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Transcriptomic profiling of sequential tumours from breast cancer patients provides a global view of metastatic expression changes following endocrine therapy

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TRANSLATIONAL RELEVANCE:

To date, next generation sequencing has been completed for a substantial number of primary breast tumours and provides an in depth understanding of the biology and heterogeneity of this disease. Here primary tumours together with matched nodal and liver metastatic patient tumours have been sequenced, offering the first in depth view of the transcriptomic changes occurring during disease progression. Patient heterogeneity was strongly evident in both primary and metastatic tumours. In addition, common adaptations in cell-cell communication were detected as metastatic endocrine-resistant disease took hold. These findings emphasise the need for ongoing personalised medicine throughout disease management. In particular, it is important to re-evaluate the biology of metastatic disease, following treatment intervention, in order to monitor adaptations and identify the optimal treatment strategy.

ABSTRACT

Purpose Disease recurrence is a common problem in breast cancer and yet the mechanisms enabling tumour cells to evade therapy and colonise distant organs remain unclear. We sought to characterise global expression changes occurring with metastatic disease progression in the endocrine resistant setting.

Experimental Design Here, for the first time, RNAsequencing has been performed on matched primary, nodal and liver metastatic tumours from tamoxifen-treated patients following disease progression. Expression of genes commonly elevated in the metastases of sequenced patients was subsequently examined in an extended matched patient cohort with metastatic disease from multiple sites. The impact of tamoxifen treatment on endocrine resistant tumours *in vivo* was investigated in a xenograft model.

Results The extent of patient heterogeneity at the gene level was striking. Less than 3% of the genes differentially expressed between sequential tumours were common to all patients. Larger divergence was observed between primary and liver tumours than between primary and nodal tumours, reflecting both the latency to disease progression and the genetic impact of intervening therapy. Furthermore, an endocrine-resistant *in vivo* mouse model demonstrated that tamoxifen treatment has the potential to drive disease progression and establish distant metastatic disease. Common functional pathways altered during metastatic, endocrine-resistant progression included extracellular matrix receptor interactions and focal adhesions.

Conclusions This novel global analysis highlights the influence of primary tumour biology in determining the transcriptomic profile of metastatic tumours, as well as the need for adaptations in cell-cell communications to facilitate successful tumour cell colonisation of distant host organs.

INTRODUCTION

The vast majority of breast cancers are steroid receptor positive and depend on estrogen for growth. The use of estrogen receptor modulators, such as tamoxifen and estrogen depletion strategies, including aromatase inhibitors, have been widely successful. Disease recurrence however is common, with between 10 and 25% of patients developing extensive local or distant metastases. Most patients with metastases will succumb to their disease, as targeted therapies which provide significant clinical benefit remain elusive.

Metastatic breast cancer secondary to endocrine treatment is heterogeneous, ranging from discreet lesions to diffuse multi-organ disease. The dominant site of metastatic involvement can have a significant bearing on clinical outcome. Metastatic disease of the liver has been reported to be a predictor of poor outcome, with a median survival of 18 months (1), whereas metastatic burden confined to the bone is thought to be more indolent (2).

Colonisation of the host organ is the ultimate step of a progressive disease path. Most breast cancer cells that enter the circulation and infiltrate distant organs die due to the hostile nature of the host microenvironment (3). For metastatic growth to be established, disseminated breast cancer cells need to survive a period of latency, prolonged exposure to therapy and subsequently re-initiate growth when appropriate. Evidence of the impact of steroid treatment on tumour progression is emerging. Steroid receptor switching and in particular loss of the progesterone receptor (PgR), an estrogen receptor (ER) target, is observed in approximately 30% of patients (4, 5). Furthermore, gain of function mutations in the *ESR1* gene have been identified in metastatic lesions from endocrine treated patients (6, 7).

The mechanisms required to enable cancer cells to overcome hostile forces and re-initiate growth at distant organs in endocrine treated patients have not been fully elucidated. Lessons from *in vitro* studies and murine models have provided important information regarding discrete tools employed by breast cancer cells to prosper at distant metastatic sites. For example, the CXCR4/SDF1 pairing

has been identified as a mediator of metastatic cell survival in the bone (8, 9); transcriptional inhibitors of differentiation, Id1 and Id3, have been associated with lung metastasis (10, 11); and serpins, as metastatic functionaries in the brain (12). However, to date, no global overview of gene expression alterations which are essential for successful colonisation of cells at metastatic sites in breast cancer patients has been undertaken.

To clearly define mediators of colonisation in endocrine treated patients we undertook gene expression analysis from sequential tumour samples. RNAseq analysis was performed on matched primary, nodal and distant metastatic tumours from breast cancer patients, all of whom had developed a liver metastasis following tamoxifen treatment. At a functional level, studies in xenograft models of endocrine resistant tumours were undertaken to examine the role of tamoxifen in the development of metastatic disease secondary to therapy resistance. Over represented pathways identified by the RNAseq analysis were further studied at the transcript and protein level in an expanded cohort of patients with metastases of multiple distant organs.

MATERIALS AND METHODS

Gene expression and bioinformatic analysis

Following ethical approval, formalin fixed paraffin embedded (FFPE) tumour samples were macrodissected to select regions with >70% tumour cells. RNA isolation was performed with High Pure FFPE RNA Micro Kit (Roche) and subjected to duplex specific thermostable nuclease (DSN) treatment. RNA Sequencing was performed by BGI (Hong Kong) using the Illumina HiSeq 2000 with >40million reads. Insufficient RNA was obtained from the nodal sample of patient one for sequencing. Reads were aligned to the hg19 genome using two different types of analysis: one for the differential expression and the other for use with the TCGA data. For the differential expression analysis, the reads were aligned using TopHat and genes were quantified using HTseq. For usage with TCGA data, the data was aligned using MapSplice and the genes were quantified using RSEM.

Full details of bioinformatics analysis are provided in Supplementary Methods. Data from this study has been deposited in the NCBI Gene Expression Omnibus (GEO) (<http://www.ncbi.nlm.nih.gov/geo/>) under accession number GSE58708.

cDNA was synthesised as previously described (13). Gene expression was assessed using Taqman technology (Applied Biosystems) on the ABI PRISM 7500 platform with the following probes: EPPK1 Hs01104050_s1; COL4A2 Hs01098873_m1; COL4A1 Hs00266237_m1; SPDEF Hs01026050_m1; KRT19 Hs00761767_s1; AHDC1 Hs00210424_m1; NCOR2 Hs00196955_m1; AKAP8L Hs00205106_m1. EIF2B Hs00426752_m1 was used as an internal control. The comparative C_T ($\Delta\Delta C_T$) method was applied to analyse relative mRNA expression levels.

Immunohistochemistry

5µm tumour sections were deparaffinised and rehydrated as previously described (4). Antigen retrieval was performed with sodium citrate at pH6 or EDTA at pH9 (Ki67). Dako Envision+ HRP kit was used as per manufacturer's instructions. Primary antibodies incubated for 1 hour at room temperature were: Ki67 (1:200, MIB-1, M7240, Dako); KRT19 (1:200, HPA002465, Sigma); LMNB2 (1:75, MAB3536, Merck Millipore); KIF12 (1:100, NBPI-86029, Novus); EVPL (1:30, Ab49397, Abcam); MCM4 (1:50, HPA004873, Sigma); p-mTOR (1:100, Ser2448, #2976, Cell Signalling); and active β -catenin (1:150, 05-665, Merck Millipore). The following primary antibodies were incubated overnight at 4°C: p-p70S6K (1:100, Thr389, #9205, Cell Signalling); p-HER2 (1:400, Tyr1248, 06-229, Merck Millipore); p-STAT3 (1:100, Tyr705, #9131, Cell Signalling); and p-ERK 1/2 (1:400, Thr202/Tyr204, #4370, Cell Signalling). ER (1:60, RM9101S, Thermo Fisher Scientific), PgR (1:60, H3569, DAKO) and HER2 (neat, 4B5, 790-2991, Ventana) were stained on an automated program. H-score was assigned by two independent assessors on a scale of 0-300 with 300 representing the highest intensity staining in 100% of cells.

Xenograft Studies

All mouse experiments were performed in accordance with the European Communities Council Directive (86/609/EEC) and were reviewed and approved by Research Ethics Committee under license from the Department of Health. Ten 6-week old female BALB/c SCID mice (Harlan) were implanted with 17-B-estradiol pellets (0.36 mg/pellet, 60-day release) alone or in combination with 4-hydroxytamoxifen pellets (5 mg/pellet, 60-day release). All pellets were obtained from Innovative Research of America and were readministered at week 8. Luciferase-tagged LY2 cells (1×10^6 cells) were mixed with 50% Matrigel (BD biosciences) and implanted into the 4th inguinal mammary gland by injection. Tumours were imaged using an IVIS whole body imaging system (Xenogen Corp) to detect luciferase activity. Five weeks after cell implantation, primary tumours were surgically removed, formalin fixed and paraffin embedded. The mice were monitored by IVIS once per week for the next 10 weeks to detect metastasis of the labelled cells. The fold increase in metastatic flux was compared by t-test between each group of animals.

RESULTS

Gene expression profiles of breast cancer patients reflect the histopathology of the original primary tumour

RNA was extracted from matched breast cancer primary, node and liver metastatic tumour tissue from three tamoxifen treated luminal breast cancer patients. The patients represented the spectrum of the luminal classification, from luminal A (patient one) to luminal B2 (patient three) (Table 1). Despite the diverse pathology of each of the primary breast tumours (Table 1, Fig. 1A and Supplementary Fig. S1), positive steroid receptor status qualified each of the patients for adjuvant endocrine treatment.

Correspondence analysis of the sequenced RNA from the individual tissue samples revealed that each of the sequential patient samples clustered together, rather than with the metastatic site (Fig 1B). The relative divergence in gene expression from the primary to the metastatic tumours was reflective of the disease latency period. Patient one with a recorded metastasis-free survival period of 11 years had the greatest difference in gene expression profile between the primary and metastatic tumours (Fig. 1B). Patient two and patient three had shorter disease-free periods and correspondingly substantially less alterations in gene expression, suggesting that some of the genes necessary for successful progression are present in the primary tumour.

RNAseq data from the breast cancer sequential samples was analysed with breast cancer samples from The Cancer Genome Atlas by correspondence analysis of almost 16,000 genes (Fig. 1C). Nodal and metastatic samples clustered in amongst the primary tumours. PAM50 genes were used to classify the breast cancer sequential samples and the TCGA samples into the major breast cancer subtypes. Tumours from patients one and two clustered on the luminal end of the horizontal axis. The tumours from patient three straddled the luminal/HER2 border. Interestingly, while the primary tumour from patient one was luminal A, the metastatic tumour from this patient was classified as luminal B

Metastatic gene expression alterations are heterogeneous in endocrine treated breast cancer patients

To define the changes involved in disease progression, differentially expressed genes (DEGs) were identified between matched primary and liver metastatic tumours from each patient. This analysis identified 1,268 upregulated and 2,606 downregulated genes between matched primary and metastatic tumours. Remarkably, less than 3% of DEGs were common to all three patients highlighting the extent of patient heterogeneity (Fig. 2A). Only 31 upregulated and 69 downregulated genes were shared by the three patients (Supplementary Tables S1 and S2). The individuality of each patient is evident in the heat maps of DEGs where each patient elevated their

own unique set of genes in the liver metastatic sample (Fig. 2B). Of note, little difference in gene expression was detected between the primary tumour and the nodal metastases, both of which were treatment naïve and were resected at the time of initial surgery (Fig. 2B).

Analysis of enriched KEGG pathway terms was performed for the DEGs of each patient and highlighted a number of important pathways that were unique to each patient (Supplementary Fig. S2 and S3). The luminal A tumour from patient one was highly endocrine dependent with strong expression of both ER and PgR. The large divergence from primary to metastasis on correspondence analysis as well as the large number of differentially expressed genes is indicative of the multiple adaptations that were required for successful tumour progression and colonisation. Following endocrine treatment, steroid receptor status remained strong, however alterations in growth factor signalling networks, including MAPkinase, Jak-STAT and ErbB2 signalling pathways were observed (Supplementary Fig. S2). As protein function cannot always be directly inferred from gene expression analysis, expression of phosphoproteins from these pathways was examined and confirmed in matched primary and metastatic tumours from a luminal A patient (Supplementary Fig. S4). In contrast to patient one, the luminal B1 primary tumour from patient two displayed limited endocrine dependence, with low steroid receptor expression and no HER2. Limited divergence on correspondence analysis and relatively low number of gene alterations suggest that the genes required for metastatic progression may have been present in the primary tumour. Pathways that were altered in the liver metastatic tumour include DNA replication and cell cycle (Supplementary Fig. S2), also confirmed at the protein level (MCM4) (Supplementary Fig. S4). The luminal B2 tumour from patient three was strongly positive for ER, PgR and HER2. From correspondence analysis, genes from the primary, node and liver metastasis clustered with the HER2 subtype using the PAM50 classification. Alterations in the WNT signalling pathway observed in the metastasis, at both the gene and phosphoprotein level (Supplementary Fig. S2 and S4), signify tumour adaptation to endocrine therapy and activation of the epithelial to mesenchymal transition network.

Of note, the mTOR pathway which is a known predictor of endocrine resistance (14), was found to be elevated in the liver metastases of patients one and three at the transcript level in comparison to the primary tumour (Supplementary Fig. S2). Expression levels of phospho-mTOR and phospho-p70S6 kinase in matched primary and metastatic tumours from endocrine treated patients confirmed activation of the mTOR pathway (Supplementary Fig. S5).

Cell to cell communication is a common pathway in liver metastasis

Although at the gene level, patient heterogeneity was very apparent in the endocrine resistant metastatic tumours, a number of functional pathways were common to all three patients (Fig. 2C). Alterations in several cancer pathways including small cell lung carcinoma, renal cell carcinoma and pathways in cancer were observed. Of interest, pathways important in cell to cell interaction were represented in gene sets that were both enhanced and suppressed in the metastatic liver. Alterations in genes important in ECM receptor interactions and focal adhesion indicate the importance of communication between the tumour and the host liver for successful colonisation.

Endocrine treatment contributes to the development of metastasis secondary to endocrine resistance

Of the sequential tumour samples sequenced here, primary and nodal tumours were biopsied at the same time. Liver tumours were biopsied following a latency period and also following treatment intervention. To determine the impact of endocrine treatment on the development of distant metastasis in endocrine resistant tumours we employed a xenograft model. Endocrine resistant cells were implanted into the mammary fat pad of immune-compromised mice in the presence of estrogen (Fig. 3A). Once established, the primary tumours were resected. Local disease recurrence occurred in animals treated and untreated with the SERM tamoxifen. Distant metastasis however was only established in the tamoxifen treated group (Fig. 3A). Metastatic disease was observed in

the bone, lung and liver (Fig. 3B). Expression of the proliferation marker, Ki67 and keratin 19, a marker of disseminated cancer cells, which was elevated in all of the patient liver metastases, was found to be enhanced in the xenograft liver metastatic tissue compared with the primary (Fig. 3B). These data demonstrate that tamoxifen has the potential to contribute to the development of metastatic disease progression in endocrine resistant tumours. Analysis of the patient RNAseq data revealed that the majority of the identified upregulated DEGs were in fact elevated in metastatic tumours relative to both matched primary and nodal tumours (Fig. 3C). These data suggest that for the majority of these genes, expression did not increase when the cells first moved away from the primary site to the node but did increase following treatment with tamoxifen, the latency period and intravasation of the distant organ. Remarkably, of the common DEGs displaying this pattern of expression (Supplementary Table S3), almost half have defined roles in cell to cell communication and cell structure (Supplementary Fig. S6).

Cellular communication and cell structure proteins are evident at other metastatic sites

An extended cohort of matched primary and metastatic patient samples were employed to examine the RNA and protein expression of shared genes from the RNAseq data. The patient cohort included a range of metastatic sites including the liver, lung, bone and peritoneum of endocrine treated patients (Supplementary Table S4). Real-time PCR confirmed the elevated expression of genes in liver metastases relative to matched primary tumours (Fig. 3D). Increases in gene expression were also detected at other metastatic sites (Fig. 3D).

At the protein level, immunohistochemical analysis of matched primary and metastatic tumours revealed strong keratin 19 protein expression, consistent with that observed in the metastatic tumours from the tamoxifen treated xenografts (Fig. 3E). Protein expression of the cellular communication proteins, lamin B2, Kif12 and envoplakin were confirmed in the metastatic tumours

of the endocrine treated patients. Greater expression of these proteins was found in the metastatic tumour cells in comparison to the surrounding normal host tissue (Fig. 3E). Furthermore analysis of unmatched metastatic brain datasets revealed enhanced expression of our defined 31 gene set in the metastatic tumours compared with normal brain tissue (Supplementary Fig. S7). Taken together these data support a role for cellular communication in breast cancer colonisation of distant organs following treatment resistance.

DISCUSSION

The ability of selected tumours to adapt and evade endocrine therapy is well established. Enhanced expression of plasticity networks (4, 15, 16), as well as mediators of EMT (17, 18), in the primary tumour has been associated with treatment failure. Information regarding changes which occur with tumour progression on treatment however is not readily available. Here we describe the first global transcriptomic analysis of sequential primary and metastatic tumour samples from endocrine resistant breast cancer patients.

Several initiating events of tumour adaptability have been proposed, including the degree of intra-tumour heterogeneity (19, 20) as well as the presence of stem cell populations (21). Though the source of the treatment adaptability is not addressed in this study, the resultant mediators which enable the successful colonisation of the tumour at a distant site are elucidated.

Here global mapping of gene expression alterations which occur in metastatic breast cancers revealed a significant degree of patient heterogeneity. The relative divergence in gene expression reflected the histology of the original tumour and the disease latency period. While disease progression in some cancers is rapid (22, 23), in ER positive breast cancer, metastasis may occur decades after resection of the primary tumour (24). The greatest degree of divergence in gene expression was observed in the ER positive PgR positive luminal A patient (patient one) with a metastasis-free survival period of >10 years. Differences in gene expression suggest that the primary

tumour underwent significant alterations over an extended period of time to successfully develop liver metastasis. In contrast, the low ER positive, PgR/HER2 negative tumour (patient two) displayed relatively little alteration in gene expression between the primary and the metastatic tumours. These data suggest that the primary tumour from patient two may harbour metastatic progression genes (8) which could have altered functions at the primary and metastatic sites. The extent of divergence in gene expression between primary and metastatic tumours would therefore appear to be inversely related to the aggressiveness of the primary tumour.

Signal pathways enriched at the metastatic site were also patient specific and closely related to the biology of the primary tumour. The ER/PgR/HER2 positive patient (patient three) displayed elevations in plasticity networks, including the WNT signalling pathways, whereas the ER/PgR positive, HER2 negative patient (patient one) had enhanced growth factor signalling. Increases in de-differentiation signals and in particular induction of steroid receptor/growth factor receptor cross talk have been well described as mechanisms of tumour adaptability to endocrine treatment (16, 25, 26). Furthermore elevations in key components of the mTOR pathway were observed in both patient one and patient three in metastatic liver tissue. Activation of the mTOR pathway has previously been associated in endocrine resistance (14, 27) and elevated expression of p-mTOR, p-4EBP1 and p-p70S6K has been reported in metastatic tumours in comparison to matched primary tissue (28).

The impact of sequential endocrine therapy in the treatment of endocrine resistant disease in patients however remains to be fully clarified. Though several clinical studies suggest that some metastatic patients respond to continued endocrine treatment, a subset of patients receives little or no benefit (29-31). Here using endocrine resistant xenografts to model sequential treatment strategies we found that the development of distant metastasis was dependent on continued tamoxifen treatment. Though the molecular mechanisms driving this continued resistance have not been described to date, *ex vivo* studies demonstrating gain of function mutations in breast metastatic patient tumours following endocrine treatment suggest that ER may remain an important

player (7). In this study the majority of DEGs were elevated in the metastatic tumours relative to both the primary and the node, with little alteration observed between the primary and the matched nodal tissue. The gene expression changes that occur during the latency period, under the pressure of treatment are therefore those that are important for the establishment of distant metastatic disease.

Though several elegant studies have employed *in vitro* and *in vivo* models to determine key players in breast cancer colonisation of distant organs, including the bone, lung and brain (8-12), there is little information regarding signalling networks important in metastatic liver disease. In this study common pathways elevated in liver metastatic tumours included pathways in cancer, ECM receptor interactions and focal adhesions. Extracellular matrix proteins have previously been shown to be important in colon cancer liver metastasis (32), moreover enhanced Claudin 2-extracellular matrix interactions have been demonstrated in liver-aggressive breast cancer *in vivo* models (33). The ability of breast tumour cells to communicate with and impact host cells would therefore appear to be essential to its ability to colonise and survive at a distant site.

Elucidation of differential gene expression in metastatic tumours is a key step to understanding the complex mechanisms controlling tumour adaptability. Here we define gene expression alterations which occur in endocrine treated metastatic liver tumours. These global studies build on model systems and clinical observations, which along with further metastatic sequencing studies will enable rational personalised treatment strategies to be developed for endocrine resistant breast cancer patients.

AUTHORS' CONTRIBUTION

Conception and design: D.M., M.M., L.Y.

Development of methodology: J.M., A.F., D.M., C.B.

Acquisition of data: J.M., A.F., D.M., F.B., D.V., S.C., C.B., J.B., L.H., P.T.

Analysis and interpretation of data: J.M., A.F., P.O., L.Y.

Writing, review, and/or revision of the manuscript: J.M., A.F., L.Y.

Study supervision: P.O., A.H., L.Y.

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REFERENCES

1. Nicolini A, Giardino R, Carpi A, Ferrari P, Anselmi L, Colosimo S, et al. Metastatic breast cancer: an updating. *Biomed Pharmacother.* 2006;60:548-56.
2. Sherry MM, Greco FA, Johnson DH, Hainsworth JD. Metastatic breast cancer confined to the skeletal system. An indolent disease. *Am J Med.* 1986;81:381-6.
3. Luzzi KJ, MacDonald IC, Schmidt EE, Kerkvliet N, Morris VL, Chambers AF, et al. Multistep nature of metastatic inefficiency: dormancy of solitary cells after successful extravasation and limited survival of early micrometastases. *Am J Pathol.* 1998;153:865-73.
4. McBryan J, Theissen SM, Byrne C, Hughes E, Cocchiglia S, Sande S, et al. Metastatic progression with resistance to aromatase inhibitors is driven by the steroid receptor coactivator SRC-1. *Cancer Res.* 2012;72:548-59.
5. Broom RJ, Tang PA, Simmons C, Bordeleau L, Mulligan AM, O'Malley FP, et al. Changes in estrogen receptor, progesterone receptor and Her-2/neu status with time: discordance rates between primary and metastatic breast cancer. *Anticancer Res.* 2009;29:1557-62.
6. Robinson DR, Wu YM, Vats P, Su F, Lonigro RJ, Cao X, et al. Activating ESR1 mutations in hormone-resistant metastatic breast cancer. *Nat Genet.* 2013;45:1446-51.
7. Jeselsohn R, Yelensky R, Buchwalter G, Frampton G, Meric-Bernstam F, Gonzalez-Angulo AM, et al. Emergence of constitutively active estrogen receptor- α mutations in pretreated advanced estrogen receptor-positive breast cancer. *Clin Cancer Res.* 2014;20:1757-67.
8. Nguyen DX, Bos PD, Massague J. Metastasis: from dissemination to organ-specific colonization. *Nat Rev Cancer.* 2009;9:274-84.
9. Kang Y, Siegel PM, Shu W, Drobnjak M, Kakonen SM, Cordon-Cardo C, et al. A multigenic program mediating breast cancer metastasis to bone. *Cancer Cell.* 2003;3:537-49.
10. Gupta GP, Perk J, Acharyya S, de Candia P, Mittal V, Todorova-Manova K, et al. ID genes mediate tumor reinitiation during breast cancer lung metastasis. *Proc Natl Acad Sci U S A.* 2007;104:19506-11.
11. Swarbrick A, Roy E, Allen T, Bishop JM. Id1 cooperates with oncogenic Ras to induce metastatic mammary carcinoma by subversion of the cellular senescence response. *Proc Natl Acad Sci U S A.* 2008;105:5402-7.
12. Valiente M, Obenaus AC, Jin X, Chen Q, Zhang XH, Lee DJ, et al. Serpins promote cancer cell survival and vascular co-option in brain metastasis. *Cell.* 2014;156:1002-16.

13. O'Hara J, Vareslija D, McBryan J, Bane F, Tibbitts P, Byrne C, et al. AIB1:ERalpha transcriptional activity is selectively enhanced in aromatase inhibitor-resistant breast cancer cells. *Clin Cancer Res.* 2012;18:3305-15.
14. Beelen K, Opdam M, Severson TM, Koornstra RH, Vincent AD, Wesseling J, et al. Phosphorylated p-70S6K predicts tamoxifen resistance in postmenopausal breast cancer patients randomized between adjuvant tamoxifen versus no systemic treatment. *Breast Cancer Res.* 2014;16:R6.
15. Pathiraja TN, Nayak SR, Xi Y, Jiang S, Garee JP, Edwards DP, et al. Epigenetic reprogramming of HOXC10 in endocrine-resistant breast cancer. *Sci Transl Med.* 2014;6:229ra41.
16. McIlroy M, McCartan D, Early S, P OG, Pennington S, Hill AD, et al. Interaction of developmental transcription factor HOXC11 with steroid receptor coactivator SRC-1 mediates resistance to endocrine therapy in breast cancer [corrected]. *Cancer Res.* 2010;70:1585-94.
17. Loh YN, Hedditch EL, Baker LA, Jary E, Ward RL, Ford CE. The Wnt signalling pathway is upregulated in an in vitro model of acquired tamoxifen resistant breast cancer. *BMC Cancer.* 2013;13:174.
18. Manavalan TT, Teng Y, Litchfield LM, Muluhngwi P, Al-Rayyan N, Klinge CM. Reduced expression of miR-200 family members contributes to antiestrogen resistance in LY2 human breast cancer cells. *PLoS One.* 2013;8:e62334.
19. Lawrence MS, Stojanov P, Polak P, Kryukov GV, Cibulskis K, Sivachenko A, et al. Mutational heterogeneity in cancer and the search for new cancer-associated genes. *Nature.* 2013;499:214-8.
20. Kleppe M, Levine RL. Tumor heterogeneity confounds and illuminates: assessing the implications. *Nat Med.* 2014;20:342-4.
21. Meacham CE, Morrison SJ. Tumour heterogeneity and cancer cell plasticity. *Nature.* 2013;501:328-37.
22. Janne PA, Freidlin B, Saxman S, Johnson DH, Livingston RB, Shepherd FA, et al. Twenty-five years of clinical research for patients with limited-stage small cell lung carcinoma in North America. *Cancer.* 2002;95:1528-38.
23. Briele HA, Das Gupta TK. Natural history of cutaneous malignant melanoma. *World J Surg.* 1979;3:255-70.
24. Lee YT. Patterns of metastasis and natural courses of breast carcinoma. *Cancer Metastasis Rev.* 1985;4:153-72.
25. Arpino G, Wiechmann L, Osborne CK, Schiff R. Crosstalk between the estrogen receptor and the HER tyrosine kinase receptor family: molecular mechanism and clinical implications for endocrine therapy resistance. *Endocr Rev.* 2008;29:217-33.
26. Montemurro F, Di Cosimo S, Arpino G. Human epidermal growth factor receptor 2 (HER2)-positive and hormone receptor-positive breast cancer: new insights into molecular interactions and clinical implications. *Ann Oncol.* 2013;24:2715-24.
27. Tokunaga E, Hisamatsu Y, Tanaka K, Yamashita N, Saeki H, Oki E, et al. Molecular mechanisms regulating the hormone sensitivity of breast cancer. *Cancer Sci.* 2014;105:1377-83.
28. Beelen K, Hoefnagel LD, Opdam M, Wesseling J, Sanders J, Vincent AD, et al. PI3K/AKT/mTOR pathway activation in primary and corresponding metastatic breast tumors after adjuvant endocrine therapy. *Int J Cancer.* 2014;135:1257-63.
29. Barrios C, Forbes JF, Jonat W, Conte P, Gradishar W, Buzdar A, et al. The sequential use of endocrine treatment for advanced breast cancer: where are we? *Ann Oncol.* 2012;23:1378-86.
30. Goss PE, Ingle JN, Pater JL, Martino S, Robert NJ, Muss HB, et al. Late extended adjuvant treatment with letrozole improves outcome in women with early-stage breast cancer who complete 5 years of tamoxifen. *J Clin Oncol.* 2008;26:1948-55.
31. Mauri D, Pavlidis N, Polyzos NP, Ioannidis JP. Survival with aromatase inhibitors and inactivators versus standard hormonal therapy in advanced breast cancer: meta-analysis. *J Natl Cancer Inst.* 2006;98:1285-91.

32. Van den Eynden GG, Majeed AW, Illemann M, Vermeulen PB, Bird NC, Hoyer-Hansen G, et al. The multifaceted role of the microenvironment in liver metastasis: biology and clinical implications. *Cancer Res.* 2013;73:2031-43.
33. Tabaries S, Dong Z, Annis MG, Omeroglu A, Pepin F, Ouellet V, et al. Claudin-2 is selectively enriched in and promotes the formation of breast cancer liver metastases through engagement of integrin complexes. *Oncogene.* 2011;30:1318-28.

FIGURE LEGENDS

Figure 1. Molecular profiling of matched tumour samples from three heterogeneous breast cancer patients. **A**, Immunohistochemical staining of classic biomarkers in primary breast tumour tissue (top) and H&E staining of matched liver metastatic tumour tissue (bottom) showing strong heterogeneity between the 3 patients. ER, estrogen receptor; PgR progesterone receptor; H&E, haematoxylin and eosin staining. **B**, Correspondence analysis of RNAseq data from primary (P), nodal (N) and metastatic liver (M) tumours of the three patients. Patient heterogeneity is evident and tumours do not cluster based on site of origin. Patient 1 shows the largest divergence between primary and metastatic tumours. **C**, Correspondence analysis of RNAseq data from the primary, nodal and metastatic samples from the three patients (n=8) and breast cancer patient samples from The Cancer Genome Atlas (n=840). Each dot or shape represents an individual tumour, coloured based on PAM50 profiling.

Figure 2. Matched primary and metastatic tumours display altered expression of genes involved in cell-cell interactions and cancer pathways. **A**, Venn diagrams showing DEGs for each individual patient between their primary and metastatic tumours. Left panel: genes upregulated in metastatic tumour relative to matched primary tumour. For all of these genes, expression in a normal liver sample was less than the primary tumour. Right panel: genes downregulated in metastatic tumour relative to matched primary tumour. DEGs exhibited a fold change >1.5 and >50 counts per million. **B**, Heat maps displaying the 1268 upregulated (left panel) and 2606 downregulated (right panel) DEGs from part A. Each patient displayed a distinct pattern of gene expression as highlighted. **C**, Network maps showing enriched KEGG pathways from the upregulated (left panel) and downregulated (right panel) DEGs. Details of pathways unique to each patient are provided in Supplementary Figs S2 and S3. Enriched KEGG pathways common to all 3 patients include a number

of cancer pathways and also cell interaction pathways such as focal adhesions and ECM receptor interactions.

Figure 3. Tamoxifen contributes to the development of metastasis secondary to endocrine resistance. **A**, Representative *in vivo* bioluminescence images of mice following orthotopic injection of luciferase expressing endocrine resistant breast cancer cells. Five mice were treated with estrogen pellets (E2 only) and five mice had estrogen and 4-hydroxytamoxifen pellets (E2+4-OHT). Primary tumours were resected at week 5. Local recurrences developed in both groups but distant metastases were detectable only in tamoxifen treated mice. Graph displays the fold increase in metastatic flux with each line representing an individual mouse. *p-value<0.05 by unpaired t-test at week 15. **B**, *Ex vivo* bioluminescence imaging of organs from the mouse xenograft model. All estrogen plus tamoxifen treated mice displayed bioluminescence in at least one organ of their bone, lung and livers (n=5). No bioluminescence was detected in the same organs from estrogen-only treated animals (n=5). Scatter plots show total flux from each *ex vivo* organ. Representative images of immunohistochemical staining of matched primary and metastatic tumours from estrogen plus tamoxifen treated mouse xenografts are shown. The metastatic marker Keratin 19 was more strongly expressed in metastatic tumours than matched primary tumours (n=3). Ki67 staining is also shown to help identify tumour cells. **C**, Heat map listing the common 31 DEGs which are upregulated in liver metastatic tumours relative to matched primary tumours for all three patients. The majority of these genes are more strongly expressed in the liver metastasis, sampled after tamoxifen treatment, than in the nodal tumour, sampled before tamoxifen. Those genes with a fold change >1.5 fold between liver and node are highlighted in bold (n=21 genes). **D**, Heatmap of TaqMan gene expression fold changes of genes up regulated in metastasis from the RNA-seq data analysis in an extended cohort of matched primary breast cancer and metastases samples. To the left of the TaqMan expression is the original gene expression fold changes for the RNA-seq data. Metastatic sites of the extended cohort include liver, peritoneum (Perit.), contralateral breast ("Contra."), bone and axilla ("Ax."). The liver metastases patients in the extended cohort exhibited a similar gene

expression pattern to the RNA-seq data. **E**, Immunohistochemical staining of four DEGs in matched tumours of 7 independent endocrine-treated patients. H-scores for each sample are reported as a heat map. Details of patient samples and treatments are provided in Supplementary Table S4.

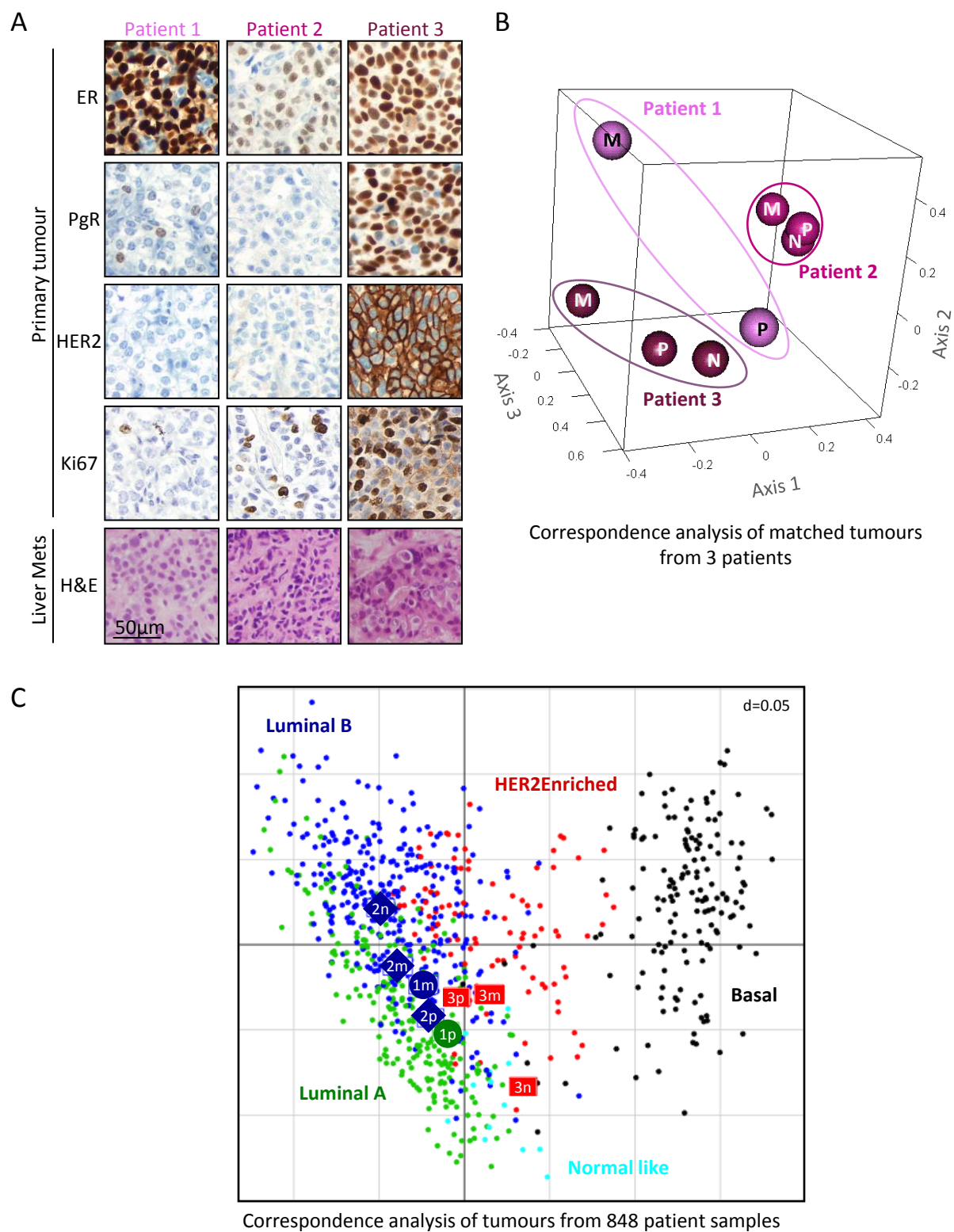
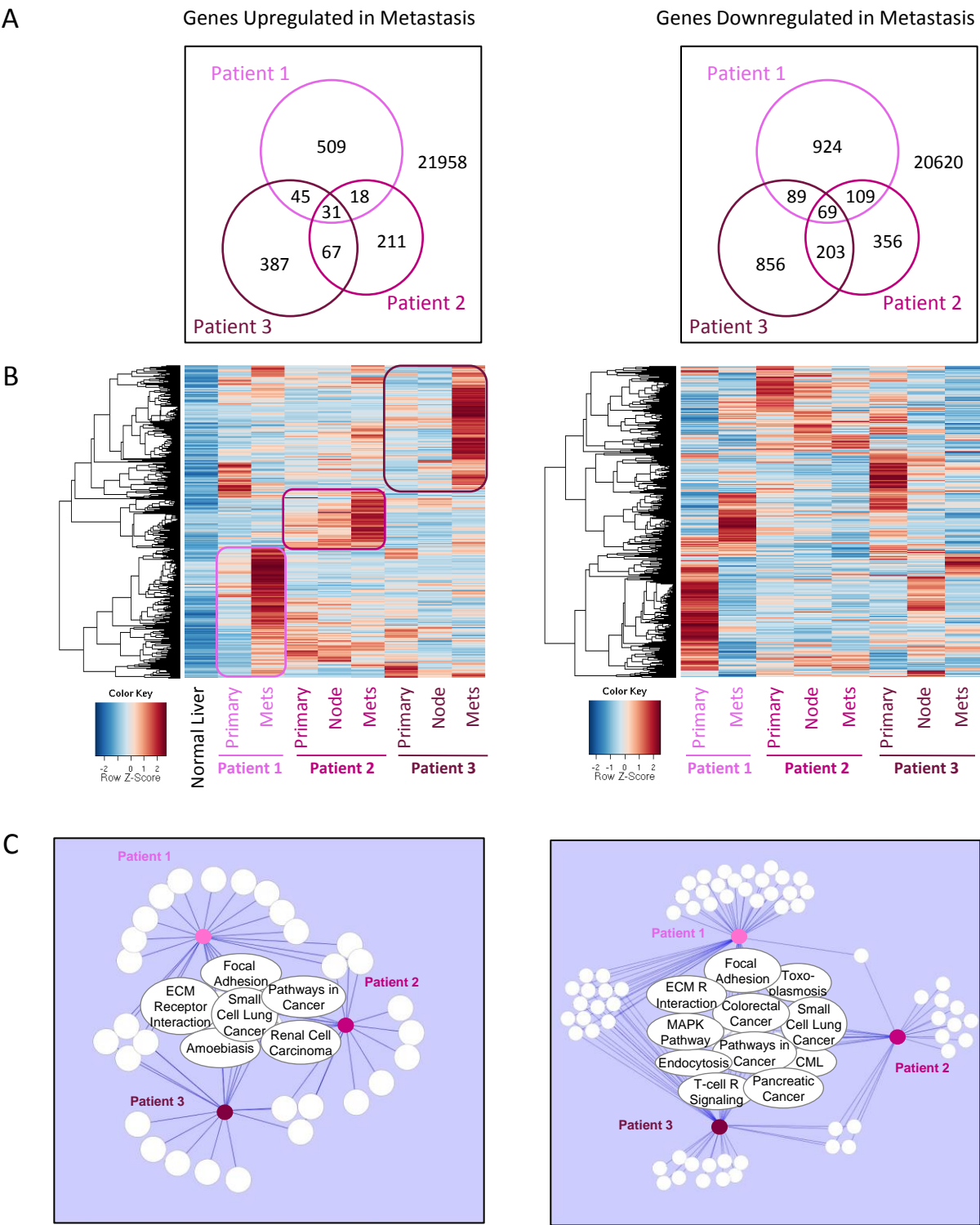


Figure 1



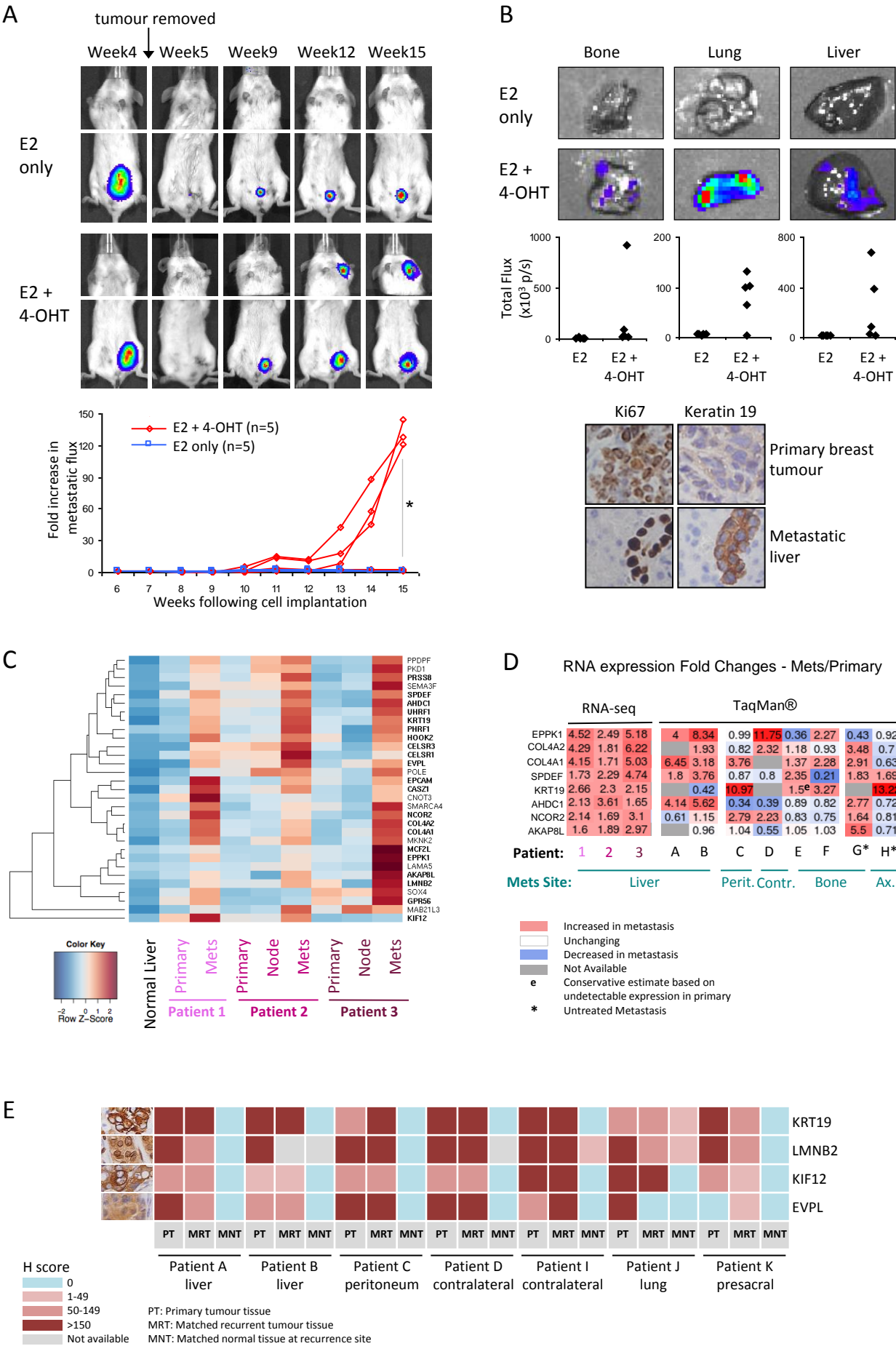


Figure 3

Table 1: Tumour and Disease Characteristics for sequenced patient samples

	Patient 1	Patient 2	Patient 3
Age at Diagnosis	34 yr	61 yr	37 yr
Time to Distant Metastasis	11 yr	5 yr	6 yr
Time to Death	12 yr	5 yr	9 yr
Primary Breast Pathology			
ER	+	+	+
PgR	+	-	+
HER2	-	-	+
P53	-	-	-
Ki67	10%	16%	38%
Tumour Size	50mm	26mm	NA
Tumour Grade	2	2	3
Lymph Node Status	+	+	+
Lobular/Ductal	Lobular	Ductal	Ductal
OncotypeDX Risk	Low	High	High
Luminal Classification	A	B1	B2
Surgery, Adjuvant Treatment			
Surgery	WLE +ANC Mastectomy ^a	WLE+ANC	Mastectomy+ANC
Chemotherapy	CMF x6 AC x4 ^a taxane x6 ^a	AC x4 taxane x4	CMF x6
Radiotherapy	yes	yes	yes
Endocrine Therapy	5 yr Tamoxifen 5yr Arimidex ^a 4yr Zoladex ^a	5 yr Tamoxifen	5 yr Tamoxifen
Herceptin	no	no	no
Metastatic Tumour Pathology			
Sites of Metastasis	Liver, bone	Liver, lung	Liver, lung, brain ^b
Biopsy	Liver	Liver	Liver
ER (liver biopsy)	+	NA	+
PgR (liver biopsy)	+	NA	+
HER2 (liver biopsy)	-	NA	+
p53 (liver biopsy)	NA	NA	-
Post-Metastatic Therapy	Fulvestrant	Arimidex	Herceptin Letrozole Fulvestrant

^aTreatment given 5 years following initial diagnosis

^bBrain metastasis diagnosed 8 years following initial diagnosis

Abbreviations: yr, years; WLE, wide local excision; ANC, axillary node clearance; CMF, cyclophosphamide methotrexate fluoracil; AC, adriamycin cyclophosphamide; NA, not available.

Supplementary Table S1: List of 31 genes which are upregulated in metastatic tumours relative to matched primary tumours for all three patients

CommonToAll3Patients_UpInMetsvsPrimary (31 genes)		Counts per Million Expression Values (genes highlighted in yellow are also down regulated in the nodes):											
GeneName	Description	PRIMARY				NODE				METASTASIS			
		Patient 1	Patient 2	Patient 3	Mean Primary	Patient 2	Patient 3	Mean Node	Patient 1	Patient 2	Patient 3	Mean Mets	
AHDC1	AT hook, DNA binding motif, containing 1	32.1980372	53.17648715	29.3071694	38.22723125	50.7415783	20.70643695	35.72400763	68.47514983	87.92471755	105.8556137	87.41849369	
AKAP8L	A kinase (PRKA) anchor protein 8-like	34.81412772	37.31367064	38.58014097	36.90264644	14.47436048	41.89441895	28.18438971	55.67815461	70.44336722	114.4688653	80.19679571	
CASZ1	castor zinc finger 1	107.4609491	105.2714186	88.03598933	100.256119	125.5528796	58.74849554	92.15068759	254.1438348	200.452817	163.6517801	206.0828106	
CELSR1	cadherin, EGF LAG seven-pass G-type receptor 1	324.5964625	562.5892081	210.7597612	365.9818106	584.6665833	146.2291788	365.447881	707.9881475	1352.927023	534.8954065	865.2701925	
CELSR3	cadherin, EGF LAG seven-pass G-type receptor 3	1.006188662	80.21537892	46.82278236	42.68144998	95.95362563	12.19914115	54.07638339	77.79226038	151.1165617	93.37264038	107.4271541	
CNOT3	CCR4-NOT transcription complex, subunit 3	93.77678334	48.48974591	50.82962193	64.36538372	59.52377454	29.53476278	44.52926866	200.2617497	73.0331969	105.1066353	126.1338606	
COL4A1	collagen, type IV, alpha 1	185.3399516	318.8786636	164.3949034	222.8711729	266.0680195	356.5038486	311.2859341	769.9525453	544.1232152	827.1218117	713.7325241	
COL4A2	collagen, type IV, alpha 2	158.5753332	256.328694	123.4106586	179.4382286	234.5171664	232.4257419	233.47114541	680.2613245	464.7449356	768.077348	637.694536	
EPCAM	epithelial cell adhesion molecule	21.33119964	32.44667012	42.58698053	32.12161677	28.62345442	26.32446248	27.47395845	119.2141133	74.58709471	75.64681828	89.81600876	
EPPK1	epiplakin 1	28.37452028	39.65704126	61.70532932	43.24563029	35.12878498	50.24119974	42.68499236	128.3067152	98.80200219	319.5641168	182.2242781	
EVPL	envoplakin	4.82970558	51.01337581	30.90990523	28.9176622	57.57217538	51.20428983	54.3882326	74.31237571	108.5138635	82.26279414	88.36301111	
GPR56	G protein-coupled receptor 56	94.1792588	36.77289281	123.5251398	84.82576379	34.47825192	78.97338745	56.72581968	145.5938841	55.55184658	231.3094955	144.1517421	
HOOK2	hook microtubule-tethering protein 2	13.88540354	26.85863249	34.11537688	24.95313764	30.41242033	8.346780787	19.37960056	56.01491765	54.90438916	51.30502032	54.07477571	
KIF12	kinesin family member 12	205.8662003	140.2417186	16.82872618	120.9788817	128.3176451	30.97939792	79.64852152	430.7199178	300.2907511	53.30229605	261.437655	
KRT19	keratin 19	138.6527977	218.2939862	204.9212235	187.2893358	220.6933389	168.8617959	194.7775674	369.3167917	502.1679744	441.2731066	437.5859576	
LAMA5	laminin, alpha 5	76.87281381	103.8293444	95.8207062	92.17428814	178.4086904	68.21888143	123.3137859	116.6322634	228.4229775	823.5017495	389.5189968	
LMNB2	lamin B2	28.17328255	42.90170827	54.4930181	41.85600297	44.23624775	21.187982	32.71211487	53.20855905	66.55862271	100.2382757	73.33515249	
MAB21L3	mab-21-like 3 (C. elegans)	150.7270616	182.0618712	291.8124016	208.2004448	214.3506417	575.6068443	394.978743	301.2906592	596.3082832	467.1128614	454.9039346	
MCF2L	MCF.2 cell line derived transforming sequence-like	33.00298813	21.27059486	38.92358436	31.06572245	21.79285735	30.65836789	26.22561262	55.45364593	51.53761058	202.9731461	103.3214675	
MKNK2	MAP kinase interacting serine/threonine kinase 2	195.603076	199.1865027	177.1023088	190.6306292	278.9160474	235.3150122	257.1155298	408.6058121	363.0941208	445.2676581	405.6558636	
NCOR2	nuclear receptor corepressor 2	235.0456715	151.9585717	145.5055168	177.5032534	166.2111956	101.9270346	134.0691151	502.1136806	256.7816125	451.7588042	403.5513658	
PHRF1	PHD and ring finger domains 1	52.92552364	57.50270983	62.7356595	57.72129766	63.10170635	37.23948351	50.17059493	89.57896649	112.139625	101.7362325	101.151608	
PKD1	polycystic kidney disease 1 (autosomal dominant)	86.93470043	105.9924557	62.96462176	85.29725931	178.8965902	78.17081237	128.5337013	159.2889141	186.2087538	273.1274561	206.2083747	
POLE	polymerase (DNA directed), epsilon, catalytic subunit	68.21959131	119.3316423	111.3901399	99.64712453	185.8898205	87.32016823	136.6049944	108.662205	179.4751966	199.3530838	162.4968285	
PPDPF	pancreatic progenitor cell differentiation and proliferation	98.80772665	110.679197	90.09664968	99.8611911	194.3467502	94.70385893	144.5253046	198.9146975	243.5734812	259.6458449	234.0446745	
PRSS8	protease, serine, 8	69.02454224	71.20241499	30.68094297	56.96930007	113.6806514	64.20600605	88.94332872	107.3151528	197.2155299	192.737108	165.7559302	
SEMA3F	sema domain, immunoglobulin domain (Ig), short basic	27.97204481	108.1555671	51.17306532	62.43355907	117.746483	70.94763669	94.34705983	113.6013961	163.6772356	266.136991	181.1385409	
SMARCA4	SWI/SNF related, matrix associated, actin dependent	116.1141716	91.93223202	144.81863	117.6216779	130.9197773	102.2480646	116.583921	238.9894983	165.1016419	280.8668996	228.3193466	
SOX4	SRX (sex determining region Y)-box 4	156.5629559	56.0606356	276.0140056	162.879199	74.64866807	235.4755272	155.0620976	295.9024507	93.8813258	447.7642528	279.1826764	
SPDEF	SAM pointed domain containing ETS transcription factor	91.96564374	89.40860212	39.26702775	73.5470912	87.33406265	27.6085826	57.47132263	159.2889141	205.1145104	186.2459618	183.5497955	
UHRF1	ubiquitin-like with PHD and ring finger domains 1	16.30025633	39.29652271	29.65061279	28.41579728	33.82771887	32.74506309	33.28639098	64.43399344	97.76607032	100.6127649	87.60427622	

Supplementary Table S2: List of 69 genes which are downregulated in metastatic liver tumours relative to matched primary tumours for all three patients

CommonToAll3Patients_DownInMetsvsPrimary (69 genes)		Counts per Million Expression Values											
GeneName	Description	PRIMARY				NODE				METASTASIS			
		Patient 1	Patient 2	Patient 3	Mean Prima	Patient 2	Patient 3	Mean Node		Patient 1	Patient 2	Patient 3	Mean Mets
ADCY7	adenylate cyclase 7	111.4857038	52.81596859	61.13292367	75.14486535	53.99424357	51.84634989	52.92029673		48.38162225	29.78304129	40.694493	39.61971885
AKAP12	A kinase (PRKA) anchor protein 12	57.55399149	56.42115416	64.91080098	59.62864887	42.60991511	73.1948469	57.90238101		28.8493664	22.27253523	23.3431601	24.82168724
ARHGAP30	Rho GTPase activating protein 30	238.6679507	75.52863768	76.01547063	130.0706863	59.36114128	170.6274611	114.9943012		15.26659078	44.28608748	49.05808512	36.2035878
ARHGAP31	Rho GTPase activating protein 31	78.48271567	52.99622787	65.71216889	65.73037081	47.32627976	80.89956763	64.11292369		29.74740115	27.84066903	38.822047	32.13670573
ARHGD1B	Rho GDP dissociation inhibitor (GDI) beta	267.4449465	205.1350589	147.9096206	206.8298753	191.2567182	276.7278861	233.9923021		34.91110098	85.98234529	81.388986	67.42747742
BTG1	B-cell translocation gene 1, anti-proliferative	162.1976124	65.43411808	61.93429158	96.52200735	82.78033126	99.03776434	90.9090478		38.50323998	26.67524568	34.45300635	33.21049734
BTG2	BTG family, member 2	421.995525	664.2554411	302.5736278	462.9415313	415.5279889	426.4883952	421.0081921		213.7322709	197.7334959	153.7902312	188.418666
BTN3A1	butyrophilin, subfamily 3, member A1	94.38049653	102.3872702	90.44009307	95.73595326	61.9632735	137.721883	99.84257824		31.76797935	49.59523832	45.18836339	42.18386035
CCND2	cyclin D2	236.6555734	173.0489073	101.773725	170.4927352	137.9130077	163.7253154	150.8191616		28.73711206	87.01827716	16.22786531	43.99441817
CDR1	cerebellar degeneration-related protein 1, 34kDa	226.9961622	262.6377687	300.055043	263.229658	104.5731886	179.776817	142.1750028		66.34231729	61.7674378	20.09758703	49.40244738
CELF2	CUGBP, Elav-like family member 2	209.2872418	61.64867323	70.6348575	113.8569242	108.58283648	139.3270331	104.9549348		11.67445177	18.38779071	14.23058958	14.76427735
CLIP3	CAP-GLY domain containing linker protein 3	53.73047457	81.29693459	60.67499915	65.2341361	67.81807099	69.82403158	68.82105129		33.78855754	35.09219213	28.08668996	32.32247987
CYR61	cysteine-rich, angiogenic inducer, 61	169.240933	101.3186784	79.67886681	119.7461594	72.37180238	163.7253154	118.0485589		55.79040896	6.086099742	21.8452033	27.90723733
DOCK10	dedicator of cytokinesis 10	92.77059467	88.86782428	51.28754645	77.64198847	78.71449967	156.0205947	117.3675472		36.03364442	35.61015806	22.34452223	31.32944157
EGR1	early growth response 1	1065.553793	663.8949226	233.6559873	654.3682345	345.1077857	776.0900982	560.598942		43.77919415	34.05626026	63.41350443	47.08298628
ELN	elastin	99.00896438	76.42993407	72.00863106	82.48250984	24.06972304	48.7965646	36.43314382		23.34890355	38.07049626	31.95641168	31.1252705
ETS1	v-ets avian erythroblastosis virus E26 oncogene	246.1137468	86.70471294	101.0868382	144.6350993	75.13656786	357.1459087	216.1412383		62.07665222	36.9050729	35.20198474	44.72790329
EZH1	enhancer of zeste 1 polycomb repressive complex 2	154.3493408	121.1342351	55.63782941	110.3738018	101.6457899	75.92360216	88.78469601		85.08879274	80.28472	33.82885768	66.40079014
FKBP1A	FK506 binding protein 1A, 12kDa	168.4359821	125.4604578	160.617026	151.5044886	116.1201503	204.0145842	160.0673673		80.48636463	65.0047249	105.9804434	83.82384432
FLJ10038	Uncharacterized Protein FLJ10038	78.88519113	57.68296911	81.39608376	72.654748	83.10559779	65.49012618	74.29786198		27.16555124	38.07049626	38.57238754	34.60281168
FMNL3	formin-like 3	224.9837849	238.6632847	215.2245253	226.2905316	185.0766542	232.5974161	204.3370351		88.23191437	88.96064942	104.9818056	94.05812312
FOS	FBJ murine osteosarcoma viral oncogene homolog	485.9891239	249.3879257	63.65150854	266.0126394	145.7194043	385.5570664	265.6382353		12.12346915	10.48881019	11.10984625	11.24070853
FYN	FYN binding protein	173.6681631	95.53741759	64.91080098	111.3721272	118.397016	137.6100929	137.6100929		27.95133165	47.3938831	30.45845488	35.26078988
GOSR2	golgi SNAP receptor complex member 2	57.35275376	111.0397155	54.37853697	74.25700209	96.27889216	71.26866672	83.77377944		34.79884663	66.81760568	35.82613341	45.81419524
GPCPD1	glycerophosphocholine phosphodiesterase GDE1	100.2163908	76.9707119	76.58787628	84.59165965	67.16753794	51.04377481	59.10565638		47.93260488	40.9193089	37.69857941	42.18349773
IFI16	interferon, gamma-inducible protein 16	64.39607439	71.20241499	95.47726281	77.02525073	49.11524566	82.98626283	66.05075424		31.655725	46.48744271	20.7217357	32.9549678
IGLL5	immunoglobulin lambda-like polypeptide 5	189.9684195	1005.846774	106.8108947	434.208696	577.8359862	224.8815362	401.3587612		8.419075798	250.9544957	12.23331385	90.53562846
IL16	interleukin 16	204.0550607	58.40406622	73.15344237	111.8708364	19.19072513	135.7950702	77.49321397		22.56312314	30.94846464	13.10712198	22.20623659
INPP4A	inositol polyphosphate-4-phosphatase, type I, 107kDa	104.2411454	58.04348767	84.94498881	82.4098773	69.69076891	64.20600608	68.45153748		32.89052279	25.25083935	48.18427699	35.44187971
KLF12	Kruppel-like factor 12	107.2597114	70.12085932	75.32853885	84.23638486	36.26721782	76.08411717	56.1756675		57.81098715	29.39456684	42.44210927	43.21588775
KLF6	Kruppel-like factor 6	161.1914237	272.552029	200.914384	211.5526122	142.6293723	359.5536339	251.0915031		93.05885116	83.39251561	91.87468359	89.44201679
LOC1001308	uncharacterized LOC100130899	58.15770469	72.64448922	86.54773463	72.44997618	82.12979821	49.75965469	65.94472645		19.98127323	45.45151084	55.67406098	40.36894835
LOC728264	uncharacterized LOC728264	249.736026	242.088211	278.8760338	256.9000903	210.1221768	266.4549251	238.288551		88.5686774	70.70235019	129.0739441	96.11499055
MAML2	mastermind-like 2 (Drosophila)	172.259499	65.43411808	74.06929141	103.9209695	61.9632735	108.0266052	84.99493934		35.58462704	20.97762039	39.69585514	32.08603419
MARS	methionyl-tRNA synthetase	96.19163613	69.39982221	108.0701872	91.2205485	89.93619487	38.04205859	63.98912673		37.94196826	38.07049626	72.02675602	49.34640685
MBNL1	muscleblind-like splicing regulator 1	249.736026	195.5813171	176.1864598	207.1679343	146.0446709	272.3939807	209.2193258		129.9905303	110.3267443	85.75802666	108.6917671
MCL1	myeloid cell leukemia 1	297.2281309	394.7678198	480.1338612	390.7099373	430.6528825	497.7570619	464.2049722		196.2205933	224.797216	198.9785946	206.665468
MFAP4	microfibrillar-associated protein 4	152.739439	80.75615675	70.17693298	101.2241762	28.46082116	21.187982	24.82440158		11.00092571	40.66032593	21.72037357	24.46054174
MIAT	myocardial infarction associated transcript (non-protein)	93.97802107	111.5804934	69.37556506	91.64469317	91.56252751	81.7021427	86.63233511		9.31711055	18.25829923	9.362229985	12.31254659
MYCBP2	MYC binding protein 2, E3 ubiquitin protein ligase	215.5256115	172.868648	157.0681111	181.8207902	124.9023466	196.1502029	210.5262748		100.8044009	100.096917	84.63455907	95.17862569
NFIB	nuclear factor I/B	91.76440601	125.041968	128.4478284	115.0439715	76.92553377	135.6351878	106.2803608		46.58555275	26.54575419	76.14613721	79.75914805
NLRC5	NLR family, CARD domain containing 5	310.1073458	135.0141996	70.52037637	171.8806406	136.7745748	155.5390497	146.1568123		18.74647544	89.47861535	40.19517407	49.47342162
PCSK7	proprotein convertase subtilisin/kexin type 7	136.4391826	52.99622787	72.69551785	87.37697611	58.06007517	77.3682373	67.71415623		18.07294938	30.94846464	36.57511181	28.53217528
PLEKHG2	pleckstrin homology domain containing, family G (with	134.4268053	89.5888614	114.4811305	112.8322657	64.89067225	156.8231698	110.856921		45.12624628	58.78913368	56.18966964	58.18966964
PLXDC1	plexin domain containing 1	69.02454224	74.80760056	131.4243378	91.75216019	53.99424357	41.73930394	47.86407375		26.82788821	28.22914348	30.45845488	28.50546219
PLXNC1	plexin C1	77.47652701	81.29693459	93.30212133	84.02519431	66.84227141	79.29441748	73.06834444		27.39005993	40.66032593	31.33226302	33.12754963
PRRX1	paired related homeobox 1	56.74904056	87.60600933	162.9064687	102.4205662	88.81276411	97.27209917	90.59543164		7.072023671	4.143727484	6.19356847	6.19356847
PTPRC	protein tyrosine phosphatase, receptor type, C	160.7889483	92.29275057	85.2884422	112.790047	53.99424357	238.2042825	146.099263		24.69595568	40.78981472	14.35541931	26.6137308
PTRF	polymerase I and transcript release factor	142.2750769	111.2199748	161.9907996	138.4952838	93.67675994	139.0223109	131.3495354		41.75861596	66.29963974	76.27096695	61.44307421
RPS12	ribosomal protein S12	222.9714076	59.12504334	282.1959866	188.0974792	53.66897705	189.0866878	121.3778324		128.5312239	32.37287097	119.3372249	93.41377323
SEC14L1	SEC14-like 1 (S. cerevisiae)	81.30004392	86.88497222	77.84716872	82.01072829	73.1849687	38.68411865	55.93454367		35.92139007	54.77489768	47.18563912	45.96064229
SEPT6	septin 6	289.581097	128.8583841	125.5858001	101.4831566	279.29961263	190.3896415	54.90438916		39.51352908	54.90438916	29.70947649	41.37579824
SLC7A6	solute carrier family 7 (amino acid transporter light	165.2161784	101.3057145	105.4371212	123.986338	87.9845957	118.2995662	103.1420809		87.78289699	59.56608258	63.91282336	70.42060098
SNED1	sushi, nidogen and EGF-like domains 1	68.21959131	179.8987599	115.7404229	121.286258	107.8258539	62.44034089	85.13309739		25.93075346	114.5999632	69.90456056	70.14512241

SPARCL1	SPARC-like 1 (hevin)	56.14532736	71.56293355	156.6101865	94.7728158	16.26332638	59.71158563	37.987456	9.092601862	33.27931135	3.744891994	15.3722684
STARD9	StAR-related lipid transfer (START) domain containing	110.6807529	68.49852582	110.4742909	96.55118986	18.37755881	85.71501808	52.04628844	43.10566809	24.73287342	31.70675222	33.18176457
STK4	serine/threonine kinase 4	141.6713637	109.9581599	104.9791966	118.8695734	112.3795853	175.4429115	143.9112484	72.85306924	66.42913122	69.77982082	69.68734043
TAGLN	transgelin	194.798125	129.2459027	217.1707045	180.4049107	62.77643982	252.3296038	157.5530218	56.23942633	50.63117019	126.7021791	77.85759189
TAGLN2	transgelin 2	240.8815658	217.7532084	355.2349478	271.2899073	189.305119	162.4411953	175.8731572	143.1242886	136.225041	200.3517217	159.9003504
TCF4	transcription factor 4	135.8354694	98.42156604	170.4624033	134.9064796	83.43086432	134.9931277	109.211996	66.34231729	41.95524077	38.19789834	48.8318188
TGFB2	transforming growth factor, beta receptor II	149.720873	101.6662331	80.59471585	110.6606073	106.3621545	174.3193064	140.3407305	50.85121782	54.64540619	45.43802286	50.31154896
TNC	tenascin C	93.97802107	103.1083073	159.1287714	118.7383666	115.1443508	202.2489191	158.6966349	60.72960009	24.73287342	99.73895677	61.7338101
TRAF1	TNF receptor-associated factor 1	138.8540354	51.37389436	66.17009341	85.46600773	47.32627976	82.50471778	64.91549877	12.90924956	24.085416	34.07851715	23.6910609
TRIM22	tripartite motif containing 22	90.75821735	109.7779006	67.88731037	89.4744761	68.79387058	94.86437395	81.82912226	23.79792092	49.85422129	15.22922744	29.62712322
TTN	titin	663.2795663	320.6812564	389.9227304	457.9611843	185.8898205	794.0677799	489.9788002	70.04671064	63.32133561	169.8932668	101.0871043
UTRN	utrophin	442.5217737	376.3813734	586.7157937	468.5396469	251.4310258	541.096116	396.2635709	257.9604825	183.4894326	305.5831867	249.0110339
VAMP1	vesicle-associated membrane protein 1 (synaptobrevin	188.5597553	171.0660553	87.23462142	148.9534773	119.8607154	95.18540398	107.5230597	48.38162225	61.50845484	53.30229605	54.39745771
WIPI1	WAS/WASL interacting protein family, member 1	256.3768712	123.2973465	102.2316495	160.6352891	117.9091162	133.3879776	125.6485469	32.32925107	53.47998284	47.56012832	44.45645408
ZEB2	zinc finger E-box binding homeobox 2	184.7362384	148.5336455	135.6601396	156.3100078	148.1589033	206.4223095	177.2906064	91.03827297	72.38573948	53.17746632	72.20049292

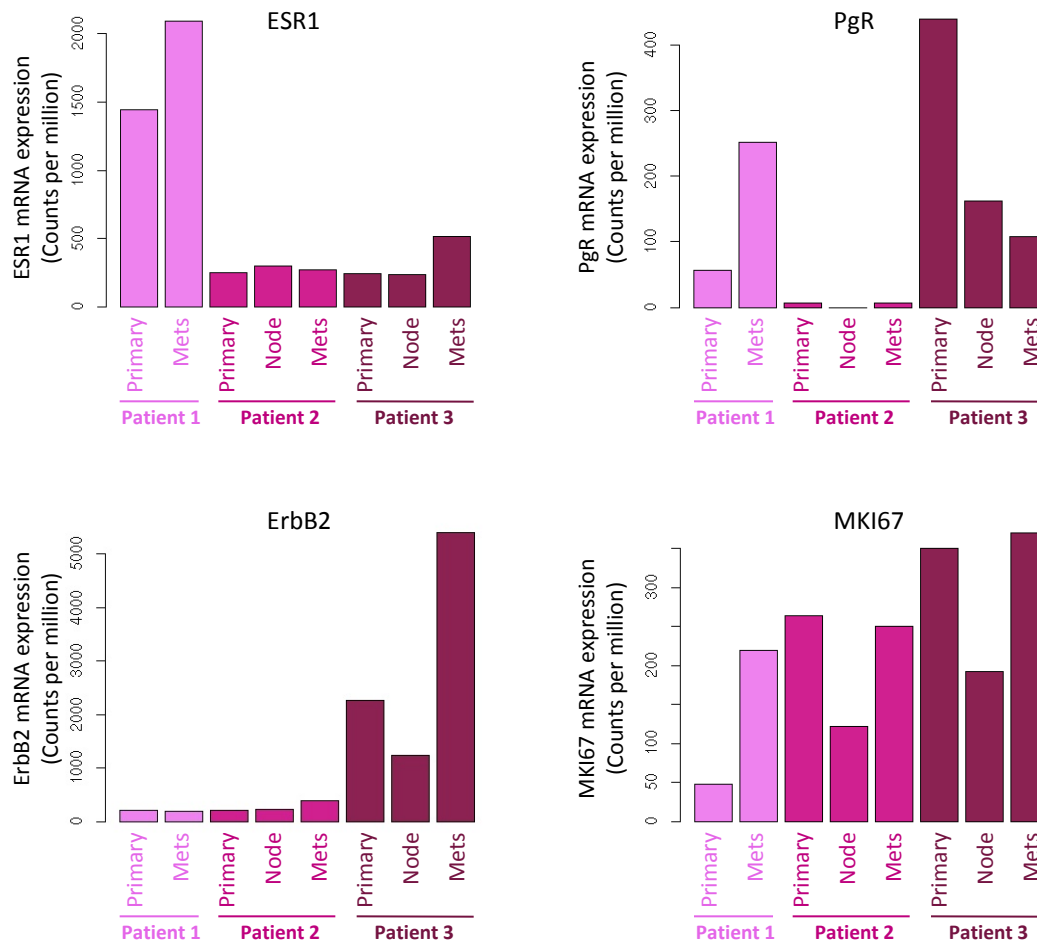
Supplementary Table S3: List of 21 genes which are upregulated in the metastatic liver relative to both matched primary and matched node in all three patients (where available)

CommonToAll3Patients_UpInMetsvsPrimary&Node (21 Genes)		Counts per Million Expression Values:											
GeneName	Description	PRIMARY				NODE				METASTASIS			
		Patient 1	Patient 2	Patient 3	Mean Prima	Patient 2	Patient 3	Mean Node		Patient 1	Patient 2	Patient 3	Mean Mets
AKAP8L	A kinase (PRKA) anchor protein 8-like	34.81412772	37.31367064	38.58014097	36.90264644	14.47436048	41.89441895	28.18438971		55.67815461	70.44336722	114.4688653	80.19679571
CASZ1	castor zinc finger 1	107.4609491	105.2714186	88.03598933	100.256119	125.5528796	58.74849554	92.15068759		254.1438348	200.452817	163.6517801	206.0828106
CELSR1	cadherin, EGF LAG seven-pass G-type receptor 1	324.5964625	562.5892081	210.7597612	365.9818106	584.6665833	146.2291788	365.447881		707.9881475	1352.927023	534.8954065	865.2701925
CELSR3	cadherin, EGF LAG seven-pass G-type receptor 3	1.006188662	80.21537892	46.82278236	42.68144998	95.95362563	12.19914115	54.07638339		77.79226038	151.1165617	93.37264038	107.4271541
COL4A1	collagen, type IV, alpha 1	185.3399516	318.8786636	164.3949034	222.8711729	266.0680195	356.5038486	311.2859341		769.9525453	544.1232152	827.1218117	713.7325241
COL4A2	collagen, type IV, alpha 2	158.5753332	256.328694	123.4106586	179.4382286	234.5171664	232.4257419	233.4714541		680.2613245	464.7449356	768.077348	637.694536
EPCAM	epithelial cell adhesion molecule	21.33119964	32.44667012	42.58698053	32.12161677	28.62345442	26.32446248	27.47395845		119.2141133	74.58709471	75.64681828	89.81600876
EPPK1	epiplakin 1	28.37452028	39.65704126	61.70532932	43.24563029	35.12878498	50.24119974	42.68499236		128.3067152	98.80200219	319.5641168	182.2242781
EVPL	envoplakin	4.82970558	51.01337581	30.90990523	28.9176622	57.57217538	51.20428983	54.3882326		74.31237571	108.5138635	82.26279414	88.36301111
GPR56	G protein-coupled receptor 56	94.1792588	36.77289281	123.5251398	84.82576379	34.47825192	78.97338745	56.72581968		145.5938841	55.55184658	231.3094955	144.1517421
HOOK2	hook microtubule-tethering protein 2	13.88540354	26.85863249	34.11537688	24.95313764	30.41242033	8.346780787	19.37960056		56.01491765	54.90438916	51.30502032	54.07477571
KIF12	kinesin family member 12	205.8662003	140.2417186	16.82872618	120.9788817	128.3176451	30.97939792	79.64852152		430.7199178	300.2907511	53.30229605	261.437655
KRT19	keratin 19	138.6527977	218.2939862	204.9212235	187.2893358	220.6933389	168.8617959	194.7775674		369.3167917	502.1679744	441.2731066	437.5859576
LMNB2	lamin B2	28.17328255	42.90170827	54.4930181	41.85600297	44.23624775	21.187982	32.71211487		53.20855905	66.55862271	100.2382757	73.33515249
MCF2L	MCF.2 cell line derived transforming sequence-like	33.00298813	21.27059486	38.92358436	31.06572245	21.79285735	30.65836789	26.22561262		55.45364593	51.53761058	202.9731461	103.3214675
NCOR2	nuclear receptor corepressor 2	235.0456715	151.9585717	145.5055168	177.5032534	166.2111956	101.9270346	134.0691151		502.1136806	256.7816125	451.7588042	403.5513658
PHRF1	PHD and ring finger domains 1	52.92552364	57.50270983	62.7356595	57.72129766	63.10170635	37.23948351	50.17059493		89.57896649	112.139625	101.7362325	101.151608
PRSS8	protease, serine, 8	69.02454224	71.20241499	30.68094297	56.96930007	113.6806514	64.20600605	88.94332872		107.3151528	197.2155299	192.737108	165.7559302
SPDEF	SAM pointed domain containing ETS transcription factor	91.96564374	89.40860212	39.26702775	73.5470912	87.33406265	27.6085826	57.47132263		159.2889141	205.1145104	186.2459618	183.5497955
UHRF1	ubiquitin-like with PHD and ring finger domains 1	16.30025633	39.29652271	29.65061279	28.41579728	33.82771887	32.74506309	33.28639098		64.43399344	97.76607032	100.6127649	87.60427622

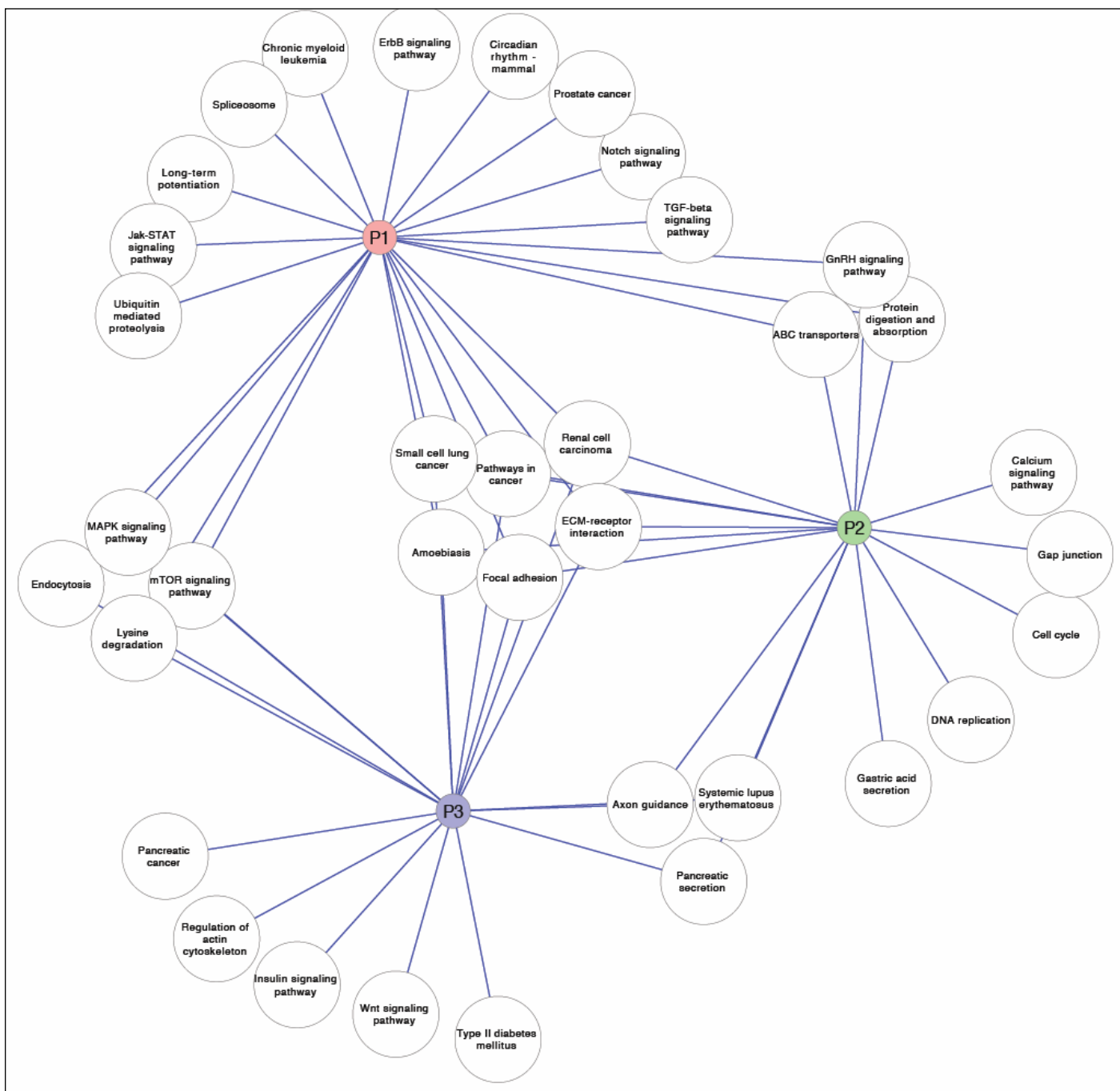
Supplementary Table S4: Patient information for patients A-L

	Patient A	Patient B	Patient C	Patient D	Patient E	Patient F	Patient G	Patient H	Patient I	Patient J	Patient K	Patient L
Age at Diagnosis	39	44	62	70	70	55	71	45	50	72	36	48
Time to Recurrence	7yr	5yr	6yr	2yr	4yr	4yr	0yr	5yr	3yr	3yr	4yr	6yr
Endocrine Treatment	Tamoxifen	Tamoxifen	Tamoxifen	AI	Tamoxifen	AI	none	none	AI	AI	Tamoxifen	Tamoxifen
Primary Breast Pathology												
ER	+	+	+	+	+	+	+	+	+	+	+	+
PgR	+	-	-	-	-	-	-	-	+	+	+	+
HER2	-	-	-	+	+	+	+	NA	+	-	+	+
Recurrent Tumour Pathology												
Site of Recurrence	liver	liver	peritoneum	contra breast	bone	bone	bone	axilla	contra breast	lung	presacral	contra breast
ER	+	+	-	+	+	+	NA	+	-	+	+	+
PgR	+	-	-	-	+	-	NA	-	-	-	+	-
HER2	-	-	-	+	+	+	NA	-	+	+	-	-

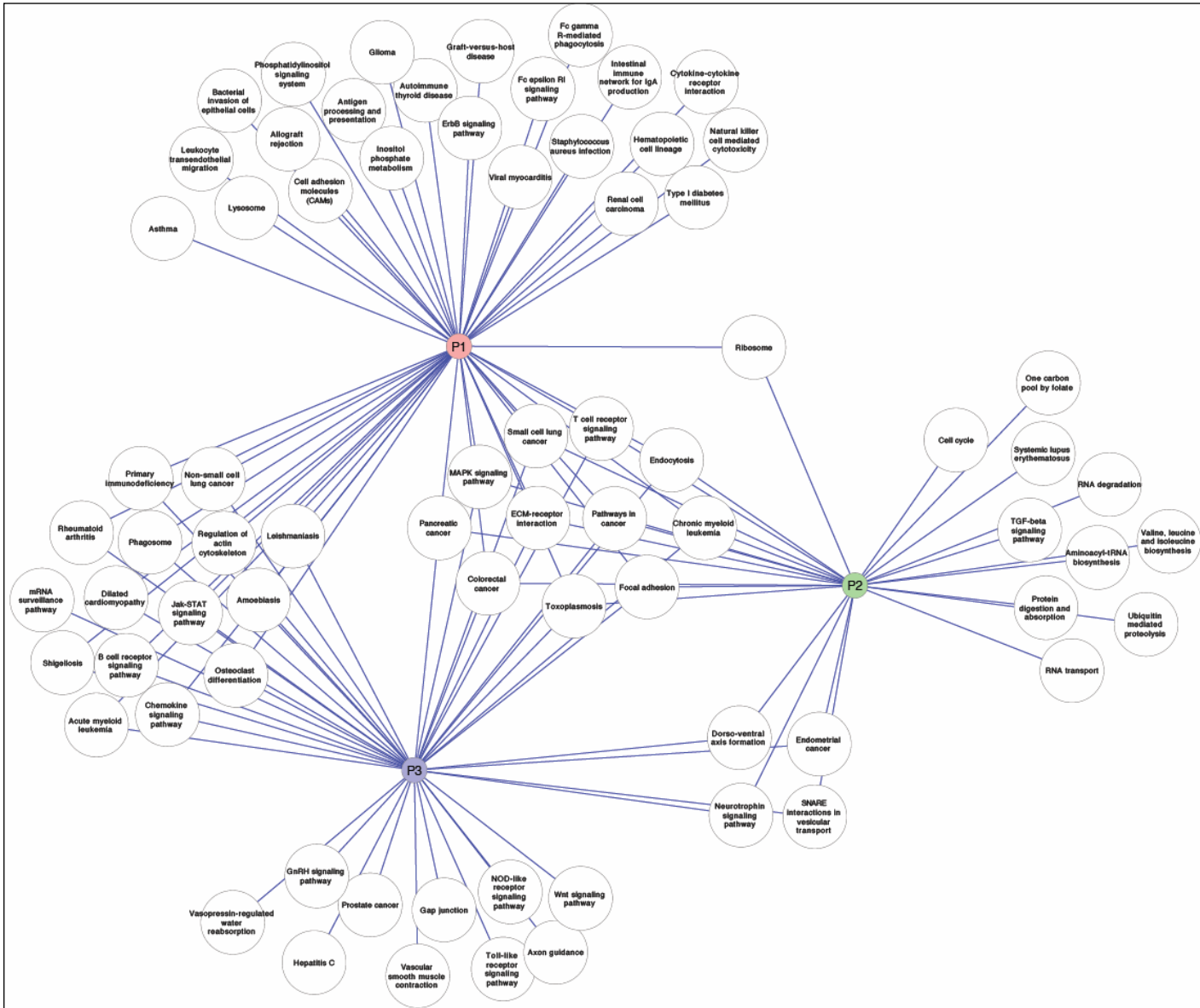
Abbreviations: AI, aromatase inhibitor; NA, not available; contra, contralatera



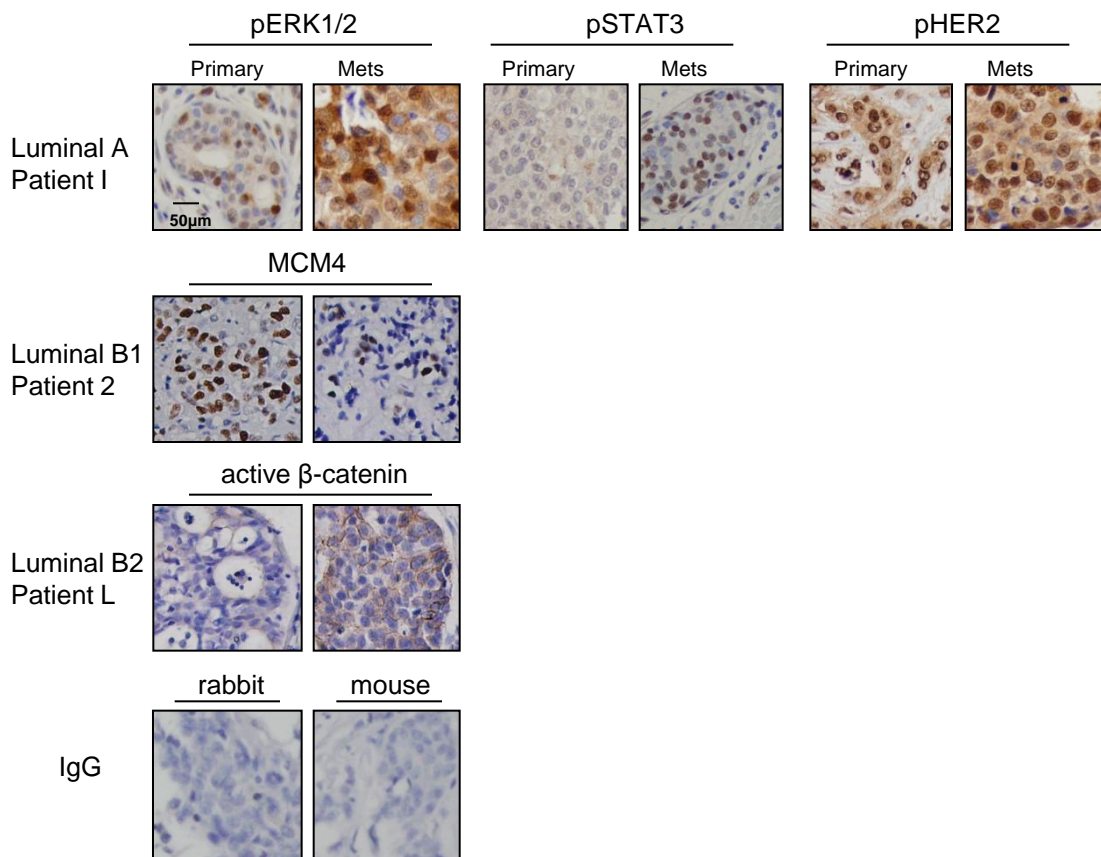
Supplementary Figure S1: Gene expression profiles of classic biomarkers in matched tumour samples from 3 patients. RNAseq data was mined for expression of the estrogen receptor (ESR1), the progesterone receptor (PgR), the human epidermal growth factor receptor 2 (ErbB2) and the proliferation marker Ki67 (MKI67). Gene expression in primary tumours largely reflected the protein expression observed by immunohistochemistry.



Supplementary Figure S2: Network map showing the enriched KEGG pathways from upregulated differentially expressed genes (DEGs). For each patient, relative gene expression was compared between the metastatic liver tumour and the primary breast tumour. In total, 1,268 genes were identified as being significantly upregulated in at least one patient (fold change >1.5 with liver metastatic tumour showing higher expression than primary tumour, >50 counts per million in the metastatic tumour, expression in a normal liver sample < expression in the primary tumour). Enriched KEGG pathway terms were identified for each patient using a hypergeometric test from the GOSTATS package in Bioconductor. The network map allows us to visualise the KEGG pathway terms which are common or unique.

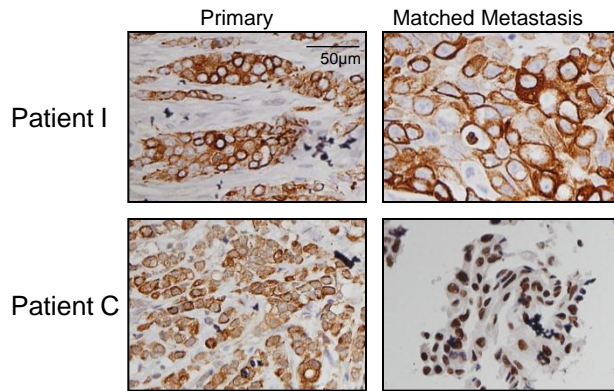


Supplementary Figure S3: Network map showing the enriched KEGG pathways from downregulated differentially expressed genes (DEGs). For each patient, relative gene expression was compared between the metastatic liver tumour and the primary breast tumour. In total, 2,606 genes were identified as being significantly downregulated in at least one patient (fold change >1.5 with liver metastatic tumour showing lower expression than primary tumour, >50 counts per million). Enriched KEGG pathway terms were identified for each patient using a hypergeometric test from the GOSTATS package in Bioconductor. The network map allows us to visualise the KEGG pathway terms which are common or unique.

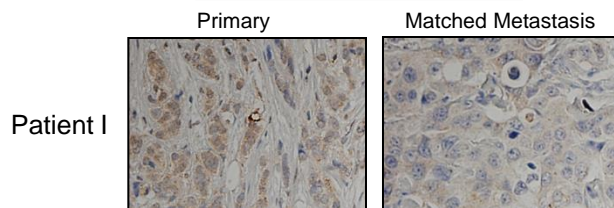


Supplementary Figure S4: Protein expression in matched primary and metastatic tissue from endocrine treated breast cancer patients. Immunohistochemistry analysis confirmed expression of phosphorylated growth factor signalling proteins, pERK1/2, pSTAT and pHER2 in both primary and metastatic tissue from a luminal A patient, Patient I (see Supplementary Table S4 for patient details). Protein expression of MCM4, a cell cycle protein, was confirmed in the matched primary and metastatic tumours from patient 2. Active β -catenin, a protein involved in the WNT signalling pathway, was detected in both primary and metastatic tumours from patient L, who presented with a luminal B2 primary tumour similar to patient 3 (see Supplementary Table S4 for patient details). Negative IgG controls for both rabbit and mouse antibodies are shown in the lower panel.

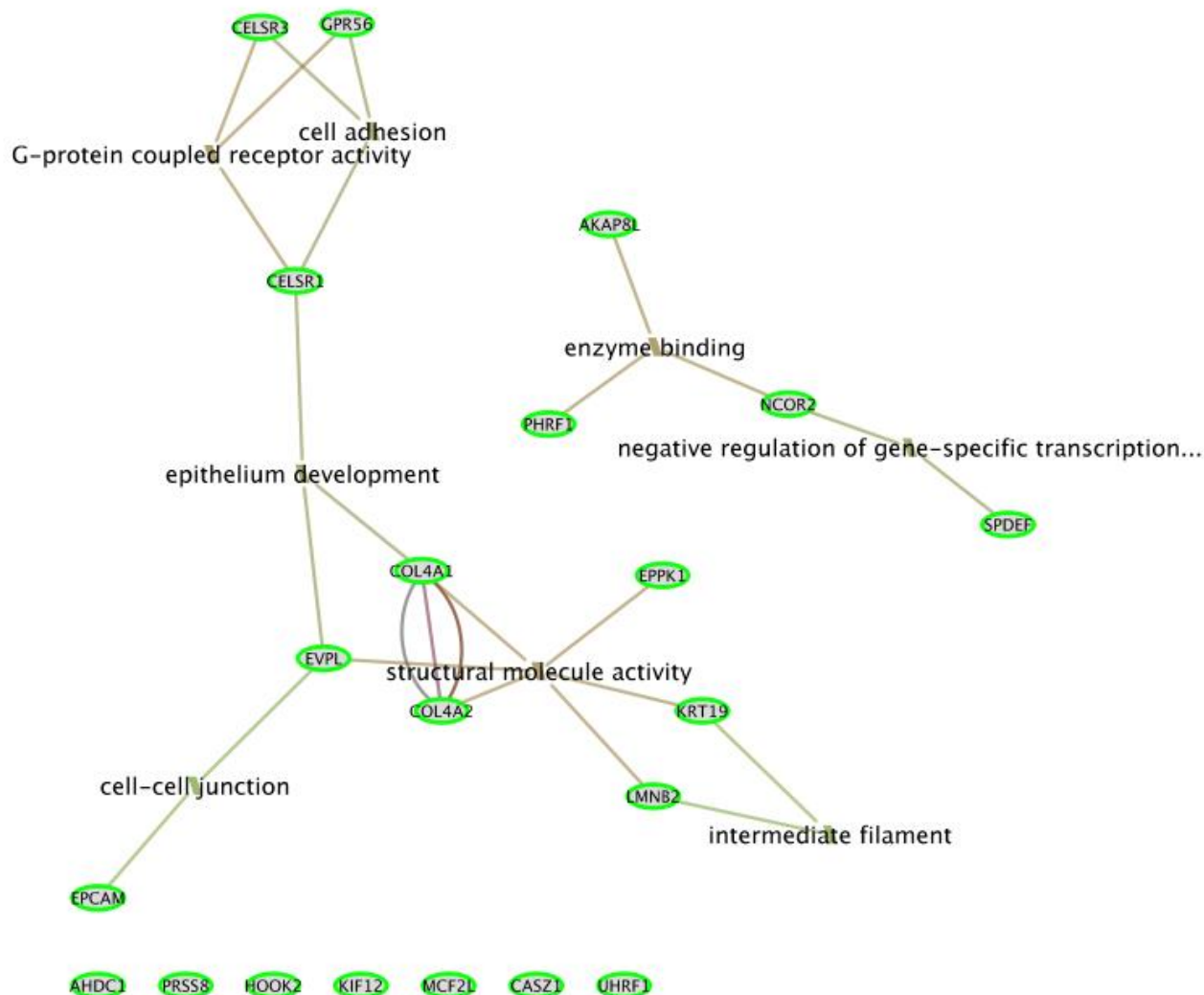
Phospho-mTOR



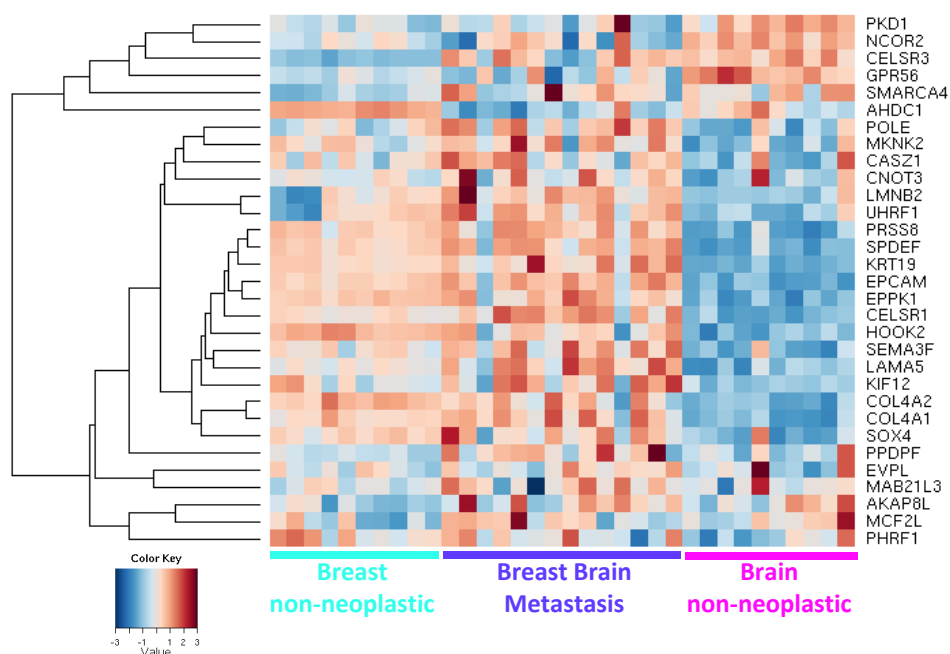
Phospho-p70S6K



Supplementary Figure S5: Phosphorylated mTOR signaling in endocrine resistant breast cancer tumours. High levels of phosphorylated mTOR protein was detected by immunohistochemistry in primary tumour and matched metastatic tissue in patient I and patient C (see Supplementary Table S4 for patient details). Phosphorylated p-70S6K, a downstream target of mTOR, was also detected at the protein level in both the primary and matched tissue from patient I (Supplementary Table S4).



Supplementary Figure S6: Interaction plot showing functional roles for 21 genes commonly upregulated in metastatic liver tumours of all three patients. These 21 genes are significantly upregulated in the liver metastatic tumours relative to both primary and nodal tumours for all three patients, where available (fold change >1.5, sequence read >50 counts per million, expression in a normal liver sample < expression in the primary tumour). The 21 genes and their expression are listed in Supplementary Table S3. The plot was generated using the Exploratory Gene Analysis Network software tool. Each of the nodes in the network represents a gene while each of the edges represents an association between the genes. Associations between genes are either from PubMed or protein-protein interaction evidence. The 21 genes were ran through enrichment analysis of KEGG/GO terms using a hypergeometric test. Significant terms were selected and added to the plot above (e.g. cell-cell junction, intermediate filament etc.).



Supplementary Figure S7: Genes upregulated in liver metastases are also strongly expressed in breast metastases to the brain. Heat map of the common 31 DEGs using publicly available microarray data from an unmatched breast cancer brain metastasis study. The majority (18) of the 31 genes showed significant upregulation in the brain metastasis relative to non-neoplastic brain. Six of these genes also showed a significant upregulation relative to the non-neoplastic breast tissue. Genes were considered significant using a p-value cut off of <0.05 and a fold change of >1.5 .

SUPPLEMENTARY METHODS

Data

RNA-seq data from the 3 patient samples is described in the main body of the paper. The following publicly available data listed here was also used in the analysis. Normal human liver RNA-seq data was downloaded from the Illumina BodyMap 2.0 Project (E-MTAB-513 from Array Express <http://www.ebi.ac.uk/arrayexpress/experiments/E-MTAB-513/>). Breast cancer samples (n=956) (1) in the form of level 3 RNA-seqV2 data were downloaded from The Cancer Genome Atlas (TCGA) data portal (<http://tcga-portals.nci.nih.gov/>). Gene expression microarray data from a breast brain metastasis study (2) were downloaded from NCBI Gene Expression Omnibus (GSE52604).

Data Analysis

The raw RNA-seq data was processed in two ways for two different types of analyses:

a) differential gene expression analysis;

b) combined analysis with the Breast Cancer Data from TCGA.

a) RNA-seq Differential gene expression analysis

Using default parameters, TopHat v1.4.1 (3) was used to align the reads from our patient samples to the hg19 genome. The reads were quantified into gene counts by using HT-Seq v0.5.3 (<http://www-huber.embl.de/users/anders/HTSeq/>). The counts were loaded into the software environment R and were processed by functions in the EdgeR package (4) from Bioconductor (5). The differential expression analysis was carried out on each patient individually using a negative binomial generalised linear model (6). A gene was considered to be differentially expressed if it exhibited a fold change of greater than 1.5 and if the gene had greater than 50 counts per million in the up regulated condition. When looking for differentially expressed genes between primary and metastasis, a normal RNA-seq liver sample from Illumina BodyMap 2.0 was used to reduce the selection of liver specific genes. Genes were filtered out if their expression was higher in the normal liver than in the primary breast cancer sample.

b) Combined analysis with TCGA

The RNA-seq data from our patient samples were processed in the same manner as the RNA-seqV2 (level 3) data from TCGA. Briefly, MapSplice v2.0 (7) was used to align the RNA-seq reads to the hg19 genome. RNA-seq by Expectation-Maximization (RSEM) v1.2.0 was used to calculate the expected counts of 20,532 human genes. The expected counts were upper quartile normalised and then filtered. They were filtered down to 15,964 genes based on two qualifying criteria: (i) if they were protein coding genes and (ii) if the mean normalised expected counts was greater than 10. The protein coding genes were retained to avoid influence from the expression differences in the non protein coding genes. Our RNA

extraction protocol produced reads from the pre-polyadenylated fraction that included a lot of non coding genes. The lowly expressed genes (mean < 10) were then removed to reduce the noise in the dataset. Out of the 956 TCGA breast cancer samples, 109 samples were removed because they were either from male patients, were normal breast samples or were outliers (n=2). The filtered normalised expected counts were then converted to a log2 scale. Each of the samples were classified by the PAM50 (8) classifier from the GeneFu package in Bioconductor. A correspondence analysis using the made4 package (9) in Bioconductor was carried out using the 15,964 human genes. Each of the samples was labeled by their PAM50 classification. OncotypeDX classifier was also carried out on our patient samples using the GeneFu package.

KEGG pathway enrichment analysis

Enriched KEGG (10) pathway terms were found using a hypergeometric test from the GStats package in Bioconductor. The hypergeometric test was carried out on the genes found to be differentially expressed in the metastasis samples vs. primary samples on each patient separately.

Gene Expression Microarray analysis:

The breast cancer brain metastasis dataset contained processed gene expression data from 35 breast to brain metastases samples, 10 non-neoplastic brain samples and 10 non-neoplastic breast samples. The brain metastasis samples were classified as the PAM50 (8) subtypes by using the 'genefu' package in Bioconductor. 12 luminal B and 2 luminal A samples were classified from the brain metastasis group. These luminal samples were assessed for differential expression against the non-neoplastic brain and the non-neoplastic breast samples. All analyses were carried out in R using the Limma package. P-values were corrected for multiple testing using the Benjamini-Hochberg method.

REFERENCES:

1. Comprehensive molecular portraits of human breast tumours. *Nature*;490(7418):61-70.
2. Salhia B, Kiefer J, Ross JT, *et al.* Integrated genomic and epigenomic analysis of breast cancer brain metastasis. *PLoS One*;9(1):e85448.
3. Trapnell C, Pachter L, Salzberg SL. TopHat: discovering splice junctions with RNA-Seq. *Bioinformatics* 2009;25(9):1105-11.
4. Robinson MD, McCarthy DJ, Smyth GK. edgeR: a Bioconductor package for differential expression analysis of digital gene expression data. *Bioinformatics*;26(1):139-40.
5. Gentleman RC, Carey VJ, Bates DM, *et al.* Bioconductor: open software development for computational biology and bioinformatics. *Genome Biol* 2004;5(10):R80.
6. McCarthy DJ, Chen Y, Smyth GK. Differential expression analysis of multifactor RNA-Seq experiments with respect to biological variation. *Nucleic Acids Res*;40(10):4288-97.

7. Wang K, Singh D, Zeng Z, *et al.* MapSplice: accurate mapping of RNA-seq reads for splice junction discovery. *Nucleic Acids Res*;38(18):e178.
8. Parker JS, Mullins M, Cheang MC, *et al.* Supervised risk predictor of breast cancer based on intrinsic subtypes. *J Clin Oncol* 2009;27(8):1160-7.
9. Culhane AC, Thioulouse J, Perriere G, Higgins DG. MADE4: an R package for multivariate analysis of gene expression data. *Bioinformatics* 2005;21(11):2789-90.
10. Kanehisa M, Goto S. KEGG: kyoto encyclopedia of genes and genomes. *Nucleic Acids Res* 2000;28(1):27-30.