1 **ESR1** mutant breast cancers show elevated basal cytokeratins and immune

2 activation

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28 Abstract

29 Estrogen receptor alpha (ER/ESR1) is mutated in 30-40% of endocrine resistant ER-30 positive (ER+) breast cancer. ESR1 mutations cause ligand-independent growth and 31 increased metastasis in vivo and in vitro. Despite the distinct clinical features and 32 changes in therapeutic response associated with ESR1 mutations, there are no data 33 about their potential role in intrinsic subtype switching. Applying four luminal and 34 basal gene set pairs, ESR1 mutant cell models and clinical samples showed a 35 significant enrichment of basal subtype markers. Among them, the six basal 36 cytokeratins (BCKs) were the most enriched genes. Induction of BCKs was 37 independent of ER binding and instead associated with chromatin reprogramming 38 centered around a progesterone receptor-orchestrated topological associated 39 domain at the KRT14/16/17 genomic region. Unexpectedly, high BCK expression in 40 ER+ primary breast cancer is associated with good prognosis, and these tumors 41 show enriched activation of a number of immune pathways, a distinctive feature 42 shared with ESR1 mutant tumors. S100A8 and S100A9 were among the most highly 43 induced immune mediators shared between high-BCKs ER+ and ESR1 mutant 44 tumors, and single-cell RNA-seq analysis inferred their involvement in paracrine 45 crosstalk between epithelial and stromal cells. Collectively, these observations 46 demonstrate that ESR1 mutant tumors gain basal features with induction of basal 47 cytokeratins via epigenetic mechanisms in rare subpopulation of cells. This is 48 associated with increased immune activation, encouraging additional studies of 49 immune therapeutic vulnerabilities in ESR1 mutant tumors.

50 Introduction

Breast cancer is characterized by a high degree of heterogeneity, originally identified through the use of immunohistochemistry and gene expression profiling^{1,2}. Broadly, molecular subtypes can be grouped into luminal (luminal A and luminal B), HER2-enriched and basal-like tumors, primarily driven by expression of ER, PR and HER2 and Ki67³. Tumors with different molecular subtypes show distinguishing clinical features and therapeutic responses^{4,5}, including metastatic spread and immune profiles^{6,7}.

58 The basal-like subtype, which represents 15-25% of all cases and overlaps with triple negative breast cancers (TNBC), is characterized by a unique gene 59 expression profile similar to that of myoepithelial normal mammary cells⁸. Basal-like 60 61 breast cancers are more aggressive and patients suffer from shorter metastasesfree survival compared to those with luminal subtypes^{8,9}. Mechanisms underlying 62 63 increased invasive properties of basal-like tumors include deregulation of the CCL5/CCR5 axis¹⁰, amplified EGFR¹¹ kinase signaling and activation of TGF-B 64 signaling¹². Despite multiple signaling aberrations providing challenges for efficient 65 therapeutic strategies, recent studies have unveiled unique vulnerabilities of basal-66 67 like breast cancers, such as higher levels of PD-L1 expression along with constitutive IFNy signaling activation¹³, in line with higher immune- infiltration 68 69 scores⁶. While the FDA has granted an accelerated approval for atezolizumab, a 70 monoclonal antibody drug targeting PD-L1, plus chemotherapy for the treatment of TNBC¹⁴, the potential application of immune therapies for patients with luminal 71 72 breast cancer remains largely unknown.

73 Among the four intrinsic subtypes, basal and luminal subtypes show opposite histochemical features and notable differences in prognosis^{15,16}, however there is 74 increasing evidence that these subtypes are on a continuum of "luminal-ness" and 75 76 "basal-ness" features. Models of breast cancer lineage evolution describe that basal 77 and luminal progenitor cells are derived from the same bipotential progenitors¹⁷, 78 indicating the potential of lineage reprogramming during cancer progression. Such 79 subtype switching during tumor evolution has been described and is critical for implementation of precision therapeutics¹⁸⁻²⁰. A recent study by Bi et al. reported 80 loss of luminal and gain of basal markers in endocrine resistant breast tumors²¹. 81 82 Mechanisms underlying the intrinsic subtype plasticity are largely unknown, with some exceptions. $JARID1B^{22}$ and $ARID1A^{23}$ have been described as essential 83 84 luminal lineage driver genes and their mutations result in luminal-to-basal subtypes 85 switches. In addition, enhancer reprogramming at GATA3 and AP1 binding sites has been highlighted as a pivotal epigenetic mechanism allowing lineage plasticity²¹. 86

87 ER is well characterized as a luminal lineage marker²⁴. Hotspot mutations in 88 its ligand-binding domain occur in 30%-40% of endocrine resistant breast tumors, 89 promoting ligand-independent ER activation and metastasis²⁵⁻²⁷. Several recent 90 studies showed that *ESR1* mutant tumors are not only associated with endocrine 91 resistance, but also gain unexpected resistance towards CDK4/6 inhibitors²⁸, mTOR 92 inhibitors²⁹ and radiation therapy³⁰ in a mutation subtype and context dependent 93 manner, suggesting potentially more complex re-wiring of ER mutant tumors.

94 We set out to examine whether *ESR1* mutations alter the "luminal-ness" and 95 "basal-ness" balance in breast cancer cell line models and clinical specimens. We

- 96 discovered that ER mutant tumors gain basal-like features, characterized by
- 97 elevated expression of basal cytokeratins as a result of epigenetic reprogramming.
- 98 Immune context analyses in clinical specimens revealed potential therapeutic
- 99 vulnerabilities accompanying the increased basal-ness in *ESR1* mutant breast
- 100 cancer, a finding of potential clinical relevance.

101 Results

102 Basal gene signatures are enriched in *ESR1* mutant breast cancer

103	To examine whether ESR1 mutations alter "luminal-ness" and "basal-ness"
104	we utilized four independent luminal and basal gene signatures (Fig. 1A,
105	Supplementary Table S1). Gene sets from Charafe-Jauffret et al. ³¹ and Huper et
106	al. ³² were obtained from MSigDB (Supplementary Fig. S1A and S1B), and in
107	addition we generated two other gene sets from i) intrinsic subtype genes ³³
108	differentially expressed between luminal (n=33) and basal (n=39) breast cancer cell
109	lines (Supplementary Table S2) $^{34-36}$ and ii) genes differentially expressed between
110	luminal and basal primary tumors in TCGA ³⁷ (Supplementary Fig. S1C and S1D).
111	Although the overlap among the different gene sets was limited (Fig.1B), likely
112	reflecting differences in methodology and sources, some well described lineage
113	marker genes (e.g. ESR1 and FOXA1 as luminal markers, and KRT6A and KRT16
114	as basal markers) were observed in 3 out of 4 gene sets.
115	As expected, all four basal gene sets were significantly enriched in basal
116	versus luminal breast cancer cell lines and tumors (Supplementary Fig. S2A and
117	S2B), and vice versa for luminal gene sets except for the Huper luminal markers,
118	likely due to its derivation from normal mammary tissue (Supplementary Fig. S2C
119	and S2D). We found concordantly increased enrichment of basal gene sets in
120	Y537S and D538G MCF7 ESR1 genome-edited mutant cells, whereas no
121	differences were observed in estrogen treated ESR1 wildtype cells (Fig. 1C). In
122	contrast, we did not observe a consistent change in the luminal gene sets (Fig. 1D).
123	The enrichment of the basal gene sets in the ESR1 mutant cells was also seen in an

124	independent CRISPR-engineered MCF7 ESR1 mutant cell model recently reported		
125	by Arnesen et al ³⁸ (Supplementary Fig. S3A) and in our T47D <i>ESR1</i> mutant cells ²⁷		
126	(Supplementary Fig. S3B). Of note, no consistent and strong alterations of luminal		
127	and basal gene sets enrichment levels were detected in ESR1 WT endocrine		
128	resistant ER+ breast cancer cell models ^{21,39-46} (8 tamoxifen resistant, 2 fulvestrant		
129	resistant and 7 long-term estradiol deprivation (LTED) models), suggesting that the		
130	"basal-ness" shift is a unique feature acquired as a result of ESR1 mutations		
131	(Supplementary Fig. S3C) ⁴⁶ .		
132	We next sought to extend our findings to clinical specimens using RNA-seq		
133	data composed of 51 intra-patient matched ER+ primary-metastatic tumor pairs (7		
134	ESR1 mutant and 44 ESR1 WT pairs) (Supplementary Table S3). Similar to		
135	observations in cell lines, ESR1 mutant metastatic breast cancers showed a		
136	significant enrichment of basal gene signatures compared to tumors with WT ESR1		
137	(Fig. 1E). We did not observe a concurrent decrease of luminal markers (Figure 1F).		
138	Taken together, these findings suggested a novel and unexpected gain of "basal-		
139	ness" in <i>ESR1</i> mutant tumors.		
140			

141 Basal cytokeratins are elevated in *ESR1* mutant breast cancer cells and

142 **tumors**

We next interrogated the union of the four basal gene sets (N=634) to identify which basal marker genes were consistently induced in *ESR1* mutant breast cancer cells. Integrating RNA-seq results from MCF7 cell models²⁷ and clinical samples identified a group of basal cytokeratins (*KRT5, KRT6A, KRT6B, KRT14, KRT16, and*

147 *KRT17*) as the top consistently increased basal markers (Fig. 2A, Supplementary 148 Fig. S4A and Supplementary Table S4). Elevated basal cytokeratins (BCKs) mRNA 149 levels were further confirmed in independent gRT-PCR experiment in ESR1 mutant 150 MCF7 cells (Fig. 2B). Analyzing fold-change expression of all basal markers in a number of MCF7 ESR1 mutant cell models previously described^{25,27,38} revealed 151 152 KRT5, 16 and 17 as the top increased basal genes (Supplementary Fig. S4B-D). In 153 the T47D ESR1 mutant cells, KRT16 was significantly increased (Supplementary 154 Fig. S4E), but the observed enrichment of basal marker genes (Supplementary Fig. 155 S3B) was also driven by other non-canonical basal genes such as WLS and HTRA1 156 (Supplementary Table S5), suggesting some context-dependent mechanisms for the 157 increased basal-ness.

We also queried *KRT* expression in overexpression models. In MCF7 cells
with stable overexpression of HA-tagged WT and mutant ER (Y537S and D538G)
(Supplementary Figure S5A and S5B), we again observed significant
overexpression of *KRT5, KRT6A, KRT6B, KRT16*, and *KRT17* (Supplementary Fig.
S5C).

Given higher BCK mRNA expression in *ESR1* mutant cells, we examined their expression at the protein level. We confirmed higher CK5 and CK16 protein levels in early passage (P6-8) *ESR1* mutant cells, but curiously expression was not detectable in later passages (P30-32) (Supplementary Fig. S6A). This finding was consistent with prior reports on slower growth of CK5+ sub-populations⁴⁷, reflecting selection forces eliminating BCK-positive subclones from luminal cell populations. To determine whether BCK expression was limited to minor sub-populations in *ESR1*

170	mutant cells, we performed IF staining for CK5, CK16 and CK17 in early passage
171	cells (below P12) (Fig. 2C). No BCK positive clones were identified in MCF7-WT
172	cells, while 0.5-1% of Y537S and D538G ESR1 mutant cells exhibited strong diffuse
173	cytoplasmic CK5/16/17 expression. In addition, 3-5% of ESR1 mutant cells
174	displayed strong BCK signals localized as foci adjacent to the nucleus
175	(Supplementary Fig. S6B), and this was again not observed in the WT cells.
176	Furthermore, co-staining of CK5+CK16 and CK16+CK17 showed that the BCK
177	proteins were predominantly (in 75%-90% imaged cells) upregulated in the same
178	sub-population of cells (Supplementary Fig. S6C and S6D). In contrast, luminal
179	cytokeratin CK8 was homogenously expressed with stronger expression at the
180	edges of each cell cluster (Supplementary Fig. S6E), suggesting that the marked
181	heterogeneity was a unique feature for BCK expression in the luminal cell
182	background.

183

BCK induction is independent of mutant ER DNA binding but requires low ER expression

Mutant ER can function in a ligand-independent manner ^{26,27}, and we thus tested whether induction of BCKs resulted from ligand-independent ER activity. We interrogated eight publicly available RNA-seq and microarray data sets with estradiol (E2) treatment in six different ER+ breast cancer cell lines ^{26,27,48-51}. In contrast to strong E2 induction of classical ER target genes such as *GREB1, TFF1* and *PGR*, expression of basal and luminal cytokeratins genes was not regulated by E2 with the exception of *KRT7* (Fig. 3A). We then examined whether BCK expression was

193 regulated via *de novo* genomic binding of mutant ER at BCK genes. We performed 194 ChIP-seq in MCF7 WT and ESR1 mutant cells in the absence and presence of E2. 195 As expected, in the absence of E2 we detected very few ER binding sites in WT 196 MCF7 cells (n=125), whereas E2 stimulation triggered substantial ER binding events (n=12,472) (Supplementary Table S6). Consistent with previous studies^{25,26}. Y537S 197 198 and D538G ER show strong ligand-independent binding, with 657 binding sites in 199 Y537S and 1,016 in D538G mutant cells (Supplementary Fig. S7A). The GREB1 200 gene locus is shown as a representative example (Fig 3B, left panel). Co-201 occupancy analyses between WT-E2 and mutant-vehicle sets demonstrated that 202 one third of all Y537S (36%) and D538G (31%) ER binding sites were not detected 203 in the WT+E2 data suggesting gain-of-function novel binding sites (Supplementary 204 Fig. S7B); however, none of them mapped to the BCKs genes with increased 205 expression in ESR1 mutant cells (-/+ 50kb of transcriptional start sites) (Fig. 3B, 206 middle and right panel).

207 We then expanded our analyses and examined potential estrogen-regulation 208 of all basal marker genes, again using the union of the four basal gene sets (N=634). 209 Comparison of E2 and ESR1 mutation-conferred fold changes of these genes in 210 MCF7 cells revealed that the top upregulated basal markers in ESR1 mutant cells 211 were not E2-induced (Supplementary Fig. S7C and S7D). In addition, only 20 basal 212 genes (3%) harbor mutant ER binding sites at -/+ 50 kb of TSS (Supplementary Fig. 213 S7E), and 18 of those were not differentially expressed between WT and mutant 214 cells (Supplementary Fig. S7F). Taken together, these analyses suggest that the

shift to "basal-ness" in *ESR1* mutant cells was not mediated via ligand-independent
binding of mutant ER to BCK gene loci.

217 To further understand interplay between ESR1 and KRT gene expression, we 218 determined expression of basal and luminal KRT genes in ER+ primary breast 219 tumors. As shown in Figure 3C, the six BCKs were significantly negatively correlated 220 with ESR1 expression, whereas the luminal KRT were mostly positively correlated 221 with ESR1 (Fig. 3C). Luminal KRT7 was again the exception, being negatively 222 correlated with ESR1 expression, in line with it being repressed by ER (Figure 3A). 223 The inverse correlation between BCK and ESR1 expression was also reflected in 224 results from ER knockdown experiments, in which loss of ESR1 significantly 225 increased expression of BCKs in MCF7 WT and mutant cells (Fig. 3D). Similar 226 results were obtained in five additional ER+ breast cancer cell lines where we 227 observed a general increase of BCK expression after ESR1 knockdown 228 (Supplementary Fig. S8). In addition, co-staining of ER and CK5/CK16/17 in MCF7 229 ESR1 mutant cells showed significantly lower ER expression in BCK+ cells than in 230 the surrounding BCK- cells (Fig. 3E). Collectively, these data demonstrate that ER 231 serves as a negative regulator of BCKs expression independent of ligand and 232 mutational status, and suggest that low ER expression is likely necessary but not 233 sufficient to facilitate BCKs overexpression in a subpopulation of ESR1 mutant cells. 234 These data also support a role for mutant ER in regulating BCK expression via 235 epigenetic regulation, a mechanism that we have recently shown to be used by 236 mutant ER ³⁸.

237

238 PR regulation of BCK expression through binding at a CTCF-driven chromatin

loop at the *KRT14*/16/17 loci in *ESR1* mutant cells

240 To investigate potential epigenetic regulation of *KRT5/6A/6B* and

241 *KRT14/16/17*, we first compared their regional epigenetic landscapes on

chromosome 12 and 17, respectively, in luminal and basal breast cancer cell lines

and tumors (Supplementary Fig. S9). Integrative analysis of ATAC-seq and ChIP-

seq profiles of H3K4me2, H3K4me3, H3K9ac and H3K27ac suggested that these

two regions are epigenetically silent in MCF7 (Supplementary Fig. S9A), consistent

with low expression. In basal breast cancer cell lines and tumors, there is an

enrichment of H3K27 acetylation (Supplementary Fig. S9B) and number of ATAC-

248 seq peaks (Supplementary Fig. S9C) at BCK loci, consistent with increased mRNA

expression (Supplementary Fig. S9E and S9F). This is also observed in ESR1

250 mutant cell models (Supplementary Fig. S9G).

251 We recently reported CCCTC-binding factor (CTCF) motif as one of the top 252 enriched motifs in unique ESR1 mutant-regulated accessible genomic regions³⁸. To 253 determine whether CTCF has a role in the epigenetic regulation of BCK, we 254 developed a CTCF gene signature by identification of the top 100 differentially expressed genes before and after CTCF knockdown in MCF7⁵² (Supplementary 255 256 Table S1). The positively correlated CTCF signature (i.e. using genes that were 257 repressed after CTCF knockdown) was significantly enriched in both MCF7 ESR1 258 mutant cells (Fig. 4A) and metastatic tumors (Fig. 4B) compared to their WT 259 counterparts, whereas E2 stimulation had no effect (Fig. 4A). CTCF is a multimodal 260 epigenetic regulator in breast cancer⁵³, in part through generating boundaries of

topological associating domains (TADs) and guiding of DNA self-interaction⁵⁴.

- 262 Mapping the genomic occupancy of CTCF and three other cohesion complex
- 263 members (RAD21, STAG1 and SMC1A) in MCF7 cells⁵⁵⁻⁵⁷ (Fig. 4C) identified five
- 264 putative TAD boundaries at the KRT14/16/17 (Fig. 4D) loci and three at the
- 265 KRT5/6A/6B (Supplementary Fig. S10A) loci. Integration of an additional MCF7
- 266 CTCF ChIA-PET dataset⁵⁸ showed that a strong chromatin loop is predicted to span
- the *KRT14/16/17* genes, further supported by the pattern of convergent CTCF motif
- orientations at the predicted TAD boundaries (Fig. 4C). Since the *KRT5/6A/6B* locus
- did not harbor strong chromatin loops (>3 linkages), we focused our further analysis
- 270 on the *KRT14/16/17* locus.
- 271 ChIP revealed strong enrichment of CTCF binding at the base of the
- 272 chromatin loops of the KRT14/16/17 locus in ESR1 mutant cells, however there was
- a lack of E2 regulation (Fig. 4E). Decreasing CTCF levels led to increased
- 274 expression of KRT14, KRT16 and KRT17 mRNA levels in ESR1 mutant cells,
- 275 potentially reflecting a role for CTCF as "classical" insulator, suppressing high
- 276 expression of these BCKs through the identified super enhancer at the KRT14,
- 277 KRT16 locus (Figure 4F). Given identification of progesterone receptor (PR) binding
- sites within this super enhancer, PR's previously identified role in regulating *KRT5*
- 279 expression in luminal breast cancer cells^{47,59}, and finally its overexpression in
- 280 multiple *ESR1* mutant cell models ^{25-27,60} (Supplementary Fig. S10C and S10D), we
- 281 tested whether PR regulates *KRT14/16/17* expression.
- 282 PR ChIP-seq revealed a ligand-inducible PR binding sites in MCF7 cells 283 approximately 32kb upstream of the *KRT14/16/17* loop region⁶¹ (Fig. 4F). This PR

binding site overlapped with a curated super-enhancer in MCF7 cells⁶², which was 284 285 additionally supported by strong active histone modifications (Fig. S9). Knockdown 286 of PR partially rescued the increased expression of KRT14, 16 and 17 in both ESR1 287 mutants (Fig. 4G and Supplementary Fig. S10E). We also observed a similar rescue 288 effect for KRT5 (Supplementary Fig. S10E), consistent with previous studies⁵⁹. 289 Furthermore, both PR agonist (P4) and antagonist (RU486) treatment increased 290 KRT5, 16 and 17 expression in Y537S ESR1 mutant cells, while only RU486 291 triggered KRT5 and KRT16 expression in D538G mutant (Fig. 4H and 292 Supplementary Fig. S10F). The marked induction effect of RU486, a PR antagonist, 293 is likely due to its previously reported partial agonism via recruitment of coactivators⁶³. The RU486-induced CK5 and CK16 increase was further examined 294 295 by IF, where CK5 (Supplementary Fig. S10G) and CK16 (Fig. 4I and 4J) positive 296 cells increased from 1% to 5%. Of note, CK17 positive cells were not increased by 297 RU486 treatment (Supplementary Fig. S10G), suggesting translational efficiency 298 differences between different BCK subtypes. Together, these data demonstrated 299 that elevated PR expression in ESR1 mutant cells was essential for BCKs induction, and this was possibly due to an orchestration with a super enhancer which is 300 301 accessible to regulate KRT14/16/17 genes via the CTCF-driven chromatin loop.

302

303 Enhanced immune activation, associated with S100A8-S100A9 secretion and 304 signaling in *ESR1* mutant tumors

Finally, we investigated whether the increased expression of basal genes in
 ESR1 mutant tumors confers basal-like features and potentially novel therapeutic

307 vulnerabilities. To identify basal cytokeratin-associated pathways enriched in ER 308 mutant tumors, we at first identified ER+ tumors with the top and bottom quantile of 309 BCK gene enrichment and then computed hallmark pathways differentially enriched 310 between these two groups (Supplementary Fig. S11A). Intersection of these BCKs-311 associated pathways with those enriched in ESR1 mutant metastases uncovered 312 seven shared molecular functions, the top four of which are all related to immune 313 responses (Fig. 5A, Supplementary Fig. S11B, S11C and Supplementary Table S7). An orthogonal approach - bioinformatic evaluation using ESTIMATE⁶⁴ - confirmed 314 315 enhanced immune activation in BCK-high vs BCK-low ER+ tumors albeit still lower 316 than in basal tumors (Fig. 5B). In addition, BCK-high tumors displayed higher lymphocyte and leukocyte fractions according to a recent biospecimens report⁶⁵ (Fig. 317 318 5C), and higher PDCD1 mRNA levels (Supplementary Fig. S11D). Intriguingly, 319 patients with BCK-high ER+ tumors experience improved outcomes (Fig. 5D), and 320 although entirely speculative at this point in time, one could hypothesize that this 321 might be due to increased anti-tumor immune activation.

Similar to BCK-high ER+ tumors, *ESR1* mutant metastatic tumors exhibited higher immune scores compared to those with *ESR1* WT (Fig. 5E). Immune cell subtype deconvolution^{66,67} revealed significantly higher CD8+ T, NK and dendritic cells, along with macrophages in *ESR1* mutant tumors. Basal breast cancers harbor high immune infiltrations at least in part due to higher tumor mutation burden (TMBs)⁶⁸, however, we did not detect higher TMB in BCK-high vs low ER+ tumors (Supplementary Fig. S11E).

329 To understand which factors might contribute to immune activation in ESR1 330 mutant and BCK-high ER+ tumors, we compared gene expression of major immune genes derived from ESTIMATE⁶⁹ (n=141) between ESR1 mutant and WT tumors, 331 332 and BCK-high vs BCK-low ER+ tumors. This analysis identified S100A8 and S100A9 333 as the two top consistently increased immune-related genes (Fig. 6A), and this 334 overexpression was also seen in MCF7 ESR1 mutant cell models (Supplementary 335 Fig. S11F). S100A8 and S100A9 are pro-inflammatory cytokines that form heterodimers and play crucial roles in shaping immune landscapes^{45,46}. As 336 337 expected, S100A8-A9 expression correlated positively with immune scores in ER+ 338 tumors (Fig. 6B). BCKs levels failed to differentiate immune scores in ER+ tumors 339 among the subset of tumors exhibit high S100A8-A9 (Fig. 6B). S100A8-A9 are 340 secreted proteins and function as heterodimers. To confirm S100A8-A9 protein 341 overexpression, we measured S100A8-A9 heterodimer levels in plasma samples 342 from patients with ESR1 WT (n=7) and mutant (n=11) tumors (Supplementary Table 343 S8) (Fig. 6C). This analysis revealed significantly higher circulatory S100A8-A9 344 heterodimers concentrations in plasma from patients with ESR1 mutations (Fig. 6D). 345 S100A8-A9 heterodimer mainly stimulates downstream cascades through two 346 receptors: toll-like receptor 4 (TLR4) and receptor for advanced glycation end 347 products (RAGE), and both of them are widely reported to impact cancer immunity. 348 A further gene set variation analysis in WCRC/DFCI primary-matched paired 349 metastatic samples revealed consistent enrichment of both pathways in ESR1 350 mutant tumors (Fig. 6E, Supplementary Table S1), suggesting both TLR4 and RAGE 351 signaling are hyperactive in ESR1 mutant tumors.

352	To further elucidate the specific cell-cell communication by S100A8/S100A9
353	signaling, we analyzed RAGE and TLR4 signaling via measuring ligand and receptor
354	expression in different cell types using single-cell RNA-seq data from two breast
355	cancer metastases. Highest expression of S100A8/S100A9 was seen in epithelial
356	cells, followed by fibroblast and macrophages. In contrast, TLR4 and AGER (RAGE)
357	showed low expression in the epithelial cells, but instead were widely expressed in
358	the stroma, especially in fibroblasts and macrophages. In general, AGER displayed
359	lower expression levels in all cell types compared to TLR4 (Fig. 6F and 6G).
360	Taken together, these data support the concept that the increase in basal-
361	ness of ESR1 mutant tumors is associated with immune activation, in part facilitated
362	by the paracrine S100A8/A9-TLR4 signaling.

363 **Discussion**

364 Recurrence of ER+ breast cancer causes over 24,000 deaths each year in 365 the US alone. Given that ESR1 mutation occur in as many as 20-30% of metastatic 366 recurrences, it is imperative to identify therapeutic vulnerabilities through dissecting 367 mechanisms of action. In this study we have uncovered a previously unrecognized 368 plasticity of ESR1 mutant cells, reflected by enrichment of basal subtype genes in 369 ESR1 mutant tumors and in particular a gain of BCK expression, resulting from 370 epigenetic reprogramming of a mutant ER-specific PR-linked chromatin loop. This 371 molecular evolution, i.e. an increase of basal-like feature in the ESR1 mutant tumors 372 was associated with immune activation including enhanced S100A8/A9-TLR4 373 signaling (Fig 7).

374 Increased plasticity of tumors has previously been shown to be associated with tumor initiation and progression^{21,46,70-72}. PAM50 intrinsic subtype switching has 375 been described to occur in as many as 40% of breast cancer metastases²⁰. Here we 376 377 show that ESR1 mutant cells gain basal-ness, and a similar observation was recently reported by Gu et al.⁷³ showing a luminal to basal switch in MCF7 ESR1 378 379 Y537S CRISPR cells compared to parental cells. However, luminal to basal subtype switching is rare in breast cancer²⁰ and we have previously reported on clinically 380 381 relevant gene expression changes in brain metastases (increased in HER2 gene expression) without clear subtype switching¹⁸. These results are in line with the 382 383 increasing appreciation of the molecular subtypes being on a continuum rather than 384 representing discrete stages. Of note, we did not observe a similar gain of basal-385 ness in a series of ESR1 wildtype endocrine resistant in vitro models, with the

386 exception being a study revealing a "luminal-to-basal" switch in an estradiol-deprived

387 T47D xenograft derived cell line, indicating a potential role for the microenvironment

in mediating a similar switch in ER wildtype tumors 74 .

389 We propose that the observed *ESR1* mutant-cancer cell state

390 interconversions are of potential clinical relevance due to increased stromal immune

391 activation associated with the induction of BCK. Using *in silico* gene expression,

392 pathway analyses and pathology information, we observed increased activation of a

number of immune-related pathways including S100A8/S100A9-TLR4 signaling and

394 increased lymphocytic infiltration. S100A8/S100A9 heterodimers exhibit pro-

inflammatory properties in different contexts in breast cancer^{75,76}, are associated

³⁹⁶ with poor prognosis in multiple cancer types³⁶ including breast cancer⁷⁷, and

397 blockade of their activity improves survival⁷⁸. We observed increased

398 S100A8/S100A9 levels in blood from patients with *ESR1* mutant tumors but given

399 complexity of tumor-cell intrinsic and extrinsic roles of the inflammatory mediators

400 and their receptors (also supported by our single cell sequencing analysis) additional

401 work is needed to understand if and how they contribute to tumor progression in

402 patients with ER mutant tumors. This should include an analysis of MDSC in this

403 setting since they have been described to play an important role in S100A8/A9

404 function^{76,79}. This is also supported by our recent studies showing an enrichment of

405 immune-suppressive macrophages in ER mutant tumors, along with increased

406 expression of interferon regulated genes⁸⁰. Together, these data imply opportunities

407 for immune therapies for patients with ER mutant tumors that should be analyzed408 further.

We and others^{26,27,38} previously identified genes that have altered expression 409 410 in ESR1 mutant cells but are not E2 regulated in WT cells. Here, all six BCK belong 411 to this group of novel, gain-of-function target genes. BCK are not regulated as a 412 result of ligand-mimicking nor *de novo* transactivation by mutant ER, and their 413 expression is strongly and negatively correlated with ER levels. A similar correlation 414 was also observed with P4-induced CK5+ luminal breast cancer cells displaying low ER and PR levels⁵⁹. One possible explanation is that ER, regardless of its liganded 415 416 status or genotype, serves as a direct epigenetic suppressor that represses BCK 417 expression to maintain luminal identity. For example, it has been shown that ER 418 silences basal, EMT and stem cell related genes by recruiting pivotal methyl-419 transferases like EZH2 and DNMTs to reshape the DNA and histone methylation 420 landscape⁸¹. More studies are required to further elucidate the regulatory network 421 between ER and BCKs. Given bi-directional interactions between tumor and stromal 422 cells in BCK regulation, it will be important to perform future studies in improved 423 model systems such as those recently described for analysis of complex regulation of CK14 expression and function⁸². 424

Assessment of BCK expression revealed that a 50-fold increase in mRNA was reflected in only ~1% cells being positive for BCK protein. This finding is consistent with a previous study showing that P4 stimulation of breast cancer cells caused a 100-fold induction of CK5 promoter activation ultimately translating to 1-10% of cells positive for CK5 protein⁵⁹. In addition, discordance between mRNA and protein of CK7 and CK14 in breast cancer tissue has been documented⁸³. It is possible that BCK protein translation in luminal cells is aberrant, resulting in poorly

432 localized or transported protein, consistent with our detection of BCK protein foci 433 rather than the broad distribution pattern over full cytoskeleton similar to what has 434 been previously reported for example for formation of CK17 foci. The discordance in 435 mRNA and protein expression may be due to the cell heterogeneity, with individual 436 cells having high mRNA and protein compared to the negative population, potentially due to heterogenous expression of miRNAs regulating BCK expression⁸⁴. These 437 438 BCKs positive cells might be pre-selected by multiple genetic and epigenetic cues 439 including but not limited to low ER expression and chromatin loop formation as 440 identified in our study. The discordance between mRNA and protein expression may 441 also help to explain differences in prognosis using mRNA expression profiling like in our study vs IHC in previous studies^{85,86}. 442

443 We provide evidence to support BCK as emerging biomarkers of ESR1 444 mutant breast cancer and its prognosis, yet their direct functional impact remains 445 ambiguous. CK14 positive cells typically lead collective invasion across major subtypes of breast cancer cells⁸⁷, and this is in line with previously identified 446 enhanced cell migration in *ESR1* mutant cells⁸⁸. In addition, as previously described, 447 448 CK5 positive luminal cells acquire stem-like properties and chemotherapy resistance^{47,59}. Importantly, we found several other consistently increased basal 449 450 marker genes such as interferon-alpha inducible protein 27 (IFI27). Previous studies have reported a role of IFI27 in regulating innate immunity in breast cancer⁸⁹ and 451 cisplatin resistance in gastric cancer³⁶. Thus, the "basal-ness" shift might confer 452 453 several broad functional alterations to ESR1 mutant tumors.

454 We identified a PR-orchestrated TAD at the KRT14/16/17 genomic locus in 455 ESR1 mutant cells, and we propose that the simultaneous generation of a *de novo* 456 CTCF loop and ER ligand-independent PR overexpression is necessary for 457 KRT14/16/17 in ESR1 mutant cells. Intriguingly, knockdown of CTCF selectively 458 increased KRT14/16/17 mRNA levels whereas knockdown of PR blocked their 459 induction in ESR1 mutant cells. This unexpected discrepancy may highlight that 460 CTCF binding may simultaneously serve as a transcriptional insulator to restrict KRT14/16/17 in an inactive compartment^{53,90}. Importantly, data indicates that CTCF 461 462 knockdown alone is not sufficient to eliminate TAD but instead promotes the formation of new chromatin interactions that alter gene expression⁹¹. We also 463 464 unexpectedly found that both PR agonist P4 and PR antagonist RU486 elevated 465 BCK expression, which was inconsistent with previous reported findings where P4 and RU486 exhibited opposite effects in regulating CK5⁵⁹. Given RU486 is well-466 467 characterized for its partial agonism, it is possible that ESR1 mutant cells uniquely 468 express a particular strong PR coactivator that confers the partial agonism of RU486 469 in this context. Another possibility is that RU486 alternatively stimulates other nuclear receptors such as glucocorticoid^{85,92} or potentially even androgen receptor⁹³ 470 471 to reprogram BCKs expression. The reversed PR pharmacological response in 472 ESR1 mutant cells is intriguing and warrants future investigation.

Our study discovered a unique aspect of *ESR1* mutant cells and addressed
the underlying mechanisms as well as its clinical relevance, albeit with some
remaining limitations, such as limited numbers of clinical samples due to inherent
difficulties of obtaining metastatic tissues. The enhanced immune infiltration requires

477 additional validation by TIL counting on *ESR1* mutant tumor sections. Confirmation

- 478 and studies in *in vivo* models should be included into future studies. Our preliminary
- 479 analysis in a ESR1 Y541S (mouse ortholog of Y537S mutation) knockin mouse
- 480 model showed overexpression of BCK at RNA and protein level in mammary
- 481 tumors⁹⁴. And finally, the *in silico* prediction of enhanced CTCF-driven chromatin
- 482 loop at the basal cytokeratin gene locus requires confirmation by orthogonal
- 483 approaches, such as chromosome conformation capture. Nonetheless, our study
- 484 serves as a robust pre-clinical report uncovering mechanistic insights into ESR1
- 485 mutations and their roles in conferring basal-like feature to ER+ breast cancer and
- 486 implicates the immune therapeutic vulnerabilities to this subset of patients.

487 Materials and methods

- Additional details are provided in the Supplementary Materials and Methodssection.
- 490 Human tissue and blood studies
- 491 51 paired primary matched metastatic samples were from DFCI (n=15) and
- 492 our Women's Cancer Research Center (WCRC) (n=36) cohorts as previously
- 493 reported^{95,96}. For all WCRC metastatic samples, *ESR1* mutations status were called
- 494 from RNA-sequencing. For bone/brain/GI metastatic lesions, *ESR1* mutations status
- 495 were additionally examined using droplet digital PCR for Y537S/C/N and D538G
- 496 mutations in *ESR1* LBD region as previously reported⁹⁷. For DFCI cohort, *ESR1*
- 497 mutations were all called from matched whole exome sequencing⁹⁸.
- 498 For the study of patients' blood, all patients provided written informed consent
- and all procedures were approved by the University of Pittsburgh Institutional
- 500 Review Broad (PRO17080172). 18 patients diagnosed with late-stage metastatic
- 501 ER+ breast cancer were recruited. Procedure to identify hotspot *ESR1* mutations
- 502 has been previously described by us⁹⁹.

503 Cell culture

- 504 Establishments of rAAV-edited (Park lab)²⁷, CRISPR-Cas9-edited (Gertz³⁸
- ⁵⁰⁵ and Ali²⁵ lab) and CRISPR-Cas9-edited T47D cells²⁷ were reported previously.
- 506 ZR75-1 (CRL-1500), MDA-MB-134-VI (HTB-23), MDA-MB-330 (HTB-127) and MDA-
- 507 MB-468 (HTB-132) were obtained from the ATCC. Development of BCK4 cells has
- 508 been previously reported¹⁰⁰.

509 S100A8/S100A9 heterodimer ELISA

510Human S100A8/S100A9 heterodimer amounts in human plasma samples511were quantified using S100A8/S100A9 heterodimer Quantikine ELISA kit (R&D

- 512 System, DS8900) following the manufacture protocol. All plasma samples were first
- 513 diluted in calibration buffer with 1:50 ratio and loaded into antibody-coated plate.

514 Chromatin-immunoprecipitation (ChIP) and sequencing analysis

515 ChIP was performed as previously described ⁵¹. ChIP-seq reads were aligned

to hg38 genome assembly using Bowtie 2.0¹⁰¹, and peaks were called using

517 MACS2.0 with p value below 10E-5 ¹⁰². We used DiffBind package ¹⁰³ to perform

518 principle component analysis, identify differentially expressed binding sites and

analyze intersection ratios with other data sets. Heatmaps and intensity plots for

520 binding peaks were visualized by EaSeq. Annotation of genes at peak proximity was

521 conducted using ChIPseeker ¹⁰⁴, taking the promoter region as +/- 3000 bp of the

522 transcriptional start site (TSS) and 50kb as peak flank distance.

523 **RNA sequencing analysis**

524 RNA sequencing data sets were analyzed using R version 3.6.1. Log2

525 (TPM+1) values were used for the RNA-seq of Oesterreich *ESR1* mutant cell models

526 and TMM normalized Log2(CPM+1) values were used for Gertz RNA-seq data.

527 TCGA reads were reprocessed using Salmon v0.14.1¹⁰⁵ and Log2 (TPM+1) values

528 were used. For the METABRIC data set, normalized probe intensity values were

- 529 obtained from Synapse. For genes with multiple probes, probes with the highest
- 530 inter-quartile range (IQR) were selected to represent the gene. For pan-breast
- 531 cancer cell line transcriptomic clustering, 97 breast cancer cell line RNA-seq data

were reprocessed using Salmon and merged from three studies³⁴⁻³⁶, batch effects
were removed using "removeBatchEffect" function of "limma¹⁰⁶" package. Gene set
variation analysis was performed using "GSVA" package¹⁰⁷. Survival comparisons
were processed using "survival" and "survminer" packages¹⁰⁸ using Cox
Proportional-Hazards model and log-rank test. Data visualizations were performed
using "ggpubr¹⁰⁹", "VennDiagram¹¹⁰" and "plot3D¹¹¹".

538 For the single cell RNA seq analysis, two fresh bilateral bone metastases 539 (BoMs) were collected from a patient initially diagnosed with ER+ primary breast 540 cancer, dissociated into single cells and a cell suspension with at least 70% viability 541 was submitted for library preparation using 10X genomics chromium platform (V3.0 542 chemistry) (Ding et al, manuscript in preparation). 6,000 cells were targeted for each 543 BoM, and the final libraries were sequenced at a depth of 67,000 reads per cell 544 using NOVAseq.

545 **Tumor Mutation Burden Analysis**

Tumor mutation burden (TMB) calculation was performed as previous
described¹¹². Briefly, TCGA mutation annotation files from 982 patients were
downloaded from FireBrowse and mutation subtypes were summarized using
"maftool" package¹¹³. Mutations subtypes were classified into truncated (nonsense,
frame-shift deletion, frame-shift insertion, splice-site) and non-truncated mutations
(missense, in-frame deletion, in-frame insertion, nonstop). TMB was calculated as
2X Truncating mutation numbers + non-truncating mutation numbers.

553 Generation of Gene Sets

- 554 For Sorlie et al., the original set of intrinsic genes were downloaded from
- 555 Stanford Genomics Breast Cancer Consortium (<u>http://genome-</u>
- 556 <u>www.stanford.edu/breast_cancer/</u>). 453 genes were annotated from 553 probes.
- 557 Expression of these 453 genes were examined in 33 luminal and 39 basal breast
- 558 cancer cell lines. Significantly higher (FDR<0.01) intrinsic genes in basal or luminal
- cells were called as basal (n=75) or luminal (n=68) markers in Sorlie gene sets. For
- the TCGA gene set, differentially expressed genes were called between basal and
- 561 Iuminal A or basal and Iuminal B ER+ tumors using raw counts. The top 200
- 562 increased genes of these two comparisons were further intersected. Overlapped DE
- 563 genes in basal (n=164) and luminal (n=139) tumors were called as TCGA gene sets.
- 564 For CTCF gene signature establishment, a previous RNA-seq data set on MCF7
- 565 cells with or without CTCF knockdown was downloaded and analyzed⁵², top 100
- 566 downregulated genes with CTCF knockdown were used as the CTCF gene
- 567 signature.

568 Chromatin interaction data analysis

569 CTCF ChIA-PET data were downloaded from GSE72816. Chromatin linkages were 570 visualized on 3D genome browser (<u>http://promoter.bx.psu.edu/hi-c/</u>) after processed 571 with ChIA-PET tool¹¹⁴. Confident TAD boundaries were defined by the colocalization 572 of CTCF and cohesion complex subunits together with called chromatin interactions.

- 573 **Data Availability**
- 574 ER ChIP-seq data from MCF7 *ESR1* mutant cell model was deposited in 575 Gene Expression Omnibus with accession number of GSE125117. MSigDB curated 576 gene sets were downloaded from GSEA website

577	(http://software.broadinstitute.org/gsea/msigdb/index.jsp). RNA-seq data and clinical
578	information from TCGA and METABRIC were obtained from the GSE62944 and
579	Synapse software platform (syn1688369) respectively. TCGA biospecimen immune
580	profile data were downloaded from Saltz et al ⁶⁵ . TCGA mutation annotation format
581	(MAF) files and methylation data were downloaded from FireBrowse website
582	(http://firebrowse.org/). Complete RNA-Seq data for the DFCI metastases samples
583	will be published separately. RNA-Seq data from the WCRC cohorts are available at
584	Lee-Oesterreich Lab Github repository (https://github.com/leeoesterreich). All the
585	raw data and scripts are available upon request from the corresponding author.
586	Sources of all public available data sets used in this study are summarized in
587	Supplementary Table S10.
500	

588

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607

608 Competing Interests

609 The authors declare no conflict of interests.

610

611 Author Contributions

- Z.L., J.M.A., A.V.L. and S.O. conceived and designed the study. Z.L., Y.W., A.B. and
- K.D. designed, performed and analyzed experiments. Z.L., A.B., N.M.P. and K.D.
- 614 performed bioinformatic analysis. J.M.A., L.M. and M.R. contributed to clinical
- sample collection and intellectual input. N.W. provided extended RNA-seq data set
- 616 (DFCI) from clinical specimens and intellectual input. Z.L., A.V.L., S.O., C.A.S.,
- J.K.R., W.J.M. and J.G. contributed to data interpretation and provided additional
- 618 intellectual input. L.B., S.A. and J.G. provided additional cell models for this study
- and intellectual input. Y.F., L.Z. and G.C.T. provided and validated biostatistical

- 620 approaches of all the analysis. Z.L., A.V.L. and S.O. developed the figures and the
- 621 manuscript. All the authors reviewed and agreed with the contents of the manuscript.

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902 Figure legends

Figure 1. Basal breast cancer gene sets are enriched in *ESR1* mutant breast cancers

A) Four pairs of luminal/basal gene sets applied in this study with gene numbers

- 906 specified in each set.
- B) Venn diagram representing the overlap of genes from the basal (left) and luminal

908 (right) gene sets. Genes overlapping in at least three gene sets are indicated.

- 909 C) and D) Dot plots showing GSVA score of the four pairs of basal (C) and luminal
- 910 (D) gene sets enrichment in MCF7 genome-edited cell models. Scores from luminal
- and basal breast cancer cell lines were used as positive controls. Dunnett's test was
- 912 used to compare with WT-vehicle set within each gene set. (* p<0.05, ** p<0.01)
- E) and F) Box plots representing basal (E) and luminal (F) gene set enrichments in
- 914 intra-patient matched paired primary-metastatic samples. Delta GSVA score for
- 915 each sample was calculated by subtracting the scores of primary tumors from the
- 916 matched metastatic tumors. Mann-Whitney U test was performed to compare the
- 917 Delta GSVA scores between WT (N=44) or *ESR1* mutation-harboring (N=7) paired

918 tumors. (* p<0.05)

919

Figure 2. Overexpression of basal cytokeratins (BCK) in *ESR1* mutant breast cancer cells and tumors

A) Correlation between basal gene fold changes (FC) in MCF7-Y537S/D538G cells

- 923 (normalized to WT vehicle) and intra-patient paired mutant tumors (normalized to
- 924 WT tumors) (N=634). Consistently increased or decreased genes in the two MCF7

925 mutant cells and tumors compared to their WT counterparts were highlighted in red

- 926 or blue respectively, and six basal cytokeratin genes are indicated. Inconsistently
- 927 changed genes among the three comparisons are labelled in black.
- 928 B) *KRT5/6A/6B/14/16/17* mRNA levels in MCF7 WT and *ESR1* mutant cells.
- 929 Relative mRNA fold change normalized to WT cells and RPLP0 levels measured as
- 930 the internal control. Each bar represents mean ± SD with three biological replicates.
- 931 Representative results from three independent experiments are shown. Dunnett's
- test was used to compare BCKs expression levels between WT and mutant cells.
- 933 C) Representative images of immunofluorescence staining on CK5, CK16 and CK17
- 934 in MCF7 WT and ESR1 mutant cells. Regions with CK positive cells were
- highlighted in the magnified images. MDA-MB-468 was included as positive control.
- 936 Images were taken under 20x magnification.
- D) Quantification of percentages of CK positive cells in MCF7 WT and *ESR1* mutant
- cells. Each bar represents mean ± SD from four different regions. Data shown are
- 939 from one representative experiment of three independent experiments. Dunnett's
- 940 test was used to compare BCKs positive cell prevalence between WT and mutant

941 cells. (* p<0.05, ** p<0.01)

942

Figure 3. Basal cytokeratins induction is independent of mutant ER genomic binding but requires low ER expression.

A) Heatmap representing fold change mRNA expression (E2/veh) of six basal cytokeratins and four luminal cytokeratins in ER+ breast cancer lines from six publicly available data sets (GSE89888, GSE94493, GSE108304, GSE3834,
GSE38132 and GSE50693). *GREB1*, *PGR*, and *TFF1* are canonical E2-regulated
genes included as positive controls.

B) Genomic track showing ER binding intensities at *KRT5/6A/6B* and *KRT14/16/17*

951 loci from ER ChIP-seq data sets of MCF7 *ESR1* mutant cells. *GREB1* locus serve as

952 a positive control.

953 C) Graphic view of Pearson correlation between expression of *ESR1* and each basal

954 or luminal cytokeratin in ER+ breast tumors in TCGA (n=808) and METABRIC

955 (n=1,505) cohorts. Color scale and size of dots represent correlation coefficient and

956 significance, respectively.

D) qRT-PCR measurement of ESR1, KRT5/6A/6B/14/16/17 mRNA levels in MCF7

958 WT and ESR1 mutant cells with ESR1siRNA knockdown for 7 days. mRNA fold

change normalized to WT cells; RPLP0 levels were measured as internal control.

Each bar represents mean ± SD with three biological replicates. Data shown are

961 representative from three independent experiments. Student's t-test was used to

compare the gene expression between scramble and knockdown groups. (* p<0.05,

963 ** p<0.01)

E) Representative images of ER, CK5, CK16 and CK17 staining in MCF7-Y537S

and D538G cells. BCKs positive cells are highlighted with white arrows. Images

966 were taken under 20x magnification.

967 F) Bar plots quantifying the ER intensities in BCKs positive (blue) and the

968 corresponding proximal negative (red) cells from each region. Each bar represents

969 mean ± SD analyzed in five different regions per group from one experiment,

- 970 representative of three independent experiments. Paired t test was applied to
- 971 compare ER intensities between BCKs positive and negative cells. (* p<0.05, **
- 972 p<0.01)
- 973

974 Figure 4. Basal cytokeratins are induced via a unique PR enhancer-associated

- 975 TAD in ESR1 mutant cells.
- A) Dot plots showing enrichment levels of CTCF gene signature in MCF7 ESR1
- 977 mutant cells. Dunnett's test was used to compare the difference. (** p<0.01)
- B) Dot plots showing enrichment levels of CTCF gene signature in *ESR1* WT (n=44)
- 979 and mutant (n=7) metastases. Mann-Whitney U test was used to compare
- 980 enrichment levels in tumors. (* p<0.05)
- 981 C) Genomic track illustrating the CTCF/cohesion complex binding at *KRT14*/16/17
- 982 proximal genomic region in MCF7 cells. CTCF and RAD21 ChIP-seq were
- 983 downloaded from ENCODE (ENCSR560BUE and ENCSR703TNG). STAG1 and
- 984 SMC1A ChIP-seq data were from GEO (GSE25021 and GSE76893). CTCF motif
- orientations of each peak is labelled with black arrows in the CTCF track. Y-axis
- 986 represents signal intensity of each track.
- 987 D) CTCF-driven chromatin loops visualized using a CTCF ChIA-PET data set in
- 988 MCF7 cells (GSE72816) at the 3D Genome Browser platform. Each linkage
- 989 represents a chromatin loop.
- E) Bar graphs displaying CTCF binding events measured by ChIP-qPCR at binding
- 991 sites 1 and 5 illustrated in (C). CTCF binding fold enrichments were normalized to

992	the average of IgG binding. Each bar represents mean \pm SD of fold changes from
993	three independent experiments. Pair-wise t-test on CTCF binding fold enrichment
994	between WT and each mutant was performed. (* p<0.05, ** p<0.01)
995	F) PR binding under R5020 and progesterone treatments visualized based on a
996	reported PR ChIP-seq data set in MCF7 cells (GSE68359). Y-axis represents signal
997	intensity of each track and is adjusted to the same scale. Super enhancer range was
998	highlighted below the genomic track.
999	G) qRT-PCR measurement of <i>KRT14</i> , 16 and 17 mRNA levels in MCF7 ESR1 WT
1000	and mutant cells with PGR siRNA knockdown for 7 days. mRNA fold change
1001	normalized to WT cells; RPLP0 levels were measured as internal control. Each bar
1002	represents mean \pm SD with three biological replicates. Data shown are
1003	representative from three independent experiments. Student's t-test was used to
1004	compare the gene expression between scramble and knockdown groups. (* p <0.05,
1005	** p<0.01)
1006	H) qRT-PCR measurement of KRT5, 16 and 17 mRNA levels in MCF7 ESR1 WT
1007	and mutant cells treated with 0.1% EtOH (vehicle),100 nM P4 or 1 μM RU486
1008	treatment for 3 days. mRNA fold change normalized to WT cells; RPLP0 levels
1009	were measured as internal control. Each bar represents mean \pm SD with three
1010	biological replicates. Data shown are representative from three independent

1011 experiments. (* p<0.05, ** p<0.01)

1012 I) Representative images of immunofluorescence staining of CK5 and CK16 in

1013 MCF7 WT and ESR1 mutant cells after 3 day treatment with 1% EthOH (vehicle) or

1014 1 µM RU486. Images were taken under 20x magnification.

1015 J) Quantification of the percentages of CK positive cells in MCF7 cells. Each bar

1016 represents mean ± SD from eight different regions combining from two independent

1017 experiments. Student's t test was used to compare % BCK+ cells before and after

1018 treatment. (* p<0.05, ** p<0.01)

1019

Figure 5. Gain of basal cytokeratin expression is associated with enhanced immune activation in *ESR1* mutant tumors.

- 1022 A) Venn diagrams showing the intersection of significantly enriched hallmark
- 1023 pathways in three sets of comparisons: BCK-high vs low in 1) TCGA ER+ tumors
- 1024 (n=202 in each group), 2) METABRIC ER+ tumors (n=376 in each group) and 3)
- 1025 ESR1 mutant (n=7) vs WT (n=44) metastatic tumors. BCKs high and low were

1026 defined by the upper and bottom quartiles of each subset. The seven overlapping

- pathways are shown in a frame, and immune-related pathways are highlighted inred.
- 1029 B) Immune scores based on ESTIMATE evaluations in basal tumors (METABRIC

1030 n=328; TCGA n=190), BCK-high (METABRIC n=376; TCGA n=202) and low

- 1031 (METABRIC n=376; TCGA n=202) subsets of ER+ tumors in TCGA and
- 1032 METABRIC. Definition of BCK-high and low groups were the same in (A). Mann
- 1033 Whitney U test was used for comparison. (** p<0.01)
- 1034 C) Lymphocytes and leukocyte fractions as determined by a reported TCGA

1035 biospecimen dataset⁶⁵ comparing among basal subtype tumors (n=161), TCGA ER+

1036 BCK-high (N=163) and low (N=179) tumors. Definition of BCK-high and low groups

1037 were the same in (A). Mann Whitney U test was applied to compare the fractions

- 1038 between BCK-high and low tumors. (** p<0.01)
- 1039 D) Kaplan-Meier plots showing the disease-specific survival (DSS) (METABRIC) and
- 1040 overall survival (OS) (TCGA) comparing patients with ER+ BCKs high vs low tumors.
- 1041 BCKs high and low were defined by the upper and bottom quartiles of each subset.
- 1042 Censored patients were labelled in cross symbols. Log rank test was used and
- 1043 hazard ratio with 95% CI were labelled.
- 1044 E) Immune scores based on ESTIMATE evaluations in *ESR1* mutant (n=7) and WT
- 1045 metastatic (n=44) lesions. Mann Whitney U test was used for comparison. (* p<0.05)
- 1046 F) Dot plot showing the enrichment level alterations of immune cell subtypes in
- 1047 ESR1 mutant metastatic lesions using Davoli⁶⁶ and Tamborero⁶⁷ immune cell
- 1048 signatures. RNA seq data from intra-patient matched ESR1 mutant (N=7) and WT
- 1049 (N=44) was used. Immune cell subtypes showing significant increase in ESR1
- 1050 mutant tumors were labelled in red (p<0.05).
- 1051

1052 Figure 6. Immune activation in *ESR1* mutant tumors is associated with

1053 **S100A8/A9-TLR4** paracrine crosstalk between epithelial and stromal cells.

- 1054 A) Three-dimensional plot showing fold change (FC) expression changes of immune
- 1055 genes from ESTIMATE (N=141)⁶⁹ comparing ER+ BCK-high vs low tumors (TCGA
- and METABRIC) and intra-patient paired *ESR1* WT/mutant tumors. Consistently
- 1057 increased/decreased genes in TCGA and METABRIC BCK-high tumors and ESR1

1058 mutant tumors were highlighted in red and blue. Inconsistently changed genes

- among the three comparisons are labelled in black.
- 1060 B) ER+ cases with BCK-high and low quantiles were further divided by the mean
- 1061 expression of S100A8 and S100A9. ESTIMATE immune scores were compared
- across all four subsets (n=188 and 101 in each group of METABRIC and TCGA)
- 1063 together with basal tumors (n=328 METABRIC and n=190 TCGA). Each
- 1064 corresponding comparison was tested using Mann Whitney U test. (**p<0.01)
- 1065 C) Graphical presentation of strategy to quantify and compare S100A8/9
- 1066 heterodimer abundance in plasma from patients with ER+ metastatic breast cancer.
- 1067 D) Box plot showing S100A8/9 heterodimer concentrations in plasma from patients
- 1068 with ESR1 WT (n=7) and mutant (n=11) metastatic breast cancer. Mann Whitney U
- 1069 test was utilized. (* p<0.05)
- 1070 E) Comparison of TLR4 (left) and RAGE (right) signaling signature enrichments in
- 1071 intra-patient matched ESR1 mutant (N=7) and WT (N=44) cohort. Delta GSVA score

1072 of each sample was calculated by subtracting the scores of primary tumors from the

- 1073 matched metastatic tumors. Mann-Whitney U test was performed between WT and
- 1074 mutant tumors. (**p<0.01)
- 1075 F) Violin plots showing S100A8, S100A9, TLR4 and AGER expression by log2
- 1076 normalized counts in different cell subtypes using single-cell RNA-seq data from two
- 1077 bone metastases from a patient with ER+ breast cancer.
- 1078 G) Percent of cells expressing S100A8, S100A9, TLR4 and AGER, using single cell
- 1079 RNA seq data shown in Figure 6F.

1080

Figure 7. Graphical presentation of proposed mechanisms and relevance of

1082 basal cytokeratin induction in *ESR1* mutant breast cancer.

- 1083 ESR1 WT cells exhibit low basal cytokeratin expression with baseline TAD
- 1084 prevalence spanning KRT14/16/17 loci. In contrast, a minor subpopulation of ESR1
- 1085 mutant cells exhibit strong basal cytokeratin expression, due to PR activated
- 1086 enhancer at the KRT14/16/17 gene locus-spanning TAD. Increased expression of
- 1087 basal cytokeratin is associated with immune activation in ESR1 mutant tumor similar
- 1088 to that seen in basal tumors, at least in part mediated via enhanced S100A8/A9-
- 1089 TLR4 paracrine crosstalk between epithelial and stromal cells, including
- 1090 macrophages.

Α

Basal /Luminal Gene Sets Used in This Study

Pair #	Pair Name	Luminal Genes	Basal Genes
1	Charafe- Jauffret ³¹	380	455
2	Huper ³²	58	53
3	Sorlie 33-36	68	75
4	TCGA 37	139	164



Β

KRT16

KRT6A

PI3

S100A2

С

412

43

0

23

Α

Basal Gene Sets

127

0

11

3

Β

D

34

AKR1C1

ESR1

FOXA1 HMGCS2 С

335

MLPH

55

15

22

Luminal Gene Sets

2

4

136

0

0

0

0

Β

D

53

Figure 1. Basal breast cancer gene signatures are enriched in *ESR1* mutant breast cancer



Figure 2. Basal cytokeratins are the leading enriched basal markers in *ESR1* mutant breast cancer cells and tumors



С



D qRT-PCR with ESR1 Knockdown





Figure 3. Basal cytokeratins are not E2 induced but negatively correlated with

Surrounding CK- Cells





Figure 4. Basal cytokeratins are induced via a unique PR enhancer associated TAD in *ESR1* mutant cells



D Kaplan Meier Plots by Basal Cytokeratin in ER+ Tumors





Figure 5. Gain of basal cytokeratins is associated with enhanced immune activation in *ESR1* mutant tumors



Figure 6. Immune activation in *ESR1* mutant tumors is associated with S100A8/A9-TLR4 paracrine corsstalk between epithelial and stromal cells.



Figure 7. Schema of proposed mechanisms of basal cytokeratin induction in *ESR1* mutant breast cancer.