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Molecular monitoring of epithelial-to-mesenchymal transition in breast cancer cells by means of Raman spectroscopy

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ABSTRACT

In breast cancer the presence of cells undergoing the epithelial-to-mesenchymal transition is indicative of 20 metastasis progression. Since metabolic features of breast tumour cells are critical in cancer progression and 21 drug resistance, we hypothesized that the lipid content of malignant cells might be a useful indirect measure 22 of cancer progression. In this study Multivariate Curve Resolution was applied to cellular Raman spectra to assess 23 the metabolic composition of breast cancer cells undergoing the epithelial to mesenchymal transition. Multivar- 24 iate Curve Resolution analysis led to the conclusion that this transition affects the lipid profile of cells, increasing 25 tryptophan but maintaining a low fatty acid content in comparison with highly metastatic cells. Supporting those 26 results, a Partial Least Square-Discriminant analysis was performed to test the ability of Raman spectroscopy to 27 discriminate the initial steps of epithelial to mesenchymal transition in breast cancer cells. We achieved a high 28 level of sensitivity and specificity, 94% and 100%, respectively. In conclusion, Raman microspectroscopy coupled 29 with Multivariate Curve Resolution enables deconvolution and tracking of the molecular content of cancer cells 30 during a biochemical process, being a powerful, rapid, reagent-free and non-invasive tool for identifying 31 metabolic features of breast cancer cell aggressiveness at first stages of malignancy. 32

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38 1. Introduction

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Genetically altered neoplastic cells have special metabolic require ments [1]. Recent molecular studies of cancer revealed that oncogenes
directly affect cellular energy metabolism [2].

An early and universal feature of tumours is the activation of lipid 42 43 metabolism secondary to the activation of the lipogenic enzymes in the malignant process [3]. Increased lipogenesis is a hallmark of many 44 cancers including breast carcinomas [4]. Lipid metabolism related 45genes in breast cancer cells have been studied describing a lipogenic 4647 phenotype [3]. Gene silencing experiments with seven genes (ACACA, ELOVL1, FASN, INSIG1, SCAP, SCAP, SCD, THRSP) indicated that reduc-48 tion of the lipidomic profiles disrupted the viability of breast cancer 49 50cells [5].

In addition, the constitutive activation of signalling cascades that
stimulate cell growth has a profound impact on anabolic metabolism
[6]. One of the principal mechanisms of aerobic glycolysis resides in

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http://dx.doi.org/10.1016/j.bbamcr.2014.04.012 0167-4889/© 2014 Published by Elsevier B.V. the activation of hypoxia-inducible factor (HIF), a transcription factor 54 activated by hypoxic, oncogenic, metabolic and oxidative stress, and 55 also involved in epithelial-to-mesenchymal transition (EMT), the initial 56 signal of the lethal metastatic phenotype of breast cancer cells [7]. In- 57 deed, the preferential expression of EMT-related genes has been found 58 in basal-like breast tumours, the most invasive breast carcinomas [8] 59 with the worst prognosis and greatest resistance to chemotherapy [9]. 60 Basal-like tumour cells express markers characteristic of the normal 61 mammary gland myoepithelium, such as epidermal growth factor 62 receptor (EGFR), p63 and basal cytokeratins CK14, CK5/6 and CK17 63 [10]. Cellular remodelling occurring as a consequence of EMT, whereby 64 cells have altered responses to agents in the circulatory system or sec- 65 ondary tumour site, could be advantageous for the process of metastasis 66 [11,12]. EMT confers mesenchymal properties on epithelial cells and has 67 been closely associated with the acquisition of aggressive traits by carci- 68 noma cells [13]. Moreover, the dynamic interactions among epithelial, 69 self-renewal and mesenchymal gene programmes determine the plas-70 ticity of epithelial tumour-initiating cells [14]. 71

A close relationship between changes in fatty acid metabolism 72 is associated to cell motility and the proclivity to breast cancer cells undergoing EMT [15]. We have previously characterized the metabolic **Q3** phenotype of breast cancer cells by using Raman microspectroscopy 75 (RS) combined with Principal Component Analysis (PCA), based on a 76 panel of small molecules derived from the global or targeted analysis 77

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of metabolic profiles of cells. RS combined with PCA has emerged as a 78 79useful technique for stratifying breast cell malignancy, being well correlated with the EMT phenotype and the expression of SREBP-1c 80 81 and ABCA1 genes [16]. Since, metabolic features of breast tumour cells are critical in cancer progression and drug resistance [9,17], there exists 82 a need to develop alternative/adjunct, user-freed, cost effective, rapid, 83 objective and unambiguous methods for the early detection and diag-84 85 nosis of metastatic breast cancer cells.

86 Recently, some unmixing algorithms have been used to decompose 87 the Raman spectra into different molecular components in areas such 88 as Raman imaging: Vertex Component Analysis (VCA) [18–20] and 89 Multivariate Curve Resolution (MCR) [21]. In the last decade, MCR algorithm has been applied in a broad range of chemical analysis 90 91[22-24] and recently to Raman images [25,26]. Multivariate Curve Resolution-Alternating Least Squares (MCR-ALS) stands out because of 92 93 its flexibility to include initial estimates and constrains in the iterative resolution procedure [21]. The output of the MCR-ALS analysis gives 94 95 more chemically and physically understandable results than classical techniques such as PCA, where only a mathematical exploration of the 96 data can be performed [27–30]. Thus, being a powerful candidate tool 97 to deconvolve key molecular signatures from the Raman spectra and 98 monitor the biochemical composition of the sample over the process 99 04 studied [31]. In the present research we used MCR-ALS algorithm to study cellular RS, which allowed the deconvolution of meaningful mo-101 lecular RS of biomolecules that are related to metabolites that change 102concentration in the cells under study. Furthermore, and supporting 103 MCR results, a Partial Least Square-Discriminant analysis (PLS-DA) 104 105was performed to test the ability of discrimination with RS of the EMT from initiating breast cells, achieving a high level of sensitivity and 106 specificity. Overall the results presented in this paper report the first 107application of MCR to RS to deconvolve and track the molecular content 108 109of initiating breast cancer cells during the EMT process, suggesting that 110RS combined with MCR and PLS-DA is a powerful diagnostic tool for identifying metabolic features of the first stages of malignancy and 111 breast cancer cell aggressiveness. 112

113 2. Material and methods

114 2.1. Cell cultures

MCF10A cells were obtained from the American Type Culture Collection and were grown in DMEM/F12 medium supplemented with 5% horse serum, 1 mM pyruvate, 2 mM L-glutamine, 0.01 mg/ml bovine insulin, 20 ng/ml EGF, 1 mg/ml hydrocortisone and 100 ng/ml tetanus toxin in 5% CO_2 -95% air at 37 °C in a humidified incubator.

 $\begin{array}{ll} \text{MDA-MB-435 cells maintained under standard conditions in a 1:1} \\ \text{(v/v) mixture of DMEM and Ham F12 medium (DMEM/F12) supplemented with 10% foetal bovine serum (FBS), 1 mM pyruvate and 2 mM L-glutamine, in the same incubator conditions as described above were used in some experiments [32]. } \end{array}$

125 **2.2.** Sample preparation

Each measurement required 3×10^5 cells, or 3×10^4 cells for 126MCF10A cells in confluent or sparse conditions. Cells were seeded in 127six-well plates (Becton Dickinson, NJ) over a quartz crystal (ESCO prod-128 ucts, Oak Ridge, NJ), which was used to reduce the background signal. 129 After 24 h, the cells were fixed with 4% cold paraformaldehyde (PFA) in 130PBS 1 \times for 15 min, washed 3 times for 10 min with PBS 1 \times (until no 131 PFA residues were present) and maintained in the same PBS $1 \times$ solution 132at 4 °C until measurements. The complete washing of the cells also 133 ensures that no residues of different culture media or other chemicals 134 are present and therefore the cells are measured under the same 135 136conditions.

2.3. Raman spectroscopy

The Renishaw Raman system (Apply Innovation, Gloucestershire, 138 UK) comprises a 514 nm laser that supplies an excitation beam of 139 about 5 mW power, which is focused onto the sample via a microscope 140 using an air $60 \times$ objective (Leica, 0.75 NA). The same objective collects 141 the scattered light from the sample and directs it to the spectrometer. 142 The spectrometer processes this scattered light, by rejecting the 143 unwanted portion and separating the remainder into its constituent 144 wavelengths. The Raman spectrum is recorded on a deep depletion 145 charge-coupled device (CCD) detector (Renishaw RenCam). The record- 146 ed Raman spectrum is digitalized and displayed on a personal computer 147 using Renishaw WiRE software, which allows the experimental param- 148 eters to be set. For experiments, one spectrum was collected per cell in 149 the cytoplasm, near the nucleus but outside the endothelial reticulum 150 [15], focusing light at 3 µ inside the quartz window. We were interested 151 in obtaining the most number of cells measured rather than study the 152 intracellular domain. That is why we acquired one Raman spectra per 153 cell, obtaining a total of 20 spectra per cell line studied: MCF10 conflu- 154 ent, MCF10 sparse and MDA-MB-435. To ensure that changes in the 155 spectra among different cell lines were due to its intrinsic biochemical 156 content and not due to changes in other factors that could be out of con-157 trol (setup alignment, background signals, temperature, conditions in 158 sample preparation ...), we measure all cell lines in the same day 159 under exactly the same conditions. Replications of the experiment 160 were performed in other different days and were comparable, providing 161 similar results. The quartz in which the cells were grown and fixed 162 was translated to a magnetic fluid chamber (Live Cell Instruments), 163 filled with PBS and closed with other quartz coverslip. For Raman 164 measurements the quartz containing the cells was placed on the top 165 (with cells inside the fluid chamber) to be measured with the upright 166 microscope. 167

Before including Raman spectra in the multivariate statistical 168 techniques, correct pre-processing must be performed. Spectra shown 169 in Fig. 1 have been pre-processed using the following methods: 170 subtracting fluorescence background [15], comic spike removal, smooth- 171 ing (5 points averaging) and multiple scattering correction (MSC); MSC 172 algorithm was used to correct for differences in global intensity among 173 the spectra. We used a custom-made Labview and Matlab code to 05 perform all these preprocessing methods. However because there is 175 still the presence of background signals from quartz or PBS, with the 176 use of MCR-ALS algorithm we subtracted those signals at the same 177 time that we looked for molecular components (explained in the next 178 section on Statistical analysis). For Raman spectral analysis, the region 179 between 1015 and 1110 cm⁻¹ was removed because it contained a 180 background-related signal that reduced the quality and interpretability 181 of our statistical models. 182



Fig. 1. Raw Raman spectra from MCF10 confluent and MCF10 sparse cells and MDA-MB-435 metastatic cells. The spectra shown have been the background subtracted similarly as in [15] and a Multiple Scatter Correction was performed.

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