of each cluster as the mean CCF of the mutations assigned to that cluster. Estimate the 95% confidence intervals as the [0.025,0.975] quantiles of the mean pi_h values of the mutations assigned to each cluster within MCMC sampling.

Finally, again using the aforementioned MCMC sampling of CCF values from all n samples,
for every pair of samples, plot the mutation density, estimated using the function kde in

the R package ks, with input parameters $x = pi_h$, bandwidth = 0.1, $w = V_h$.

Taking a conservative approach, clusters were identified as subclonal only if the 95%
confidence intervals of the posterior estimate of the proportion of cells excluded the value
1. Clusters containing less than 1% of all mutations identified in a tumor were not included
in phylogenetic reconstruction.

749 Occasionally, copy number states are incorrectly called in small regions of some cancer 750 genomes. As a consequence, mutations falling in these regions have inaccurate estimates 751 of CCF and can cause artefact clusters. Such clusters may be identified after mutation 752 clustering since they contain a small percentage of mutations (less than 2.5%), the 753 mutations within them are located in localized regions of the genome, and, often, they 754 cannot be placed on the phylogenetic tree because they have discordant CCF values. We 755 excluded these clusters from phylogenetic tree construction. The number of clusters 756 excluded in total was seven (5 in P2, 1 in P3, 1 in P10). Two samples had low tumor content 757 (36% in P3_E1, 14% in S5_T1). As a result, CCF estimates for subclones found in these 758 samples are imprecise and led to violations of the sum rule (see below). The CCF values of 759 the relevant clusters were manually corrected to enable them to be placed on the 760 phylogenetic tree, as follows: P3 E1 only cluster adjusted from 1 to 0.85; S5 E1 truncal 761 cluster adjusted from 0.85 to 1.

762 To determine the most likely phylogenetic tree, we applied two rules, previously described⁵². Briefly, the 'sum rule' (which is an extension of the pigeonhole principle 763 764 described in Ref 11), asserts that if a subclone A is ancestral to both subclones B and C and 765 if the summed CCFs of B and C exceed the CCF of A in any sample, the relationship 766 between the subclones must be linear. The 'crossing rule' is applied to tree construction 767 from multiple samples. It asserts that if the CCF of B is higher than the CCF of C in sample X 768 and the CCF of B is lower than the CCF of sample C in sample Y then B and C must be in 769 separate branches of the phylogenetic tree, i.e. they are not collinear. For all clonally 770 related samples, the same underlying phylogenetic tree must exist. This exerts much 771 greater stringency to the inferred ordering of subclonal clusters present in more than one 772 sample and defines their position on the phylogenetic tree unequivocally. Note that P9 773 contains two independent cancers derived from Barrett's esophagus and adenocarcinoma 774 regions. CCF values are reported relative to the dominant cancer, so in P9 D4, which 775 contains both cancers, the two cancers are reported with CCFs of 100% and 69%. This apparent violation of the sum rule results from the mathematical convenience ofnormalizing to the dominant cancer.

It should be noted that the sum rule and crossing rule only strictly apply when the infinite sites assumption (ISA) is obeyed. The ISA states that each mutation only occurs once during the lifetime of a tumor and that mutations never revert to normal. A recent study⁵³ has shown, through analysis of targeted sequencing of single cells , that the ISA is not always followed in real data, for two reasons:

Copy number alterations (CNAs), specifically losses and loss of heterozygosity,
 have the effect of removing mutations in the deleted region, resulting in the
 apparent 'reversion' of a mutation.

• The same mutation may occur on more than one occasion, particularly if the

786

787 mutation is a driver mutation.

788 In our study, we take account of CNAs when calculating the CCF of each mutation. In 789 regions that have undergone gain of one or both alleles, a mutation may be present on 790 more than one chromosome copy, up to the number of copies of the most amplified 791 chromosome copy. Conversely, if one or both chromosome copies have undergone loss in 792 a particular sample, a mutation may be lost in that sample. In the situation where a 793 mutation is unobserved in a sample and that sample has a copy number state lower than 794 that observed in another sample in which the mutation is observed, we do not call the 795 mutation as absent. Rather, we cluster it based on its CCF in the remaining samples, 796 treating its CCF in the target sample as unknown.

797

798 Identification of cancer cell fraction

799 For each mutation we calculated the mutation copy number as previously described, using 800 the mutant allele burden, tumor cellularity and locus specific copy number in the tumor and matched normal⁴⁹. The mutation copy number reflects the percentage of tumor cells 801 802 within a sample carrying that mutation, and permits the cross-comparison of the mutation 803 in related samples despite differences in tumor purity and/ or copy number profiles. 804 Mutations present on multiple copies of a chromosomal segment will have a mutation 805 copy number greater than 1. To group mutations according to the percentage of cells 806 containing it, or cancer cell fraction (CCF), the number of chromosomes carrying the 807 mutation must be determined. For all mutations within amplified regions with a major

allele copy number, the observed fraction of mutated reads was compared to the expected fraction of mutated reads resulting from a mutation present assuming a binomial distribution³⁷.

811

812 Annotation of the trees with mutations

813 We annotated each tree with oncogenic or putative oncogenic alterations including 814 substitutions and copy number changes. For substitutions, cluster assignment information 815 from a multidimensional Dirichlet process was used.

For rearrangements and copy number changes, branch assignment was achieved by considering the set of samples containing the variant and the subclonal fraction of the associated copy number segment where applicable. All potential driver alterations were annotated. For substitutions, structural variants and copy number events, these included a set of genes compiled from the TARGET database from the Broad Institute and multiple sequencing datasets for OAC^{14-16,18,19}.

822

823 Shallow Whole Genome Sequencing for Subclone Identification

824 For shallow whole genome sequencing, samples were sequenced to a median depth of 825 ~1x. It was not therefore feasible to call mutations de novo for these samples, but we were 826 able to count the number of mutations from each subclone that reported a mutant read in 827 1x WGS sequencing. We performed simulations of 1x WGS data in order to ascertain the 828 correlation between the number of mutations identified and the CCF of each subclone. 829 First, we simulated subclones with CCF values between 0.01 and 1.00, assuming 1000 830 mutations per subclone, sequencing depth drawn from a Poisson distribution with 831 expected value 1, and binomial sampling of WT and mutant reads. The correlation 832 between the number of mutations detected and the CCF of the subclone was very high 833 (Pearson r = 0.992, Extended Data Fig. 4). In order to test whether subclones containing 834 fewer mutations also had good correlations between CCF and number of detected 835 mutations, we performed further simulations of subclones containing between 50 and 836 1,000 mutations and ascertained that the correlation remained very high (> 0.997) for 837 cluster sizes as small as 200 (Extended Data Fig. 5). Of the 169 subclones identified in our 838 study, only two contained fewer than 200 mutations, indicating that the number of

839 mutations detected is a good proxy for the CCF of a subclone.

SNVs from libraries sequenced to a minimum of 1x following filtering, were allocated to subclones previously identified at 50x WGS. Mapping quality and base quality of 10 were used. This resulted in tabulated counts for SNVs being allocated to subclones identified at 50x WGS for each sample. Normalization was performed according to the number of SNVs assigned to each subclone from 50x WGS, and to the total number of SNVs in that sample in order to account for potential differences in coverage, using the following equation:

846 $CCF_{cluster} = n_{cluster}/n_{truncal} \times H_{truncal}/H_{cluster}$

in which $n_{cluster}$ and $n_{truncal}$ are the numbers of loci in the target cluster and the truncal cluster that have mutant reads in the target sample and $H_{cluster}$ and $H_{truncal}$ are the number of mutations identified from 50x WGS in the target and truncal clusters. For each 1x WGS sample, this provides an estimate of the CCF of each subclone within that sample.

In all cases, near equal coverage was obtained and in cases of low cellularity further sequencing was performed in order to achieve this. After normalization, the GENE-E package (https://software.broadinstitute.org/GENE-E/download.html) was used to cluster the 1x WGS samples according to the similarity of their CCF profiles using Pearson correlation.

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857 Data Availability

Sequencing data that support the findings of this paper have been deposited in the European Genome-phenome Archive with the accession code EGAD00001005434.

860

861 **Code Availability**

862 All code required to reproduce the analysis outline in this manuscript can be found in the

- 863 main and supplementary methods. There are no restrictions to the accessibility of this code.
- 864

865 **Method-only references**

- 866 49. Bolli, N. *et al.* Heterogeneity of genomic evolution and mutational profiles in multiple
 867 myeloma. *Nat Commun* 5, 2997 (2014).
- Martincorena Inigo, R.K.M., Gerstung Moritz, Dawson Kevin J, Haase Kerstin, Van
 Loo Peter, Davies Helen, Michael R. Stratton Michael R, Campbell Peter J. Universal
 Patterns Of Selection In Cancer And Somatic Tissues. *Cell* (2017).
- 51. Japanese Gastric Cancer, A. Japanese classification of gastric carcinoma: 3rd English
 edition. *Gastric Cancer* 14, 101-12 (2011).

- 52. Jiao, W., Vembu, S., Deshwar, A.G., Stein, L. & Morris, Q. Inferring clonal evolution
 of tumors from single nucleotide somatic mutations. *BMC Bioinformatics* 15, 35
 (2014).
- Kuipers, J., Jahn, K., Raphael, B.J. & Beerenwinkel, N. Single-cell sequencing data
 reveal widespread recurrence and loss of mutational hits in the life histories of tumors. *Genome Res* 27, 1885-1894 (2017).

a OVERALL STRATEGY TO IDENTIFY CLONAL EVOLUTION IN METASTATIC EAC



b SAMPLING STRATEGY

c SEQUENCING STRATEGY





Figure 3

а





Case	Pre-Diaspora	Post-Diaspora	p value	Survival (months)
P1	21% 79%	3% 97%	9.10E-08	20*
P2	16% 84%	100%	9.09E-05	12
P3	16% 84%	9% 91%	3.19E-05	14
P4	5% 95%	100%	1.56E-95	5
P6	6% 94%	100%	3.10E-14	5
P8	100%	6% 5.87E-38 94%		30
P9	8% 92%	100%	5.82E-72	12
P10	21% 79%	100% 6.58E-161 100% 1.11E-14		4
S4	9% 91%			37*

 Percentage of Clocklike
 Signature Percentage Non-Clocklike Signature

С





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P4









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Truncal clusters detected at S-WGS Branch clusters detected at S-WGS Clusters not detected at S-WGS

а

Diaspora Model of Metastatic Spread



b

Features of Diaspora

	DEFINING		ASSOCIATED		
Case	Multiple subclones from primary spread to multiple metastatic sites	Stellate pattern of three or more subclones derived from the same ancestor found in metastatic sites	Lack of Signature 1 mutations, indicating rapid accumulation of mutations and near- synchronous spread	Spread of at least one subclone to organs of different types, including both lymph nodes and distant organs	Evidence for selection of subclones within the diaspora, indicative of an evolutionary niche (driver amplifications)
P1	✓	×	\checkmark	*√	×
P2	✓	√	\checkmark	\checkmark	\checkmark
Р3	✓	×	×	\checkmark	\checkmark
Р4	✓	√	\checkmark	\checkmark	\checkmark
Р6	✓	✓	\checkmark	\checkmark	\checkmark
Р8	✓	√	×	*√	×
Р9	✓	✓	×	\checkmark	×
P10	✓	×	\checkmark	\checkmark	×
S3	×	×	×	×	\checkmark
S4	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark

Overall Methodology



1 Battenberg Algorithm

2 SNV clustering using Dirichlet Process



3 Clustering Results

Cluster No.	SNVs in cluster	seen in (CCF%)
1	5697	All (100%)
2	1913	D1 (96%)
3	1663	E1, E2, L1
4	1382	E1 (81%), E2 (4%), L1 (3%)
5	1332	E1 (21%), E2 (94%)
6	1139	L1 (94%)
7	1096	L1 (39%)
8	322	D1 (40%)
9	302	E2 (36%)
10	300	E1 (28%)
11	237	E1, E2 (95%), L1 (3%)

- 4 Unscaled Tree
- 5 Final Tree with scaled branch lengths and Gene Annotation























P10.05 P10.05 P10.06

P10_D7 P10_D2 P10_D4 P10_D8

Number of SVs

150

0

PIDEZ P10_E3

Pt0_Et

PTOLM PIOLO

P10.64



S1



BND DEL DUP INS INV ME





PS. Ed P5.41 Ps. E. BND DEL DUP INS INV ME

P5.12

PS, E3

0

PSET PS_E2













Snap Frozen at Autopsy

Diagnostic FFPE biopsy







