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Differential Expression of PGC1 α in Intratumor Redox Subpopulations of Breast Cancer

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Abstract

Our previous studies indicate that the mitochondrial redox state and its intratumor heterogeneity are associated with invasiveness and metastatic potential in human breast cancer cell models and mouse xenografts. To further study the molecular basis of redox heterogeneity, we obtained the fluorescence images of Fp (oxidized flavoproteins containing flavin adenine dinucleotide, i.e., FAD), NADH (reduced nicotinamide adenine dinucleotide), and the Fp redox ratio ($FpR = Fp/(Fp + NADH)$) of MDA-MB-231 xenografts by the Chance redox scanner, then isolated the intratumoral redox subpopulations by dissection according to the redox ratio image. A total of 12 subpopulations were isolated from 4 tumors (2–4 locations from each tumor). The 12 subpopulations were classified into 3 FpR groups: high FpR (HFpR, $n = 4$, FpR range 0.78–0.92, average 0.85), medium FpR (MFpR, $n = 5$, FpR range 0.39–0.68, average 0.52), and low FpR (LFpR, $n = 3$, FpR range 0.15–0.28, average 0.20). The RT-PCR (reverse transcription polymerase chain reaction) analysis on these redox subpopulations showed that PGC-1 α is significantly upregulated in the HFpR redox group compared to the MFpR group (fold change 2.1, $p = 0.008$), but not significantly different between MFpR and LFpR groups, or between HFpR and LFpR groups. These results indicate that optical redox imaging (ORI)-based redox subpopulations

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exhibit differential expression of PGC1 α gene and suggest that PGC1 α might play a role in redox mediation of breast cancer progression.

1 Introduction

Intratumor heterogeneity has been studied at genetic, epigenetic, cellular and molecular levels [1–3] and is likely associated with metastasis [1, 4]. Employing the ORI technique, we previously identified intratumor redox heterogeneity in cancer mouse models and clinical breast tumor specimens based on fluorescence signals from NADH and Fp and the calculated redox ratio FpR [5]. NADH and Fp signals are contributed mainly by mitochondria. The redox ratio can be regarded as a surrogate indicator of the mitochondrial redox state and is shown to correlate linearly with NAD-coupled redox potential NAD^+/NADH as determined biochemically [6, 7]. Our studies have shown that the heterogeneity-based mitochondrial redox indices (NADH, Fp and FpR) can differentiate between tumors with different levels of aggressiveness and between premalignant and normal tissues [8–10]. Higher mitochondrial redox indices were found in breast cancer clinical specimens compared to the adjacent normal tissues [11–13].

Redox imbalance in cancer associates with biochemical changes [14, 15] and altered gene expressions that may modulate cellular metabolic and signaling pathways leading to cancer progression and metastasis [16–19]. However, no studies have reported on the differential gene expression associated with ORI-based redox states. It is likely that abnormal redox states as detected by ORI are associated with genome wide transcriptome profiles mediating cancer invasion and metastasis. In this study, we examined PGC1 α gene specifically since it is a master regulator in the mitochondrial biogenesis and function, energy production and homeostasis, and cancer progression [20–23].

2 Materials and Methods

MDA –MB-231 cell culture and xenograft

MDA-MB-231 cells were cultured in RPMI 1640 medium with 10% FBS and subcutaneously inoculated into hind thighs of athymic nude mice [24]. The induced tumors (diameter 12–14 mm) were harvested, snap-frozen, and stored in liquid nitrogen before redox scanning. All animal experiments were performed according to an Institutional Animal Care and Use Committee approved protocol.

Tissue embedding and redox scanning

The frozen mouse xenografts were embedded in a chilled mounting media (H_2O : ethanol: glycerol = 60:30:10, freezing point $-30\text{ }^\circ\text{C}$). NADH (100 μM) and FAD (100 μM) frozen solution standards were placed adjacent to the tumor and used as reference standards [24, 25]. The embedded tissue samples and standards were shaved flat to expose a surface plane, which was then scanned two-dimensionally under liquid nitrogen by the Chance redox scanner [26, 27]. The fluorescence signals of NADH and Fp were collected through optical filters for NADH (Excitation $365 \pm 13\text{ nm}$; Emission $455 \pm 35\text{ nm}$) and Fp channels

(Excitation 440 ± 10 nm; Emission 515 ± 15 nm), respectively. The image resolution is 200 μm .

Imaging analysis

A customized MATLAB[®] program was used to reconstruct images of NADH and Fp and the redox ratios in tumors. The nominal concentrations of both Fp and NADH in the tissues were obtained by comparing their fluorescence intensities with that of the corresponding standards. The redox indices, including nominal concentrations of NADH, Fp, and FpR were extracted from each image.

Tumor tissue dissection

After imaging, subpopulations with different redox states were isolated by dissection using a surgical knife from the four embedded frozen mouse xenografts according to their redox images. A total of 12 subpopulations (each sample ~2 to 5 mm) for 12 regions were obtained. The dissected tissues were immediately frozen and stored in liquid nitrogen until use for RNA isolation.

RT-PCR

RNA was isolated from dissected tumor tissues with RNeasy Mini Kit (QIAGEN, Venicia, CA, USA). First strand cDNA was synthesized from the isolated RNA with SuperScript[®] First-Strand Synthesis System for RT-PCR (Invitrogen, Carlshad, USA). PGC1 α gene expression was analyzed by PCR of cDNA in three repeated experiments for each dissected specimen. PCR was performed with the following denature steps: 95 °C for 4 min; followed by 30 cycles of 95 °C (each for 30 s), 58 °C for 1 min, and 72 °C for 30 s; then an extension step at 72 °C for 4 min. GAPDH was used as reference gene as in previous studies on cancers including breast cancer [28, 29]. PCR reaction volume (20 μl) contains various concentrations of cDNA and dH₂O, 1.6 μl primer mix and 10 μl BlueTaq mix. The RT-PCR primers were forward 5' tgaagacggattgcctcatt 3' and reverse 5' gctgggtccagtaagagctt 3' for PGC1 α ; and forward 5' ggagcgagatccctccaaat 3' and reverse 5' ggctgtgtcactactctcatgg 3' for GAPDH. PCR products were separated by electrophoresis with 2.5% of agarose gel. Gel images were taken under UV light.

Quantitative analysis of gel images and statistical analysis of gel image data

The PCR gel images were analyzed using Bio-Rad Quantity One software (v4.5.0, Bio-Rad, Hercules, CA, USA). The trace quantity (intensity integrated over band area) of a specific gene band was normalized to the reference gene GAPDH from the same tissue sample to represent the level of gene expression. The data were expressed as mean \pm standard deviation (SD) of replicates of at least three independent determinations. A p value from an unpaired Student's t-test of <0.05 was considered statistically significant.

3 Results

3.1 Intratumor Redox Subpopulations Dissected from Mouse Xenografts

Redox scanning showed that all four xenografts are heterogeneous in terms of the ORI-based redox state. The redox heterogeneity and subpopulations in a representative tumor (T62) are shown in Fig. 1 as an example, where the left image is the FpR image and the right one is the corresponding white light photo (taken with a regular camera) with dissection marks.

Based on the ranges of redox indices Fp, NADH, and redox ratio Fp/(Fp + NADH) of the 12 isolated subpopulations, these specimens were divided into high, medium, and low groups: HFpR (n = 4, FpR 0.78–0.92), MFpR (n = 5, FpR 0.39–0.68), and LFpR (n = 3, FpR 0.15–0.28); HFp (n = 4, Fp 110–341 μ M), MFp (n = 3, Fp 84–90 μ M), and LFp (n = 5, Fp 26–67 μ M); and H-NADH (n = 5, NADH 120–162 μ M), M-NADH (n = 2, NADH 48–101 μ M), and L-NADH (n = 5, NADH 14–31 μ M).

3.2 Differential Expression of PGC1 α Among Redox Subpopulations

The RT-PCR results show that PGC1 α is expressed differently among the 12 studied redox subpopulations. Figure 2a shows representative PGC1 α expressions from low, medium, and high FpR groups. Statistical analysis of the RT-PCR results shows that PGC1 α is significantly upregulated in the HFpR group compared with that of MFpR (fold change 2.10, p = 0.008), but shows no significant difference between HFpR and LFpR groups or between LFpR and MFpR groups (Fig. 2b). There is no significant difference in PGC1 α expression among the three Fp groups or among the three NADH groups.

4 Discussion and Conclusions

We previously observed distinct intratumor redox heterogeneity and identified the ORI-based redox state as potential diagnostic/prognostic biomarkers for breast tumor [5, 13]. This paper reports for the first time the differential gene expressions among the redox subpopulations based on ORI mapping. The gene expression level of PGC1 α is upregulated in more oxidized subpopulations in breast tumor xenografts of MDA-MB-231. PGC1 α , as a master regulator of mitochondrial function and bioenergetics, was shown to mediate metastasis of several types of cancer including human melanoma, prostate and breast cancer [21, 30, 31]. Thus, the results from this study that ORI-based redox heterogeneity is associated with differential gene expression of PGC1 α support the ORI redox indices as prognostic biomarkers of breast cancer. However, the details of how the redox state and PGC1 α may interact to regulate cancer progression remain unclear. Further studies are underway along this direction.

The tumor in Fig. 1 showed a similar pattern between the redox imaging and the white light photo. However, this is not always the case. Some small aggressive tumors such as MDA-MB-231 tumors (<5 mm) may not show distinct histological difference, but clearly show distinct redox difference. Although some histological differences were reported between cores and rims of human melanoma xenografts [32], we need to investigate more specifically the histological basis of breast cancer xenografts in the future.

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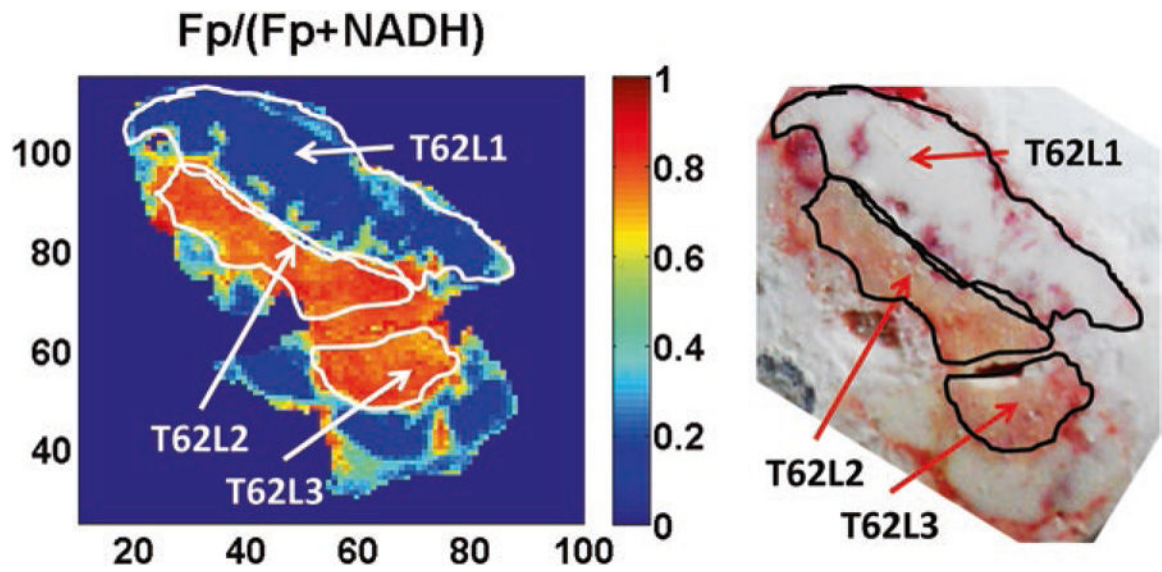


Fig. 1. Dissected redox subpopulations from a representative MDA-MB-231 xenograft. (Left) An FpR image showing three subpopulations dissected from tumor T62 with dissection marks; (Right) The corresponding white light image taken with a regular photo camera

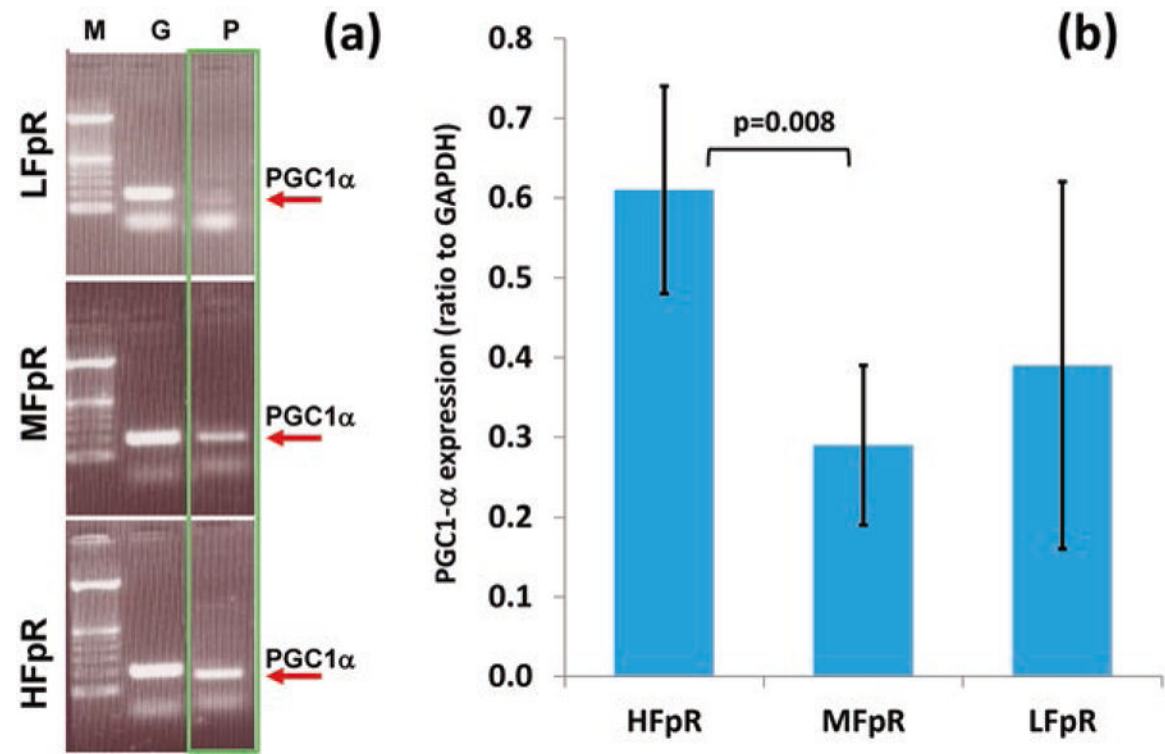


Fig. 2. PGC1 α gene expression in three FpR groups. (a) Representative PGC1 α expression levels in three FpR groups, where lane M shows the molecular weight standards, lane G shows GAPDH level, and lane P shows PGC1 α level; (b) PGC1 α expression in three FpR groups (mean \pm SD), with significant difference between HFpR and MFpR groups