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L-methionine-induced alterations in molecular signatures in

MCF-7 and LNCaP cancer cells

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Abstract

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Background—Methionine inhibits proliferation of breast and prostate cancer cells. Here, we determined the inXuence of L-methionine on functional molecular signatures in these cell lines.

Method—MCF-7 and LNCaP cells were treated with L-methionine (5 mg/ml) for 72 h. Changes in molecular signatures of these cells were examined by microarray analysis of 15,814 probes in triplicate experiments.

Results—In LNCaP cells, 325 genes were up-regulated by methionine, and 517 genes downregulated. In MCF-7 cells, 86 genes were up-regulated and 135 genes down-regulated. Ninetyeight genes were regulated in the same direction by methionine in both cells lines, and five other genes were changed in expression in opposite directions.

Conclusion—Several of the up-regulated genes encode proteins involved in cellular redox regulation, suggesting that methionine may enhance antioxidant mechanisms. Many of the downregulated genes belong to protein kinase families that may be related to the anti-proliferative effects of methionine on breast and prostate cancer cells.

Keywords

Methionine; Gene expression; Prostate cancer; Breast cancer; MCF-7 cells; LNCaP cells

Introduction

Recent investigations using new research tools have pointed to the potential of peptides as future therapeutic agents. Peptides possess a variety of potential clinical benefits, with applications in some of major categories of diseases, including cancer (Shrivastava et al. 2009; Vazquez et al. 2009). In addition to peptides, amino acid analogs are also potential candidates as new therapeutic agents. For example, our recent study (Benavides et al. 2007) has suggested that the amino acid methionine is a promising candidate anti-cancer compound, opening the prospect for development of methionine analogs as therapeutic agents.

Methionine is an essential amino acid that plays a key role in protein synthesis and in a number of other biochemical and cellular processes. Methionine is also implicated in DNAmethylation and protein-methylation by serving as the methyl-group donor, thereby serving an important role in regulation of gene expression and protein functions. Furthermore, methionine is required for the biosynthesis of the polyamines spermine and spermidine, which are critically involved in a number of cellular activities including cell proliferation (Thomas and Thomas 2001).

One approach to identifying significant molecular events of malignant transformation and tumor progression and to characterizing both known and suspected oncogenic pathways is to establish molecular signatures using gene expression profiling (Sweet-Cordero et al. 2005). While such expression signatures of cancers are frequently confounded by the complexity of human tumors (Ji et al. 2003; Giustarini et al. 2004; Ji et al. 2004; Fagerholm et al. 2008), they can be more informative when applied to in vitro systems and to genetically modified animal models where experimental variables can be controlled (Huang et al. 2003).

We have previously shown that L-methionine possesses strong inhibitory effects on cell proliferation in both breast and prostate cancer cell lines and is associated with posttranslational modification of the tumor suppressor gene p53 (Benavides et al. 2007; Benavides et al. 2010). In the present study, we have employed a global genomic approach to more comprehensively define gene signatures induced by L-methionine exposure of these cell lines and to understand the mechanisms that may underlie the methionine-mediated inhibitory effects on cell survival and cell cycle progression.

Materials and methods

Cell lines and cell culture

Wild-type p53-expressing LNCaP prostate and MCF-7 breast cancer cells were obtained from the American Type Culture Collection (ATCC; Manassas, VA). MCF-7 cells were cultured in Minimum Essential Medium (MEM; Eagle; Invitrogen, Grand Island, NY) containing 2 mM L-glutamine (Mediatech-Cellgro, Manassas, VA) 1.5 mg/L sodium bicarbonate, 0.1 mM non-essential amino acids, and 1 mM sodium pyruvate supplemented with 10% fetal bovine serum (FBS; vol/vol) (HyClone Lab Inc.; Logan, UT) and 10 mg/mL insulin (Pratt and Pollak 1993; Takahashi and Suzuki 1993). LNCaP (Horoszewicz et al. 1983) cells were cultured in RPMI 1640 media (Mediatech-Cellgro; Herndon, VA) supplemented with 10% FBS, 2 mM L-glutamine, antibiotic–antimycotic solution (1X; Mediatech-Cellgro), and MEM vitamin solution (1X; Mediatech-Cellgro). All cells were incubated at 37° C in a humidified atmosphere containing 5% CO₂.

Experimental design

Cells were seeded in 6-well plates at a concentration of 100,000 cells per well in 2 ml of media. After 48 h, cells were given fresh media to which L-methionine had been added at a concentration of 5 mg/ml or control media without additional thionine. After 72 h, cells were harvested. Three independent experiments were carried out for this study.

RNA isolation and microarray analysis

Cells were washed three times with ice-cold PBS and harvested using a trypsin. Total RNA was extracted using the RNase Mini Kit (Qiagen, Valencia, CA) according to the manufacturer's instructions. The concentration and purity of total RNA were determined spectrophotometrically at 260 and 280 nm. The quality of the RNA preparation was further evaluated by agarose gel electrophoresis. Biotin-labeled cRNA was generated from highquality total RNA using the Illumina TotalPrep RNA Amplification kit (Ambion, Austin, TX). BrieXy, 350 ng of total RNA with high 260/280 absorbance ratio (>1.8) (Gallagher 2001) and RIN number was reverse-transcribed with an oligo primer bearing T7 promoter. The first strands of cDNA, produced in the reaction, were used to make the second strands of cDNA. The purified second strands of cDNA along with biotin UTPs were used to generate biotinylated, antisense RNA of each mRNA in an in vitro transcription (IVT) reaction. The size distribution profiles for the labeled cRNA samples were evaluated by a bioanalyzer. Purified labeled cRNAs (1.5 μg) were hybridized to the Sentrix Human-6 v2 Expression Beadchip by overnight incubation at 55°C. Signals were developed with Streptavidin-Cy3. The Illumina BeadArray Reader was used to scan the chips.

Microarray data analysis

The Illumina Human WG6 V3 Expression BeadChip (Illumina: San Diego, California) was used to measure genome-wide gene expression levels. For quality control and statistical power consideration, samples from each cell line and treatment were analyzed in triplicate. The samples from the same cell lines (LNCaP and MCF-7) were laid out on the same slide (6 samples/slide) to avoid possible batch effects across slides. The Illumina Bead Array technology is based on randomly arranged beads, with each bead binding many (usually over 30) identical copies of a gene-specific probe. This redundant design yields high confidence calls and robust estimations. To take advantage of this unique feature of Illumina BeadArray, we used the Bioconductor lumi package (Du et al. 2007; Dalle-Donne et al. 2009) to preprocess Illumina data with default settings. Basically, each array was Variance-Stabilizing Transforms (VST) transformed and then followed by quantile normalization across all samples (Lin et al. 2008). Probes with intensity lower than or around background

levels were filtered. A total of 15,814 probes were used for further analysis. To identify differentially expressed genes, routines implemented in Illumina Bioconductor package (Smyth 2004) to fit linear models to the normalized expression values were applied. The variance used in the t-score calculation was corrected by an empirical Bayesian method (Smyth 2004) for better estimation relative to a small sample size. To control the effects of multiple testing and reduce false positives, *P*-values were further adjusted based on False Discovery Rate (FDR) (Benjamini and Hochberg 1995). We called genes with FDRadjusted *P*-value $\langle 0.01 \rangle$ and a fold change >1.5 as differentially expressed genes.

Pathway analysis

Pathway Analysis was performed using Ingenuity Pathway Analysis (IPA 7.0) commercial software (www.Ingenuity.com). IPA information was extracted by Ingenuity from the scientific literature, including information about genes, drugs, chemicals, cellular and disease processes, and signaling and metabolic pathways. Expression data sets containing significant differentially expressed gene identifiers (Entrez Gene identifiers) and their corresponding expression values as fold changes were uploaded as a tab-delimited text file. Each gene identifier was mapped to its corresponding gene object in the Ingenuity Pathways Knowledge Base (IPKB). To start building networks, the application program queries the IPKB for interactions between focus genes and all other gene objects stored in the knowledge base and generates a set of networks. The program then computes a score for each network according to the fit of the network to the set of focus genes. The score indicates the likelihood of the focus genes in a given network being found together due to random chance. A score of larger than 2 indicates that there is a less than 1 in 100 chance that the focus genes were assembled randomly into a network due to random chance.

Quantitative reverse transcription–PCR (QRT–PCR)

Reverse transcription was carried out with a High Capacity cDNA Reverse Transcription Kit (Applied Biosystems, Foster City, CA) on the total RNA isolated from LNCaP and MCF-7 cells that had been cultured for 72 h with or without L-methionine (5 mg/ml) using the RNeasy kit from Qiagen. PCR conditions and sequence for each primer are shown in Table 1. PCR products were separated on a 1% agarose gel. Expression of a housekeeping gene, glyceraldehyde-3-phosphate dehydrogenase (GAPDH), was used to normalize the PCR. For real-time PCR, cDNA was mixed with primers and SYBR Green PCR Master Mix according to the manufacturer's instructions (Applied Biosystems). Real-time PCR was carried out by an ABI7900-HT sequence detection system from Applied Biosystems for relative quantitation of mRNA levels, and the mRNA levels in methionine-exposed cells were plotted as fold increase compared with untreated samples. GAPDH was used for normalization ΔC_t values (target gene C_t minus GAPDH C_t) for each triplicate sample was averaged. ${}^{\Delta\Delta}C_t$ was calculated as previously described, and mRNA amplification was determined by the formula $2^{-\Delta\Delta}C_t$. For the real-time PCR of AKR1C2, we used the Taqman Gene Expression Assay from Applied Biosystems; the AKR1C2 Assay ID was Hs00912742 m1, and the GAPDH the assay ID was Hs99999905 m1.

Results

In this study, we compared the effects of treatment for 72 h with L-methionine at 5 mg/ml on gene expression in LNCaP prostate cancer cells and MCF-7 breast cancer cells, using as criteria a false discovery rate (FDR) *P*-value of greater than 0.01 and a change in expression of greater than 1.5-fold to select genes of interest in three independent experiments. Heatmaps were created to visualize the overall expression patterns of genes differentially expressed in LNCaP and MCF-7 cells treated with methionine or media without methionine added; the expression profile detected by each probe was first standardized with zero mean

and one standard deviation. Clear differences in expression patterns were observed between LNCaP cells treated L-methionine and LNCaP controls and, to a lesser extent, between MCF-7 cells treated with methionine and MCF-7 cells given media without methionine added (Fig. 1).

The expression patterns of the majority of these genes were changed by methionine in the same direction in both cells lines. Volcano plots were created to indicate the strength of biological effects (fold change) versus the reproducibility of the result (*P*-value); each gene is represented as a dot in these plots (Fig. 2). Treatment with L-methionine caused significant changes in expression of a total of 842 genes out of 15,814 probed in LNCaP cells, with 325 genes being up-regulated and 517 genes being down-regulated, while in MCF-7 cells, L-methionine treatment brought about a significant up-regulation of 86 genes and a down-regulation of 135 genes (Supplemental Tables 1 and 2). To explore the functional relevance of these findings and the observed commonalities and differences between the breast and prostate cancer cell lines, we first used Venn diagrams to identify overlapping gene signature responses to methionine between LNCaP and MCF-7 cells, again using the same criteria: FDR-adjusted *P*-value of $\langle 0.01 \rangle$ and a fold change of >1.5 . This analysis revealed that 98 genes were modified by L-methionine treatment in the same direction in both cell lines, of which 10 genes were up-regulated and 88 genes were downregulated (Table 2 and Fig. 3). In addition, the expression of the following five other genes was changed in opposite directions by L- methionine in LNCaP and MCF-7 cells (Fig. 3). H1 histone family member 0 (H1F0), Centromere Protein N (CENPN), and Acetyl-Coenzyme A Acetyltransferase 2 were up-regulated in LNCaP cells by 1.99 (*P* < 0.001), 1.65 (*P* < 0.00005), and 1.55-fold (*P* < 0.00001), respectively. In MCF-7 cells, these genes were down-regulated by 1.66 (*P* < 0.0005), 1.59 (*P* < 0.0005), and 1.57-fold (*P* < 0.002), respectively. In addition, Tumor Necrosis Factor Super Family member 2 (TNF-2) and Dehydrogenase/Reductase member 2 (DHRS2) were down-regulated in LNCaP by 1.75 and −1.52-fold, respectively (*P* < 0.0005), but they were up-regulated were up-regulated in MCF-7 cells by 1.51 ($P < 0.0005$), and 1.55-fold ($P < 0.001$), respectively. There was more variation in the response of MCF-7 cells to methionine in repeat experiments than for LNCaP cells (Fig. 3), which we cannot explain.

The 98 genes whose expression was modified by L-methionine treatment in the same direction in both cell lines were then analyzed using the data-mining tool IPA 7.0 (www.Ingenuity.com). Using this tool, we searched for functional relationships between differentially expressed genes identified in these microarray studies and those genes annotated in the Ingenuity knowledge base, the largest manually gene annotation database based on functional information available in published studies (www.Ingenuity.com). Six networks were identified that were defined as groups of two or more genes that are linked by a functional association, based on peer-reviewed published data.

Of these six functional network groups (summarized in Table 3), in group 1 (cancer, cell cycle and reproductive system diseases), only the LAMA3 gene was up-regulated, whereas 86% of genes were down-regulated and 11% was unchanged in their expression. In group 2 (cell cycle, cell assembly/organization, DNA replication, recombination and repair), only GPX8 was up-regulated, while 57% of genes were down-regulated and 40% remained unchanged. In group 3 (DNA replication, recombination and repair, cancer, and gastrointestinal disease), two genes (NQO1 and C8ORF4) were up-regulated. Fifty-one percent of the remaining genes were down-regulated, and 43% genes were unchanged. In group 4 (cancer, gastrointestinal disease and cell cycle), two genes (ATF3 and SH3BGRL) were up-regulated. In this group, 46% of genes were down-regulated, and 49% genes were unchanged. In group 5 (cancer, gastrointestinal disease, genetic disorders), four genes (AFF3, AKR1C2, RBM4B, and VAMP5) were up-regulated. Thirty-four percent of genes in

this group were down-regulated, and 54% genes were unchanged. Of the two genes in group 6 (cell death, neurological diseases, nervous system development and function), the FBXO38 gene was down-regulated, and KLF7 was unchanged.

To confirm the expression of key genes differentially regulated by the L-methionine treatment, the expression of the ten genes that were up-regulated (e.g., AFF3, AKR1C2, ATF3, C8ORF4, GPX8, LAMA3, RBM4B, NQO1, SH3BGRL, and VAMP5) was further assessed by real-time RT–PCR analysis, and their expression was compared with the changes in expression patterns detected by the micro-array analysis. There was strong induction of the mRNA expression of each of these genes after treatment with L-methionine using both microarray analysis and real-time RT–PCR (Table 1).

Discussion

In the present study, we identified, using microarray analysis, 98 genes whose expression was increased (10 genes) or decreased (88 genes) by treatment with L-methionine for 72 h at the non-apoptosis-inducing concentration of 5 mg/ml in both prostate cancer LNCaP cells and breast cancer MCF-7 cells. The expression of five genes was modified in opposite directions in these two cell lines. The dysregulation of these 98 genes in both breast and prostate cancer cells suggests that the affected genes are potential common regulatory targets of methionine and, possibly, methionine analogs. On the other hand, there were considerable differences between these two cell lines in the effect of methionine. Lmethionine caused significant expression changes of 842 genes in LNCaP cells, whereas only 221 genes were affected in MCF-7 cells. It is not clear why this difference occurred, but of note, MCF-7 cells are also less sensitive to inhibition of cell proliferation caused by methionine (Benavides et al. 2010).

Three of the genes up-regulated by L-methionine in both cell lines, NAD (P)H:quinone oxidoreductase (NQO1), SH3BGRL (SH3BGRL), and glutathione peroxidase 8 (GPX8), are associated with changes in cellular thiol redox balance and are involved cellular defense against oxidative stress (Forthoffer et al. 2002; Mazzocco et al. 2002; SantaCruz et al. 2004; Toppo et al. 2008; Yin et al. 2005). These findings suggests that L-methionine may induce antioxidant effects and consequently regulate the cellular pool of glutathione, which is required for maintaining the reduced state of cellular protein thiol groups (Metayer et al. 2008). It is conceivable that the induction of these antioxidant-related genes by Lmethionine may bring about increased antioxidant capacity in cancer cells. Of note, methionine also serves as a precursor of glutathione, a tripeptide that is a regulator of intracellular redox homeostasis, which by reducing a sulfhydryl-containing reactive oxygen species (ROS) safeguards cells from oxidative stress (Anderson 1998). NQO1 serves as a quinone reductase in connection with conjugation reactions of hydroquinones involved in detoxification pathways in addition to other functions (Bello et al. 2001). Reduced expression of NQO1 has been detected in breast cancer cells and is believed to be a strong prognostic and predictive factor in breast cancer (Fagerholm et al. 2008). SH3BGR has been suggested to represent a novel class of thioredoxin fold proteins belonging to the thioredoxin superfamily (Yin et al. 2005). GPX8 reduces hydrogen peroxide by transferring the energy of the reactive peroxides to a glutathione (Toppo et al. 2008).

Interestingly, also up-regulated was aldo-keto reductase family 1, member C2 (AKR1C2), which catalyzes the inactivation of the potent androgen 5-alpha-dihydrotestosterone (5 alpha-DHT) to 5-alpha-androstane-3-alpha, 17-beta-diol (3-alpha-diol), thereby possibly reducing androgen activity in LNCaP cells (Lou et al. 2006). The expression of AKR1C2 is known to be reduced in both prostate cancer and breast cancer cells (Ji et al. 2003, 2004).

A large number of genes that were down-regulated by L-methionine are members of protein kinase families. It is likely that many of these genes are associated with control of cell proliferation. Pathway analysis indicated down-regulation of a large number of genes involved in cancer, cell cycle, cell assembly and/or involved in organization, cell replication, recombination/repair of DNA, gastrointestinal disease, and genetic disorders by Lmethionine treatment. This could be consistent with the inhibitory effects of L-methionine on LNCaP and MCF-7 cell growth (Benavides et al. 2007; Benavides et al. 2010). On the other hand, no effects in gene groups associated with methionine metabolism specifically were detected by pathway analysis. Nevertheless, the array data generated in this study form the basis of future studies with multiple methionine doses and time points comparing not only breast and prostate cancer cells, but also cancer cells with non-tumorigenic cells from the same tissues. This is particularly important in view of the fact that methionine only inhibits cell cycle progression in breast and prostate cancer cells but not in non-tumorigenic breast and prostate epithelial cells (Benavides et al. 2010). Although such future hypothesis generating studies would also provide sufficient data to develop and test models that explore how methionine may selectively affect cancer cells, they should also focus on generating experimental evidence that the observed changes in expression of the genes have functional significance. For example, the potential modifying effects of methionine on antioxidant mechanisms would be one fruitful area of future investigation.

In summary, this study indicates that L-methionine induces common changes in molecular signatures of MCF-7 breast cancer cells and LNCaP prostate cancer cells, down-regulating genes belonging to protein kinase families, which may be related to the anti-proliferative effects of this amino acid on these cells. L-methionine also up-regulated some genes involved in cellular redox regulation suggesting antioxidant activity–enhancing properties of this amino acid. Future studies of the mechanisms and consequences of cellular and molecular effects of L-methionine and development of methionine analogs that lack the potential negative effects of methionine itself on the well-known methionine-dependence of many tumor cells (Judde et al. 1989) may eventually lead to exploitation of analogs of this amino acid in cancer therapy.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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References

- Anderson ME. Glutathione: an overview of biosynthesis and modulation. Chem Biol Interact 1998;111–112:1–14.
- Bello RI, Gomez-Diaz C, Navarro F, Alcain FJ, Villalba JM. Expression of NAD(P)H:quinone oxidoreductase 1 in HeLa cells: role of hydrogen peroxide and growth phase. J Biol Chem 2001;276:44379–44384. [PubMed: 11567026]
- Benavides MA, Oelschlager DK, Zhang HG, Stockard CR, Vital-Reyes VS, Katkoori VR, Manne U, Wang W, Bland KI, Grizzle WE. Methionine inhibits cellular growth dependent on the p53 status of cells. Am J Surg 2007;193:274–283. [PubMed: 17236862]
- Benavides MA, Hagen KL, Fang W, Du P, Lin S, Moyer MP, Yang W, Bland KI, Grizzle WE, Bosland MC. Suppression by L-methionine of cell cycle progression in LNCaP and MCF-7 cells but not benign cells. Anticancer Res. 2010 Accepted for publication.

- Dalle-Donne I, Rossi R, Colombo G, Giustarini D, Milzani A. Protein S-glutathionylation: a regulatory device from bacteria to humans. Trends Biochem Sci 2009;34:85–96. [PubMed: 19135374]
- Du P, Kibbe WA, Lin SM. nuID: a universal naming scheme of oligonucleotides for illumina, affymetrix, and other microarrays. Biol Direct 2007;2:16. [PubMed: 17540033]
- Fagerholm R, Hofstetter B, Tommiska J, Aaltonen K, Vrtel R, Syrjakoski K, Kallioniemi A, Kilpivaara O, Mannermaa A, Kosma VM, Uusitupa M, Eskelinen M, Kataja V, Aittomaki K, von Smitten K, Heikkila P, Lukas J, Holli K, Bartkova J, Blomqvist C, Bartek J, Nevanlinna H. NAD(P)H: quinone oxidoreductase 1 NQO1*2 genotype (P187S) is a strong prognostic and predictive factor in breast cancer. Nat Genet 2008;40:844–853. [PubMed: 18511948]
- Forthoffer N, Gomez-Diaz C, Bello RI, Buron MI, Martin SF, Rodriguez-Aguilera JC, Navas P, Villalba JM. A novel plasma membrane quinone reductase and NAD(P)H: quinone oxidoreductase 1 are upregulated by serum withdrawal in human promyelocytic HL-60 cells. J Bioenerg Biomembr 2002;34:209–219. [PubMed: 12171070]
- Gallagher S. Quantitation of nucleic acids with absorption spectroscopy. Curr Protoc Protein Sci. 2001 Appendix 4: Appendix 4K.
- Giustarini D, Rossi R, Milzani A, Colombo R, Dalle-Donne I. S-glutathionylation: from redox regulation of protein functions to human diseases. J Cell Mol Med 2004;8:201–212. [PubMed: 15256068]
- Horoszewicz JS, Leong SS, Kawinski E, Karr JP, Rosenthal H, Chu TM, Mirand EA, Murphy GP. LNCaP model of human prostatic carcinoma. Cancer Res 1983;43:1809–1818. [PubMed: 6831420]
- Huang E, Ishida S, Pittman J, Dressman H, Bild A, Kloos M, D'Amico M, Pestell RG, West M, Nevins JR. Gene expression phenotypic models that predict the activity of oncogenic pathways. Nat Genet 2003;34:226–230. [PubMed: 12754511]
- Ji Q, Chang L, VanDenBerg D, Stanczyk FZ, Stolz A. Selective reduction of AKR1C2 in prostate cancer and its role in DHT metabolism. Prostate 2003;54:275–289. [PubMed: 12539226]
- Ji Q, Aoyama C, Nien YD, Liu PI, Chen PK, Chang L, Stanczyk FZ, Stolz A. Selective loss of AKR1C1 and AKR1C2 in breast cancer and their potential effect on progesterone signaling. Cancer Res 2004;64:7610–7617. [PubMed: 15492289]
- Judde JG, Ellis M, Frost P. Biochemical analysis of the role of transmethylation in the methionine dependence of tumor cells. Cancer Res 1989;49:4859–4865. [PubMed: 2503245]
- Lin SM, Du P, Huber W, Kibbe WA. Model-based variance-stabilizing transformation for Illumina microarray data. Nucleic Acids Res 2008;36:e11. [PubMed: 18178591]
- Lou H, Du S, Ji Q, Stolz A. Induction of AKR1C2 by phase II inducers: identification of a distal consensus antioxidant response element regulated by NRF2. Mol Pharmacol 2006;69:1662–1672. [PubMed: 16478829]
- Mazzocco M, Maffei M, Egeo A, Vergano A, Arrigo P, Di Lisi R, Ghiotto F, Scartezzini P. The identification of a novel human homologue of the SH3 binding glutamic acid-rich (SH3BGR) gene establishes a new family of highly conserved small proteins related to thioredoxin superfamily. Gene 2002;291:233–239. [PubMed: 12095696]
- Metayer S, Seiliez I, Collin A, Duchene S, Mercier Y, Geraert PA, Tesseraud S. Mechanisms through which sulfur amino acids control protein metabolism and oxidative status. J Nutr Biochem 2008;19:207–215. [PubMed: 17707628]
- Pratt SE, Pollak MN. Estrogen and antiestrogen modulation of MCF7 human breast cancer cell proliferation is associated with specific alterations in accumulation of insulin-like growth factorbinding proteins in conditioned media. Cancer Res 1993;53:5193–5198. [PubMed: 7693333]
- SantaCruz KS, Yazlovitskaya E, Collins J, Johnson J, DeCarli C. Regional NAD(P)H: quinone oxidoreductase activity in Alzheimer's disease. Neurobiol Aging 2004;25:63–69. [PubMed: 14675732]
- Shrivastava A, Nunn AD, Tweedle MF. Designer peptides: learning from nature. Curr Pharm Des 2009;15:675–681. [PubMed: 19199988]

- Smyth G. Linear models and empirical Bayes methods for assessing differential expression in microarray experiments. Stat Appl Genet Mol Biol. 2004 3: Article 3.
- Sweet-Cordero A, Mukherjee S, Subramanian A, You H, Roix JJ, Ladd-Acosta C, Mesirov J, Golub TR, Jacks T. An oncogenic KRAS2 expression signature identified by cross-species geneexpression analysis. Nat Genet 2005;37:48–55. [PubMed: 15608639]
- Takahashi K, Suzuki K. Association of insulin-like growth-factor-I-induced DNA synthesis with phosphorylation and nuclear exclusion of p53 in human breast cancer MCF-7 cells. Int J Cancer 1993;55:453–458. [PubMed: 8375929]
- Thomas T, Thomas TJ. Polyamines in cell growth and cell death: molecular mechanisms and therapeutic applications. Cell Mol Life Sci 2001;58:244–258. [PubMed: 11289306]
- Toppo S, Vanin S, Bosello V, Tosatto SC. Evolutionary and structural insights into the multifaceted glutathione peroxidase (Gpx) superfamily. Antioxid Redox Signal 2008;10:1501–1514. [PubMed: 18498225]
- Vazquez E, Ferrer-Miralles N, Mangues R, Corchero JL, Schwartz S Jr, Villaverde A. Modular protein engineering in emerging cancer therapies. Curr Pharm Des 2009;15:893–916. [PubMed: 19275653]
- Yin L, Xiang Y, Zhu DY, Yan N, Huang RH, Zhang Y, Wang DC. Crystal structure of human SH3BGRL protein: the first structure of the human SH3BGR family representing a novel class of thioredoxin fold proteins. Proteins 2005;61:213–216. [PubMed: 16080146]

Fig. 1.

Heatmap showing the overall expression patterns of all genes differentially expressed in either LNCaP and MCF-7 cells, comparing methionine-treated (*Pos*) and untreated control (*Neg*) cells in three independent replicate experiments (1, 2, and 3). Each row represents one sample and each column one probe. The expression profile of each probe was first standardized (zero mean and one standard deviation). The *red color* represents higher than average, *green color* represents lower than average, and *black* means close to the average

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Fig. 2.

Volcano plots in which each point represents the expression of a gene plotted as a function of fold change (Log2 (fold change), *x*-axis) after methionine exposure compared to untreated controls (*Pos–Neg*) and the statistical significance (−Log 10 (*P*-value), *y*-axis). *Vertical dotted lines* represent fold changes of ± 1.5, respectively. The *horizontal dotted line* represents an FDR of 0.01. The *red dots* represent differentially expressed genes with a FDR < 0.01 and fold change > 1.5

Fig. 3.

Heatmap of genes the expression of which was altered by L-methionine treatment in both LNCaP and MCF-7 cells, comparing methionine-treated (*Pos*) and untreated control (*Neg*) cells in three independent experiments (1, 2, and 3). Of these 103 genes, 98 were regulated in the same direction by this treatment, with 10 genes up-regulated and 88 genes downregulated in both cancer cell lines. The expression of five other genes was changed in opposite directions in the two cell lines (see text of results)

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Table 1

Primer sequences used and results of confirmative quantitative RT-PCR analysis of 10 genes that were observed to be up-regulated by L-methionine Primer sequences used and results of confirmative quantitative RT–PCR analysis of 10 genes that were observed to be up-regulated by L-methionine

Table 2

Comparisons between methionine-treated and control LNCaP and MCF-7 cells revealing that the expression of 98 genes was changed in the same Comparisons between methionine-treated and control LNCaP and MCF-7 cells revealing that the expression of 98 genes was changed in the same direction in both cell lines based on the following criteria: a false discovery rate (FDR)-adjusted P-value <0.01 and a fold change >1.5 *P*-value <0.01 and a fold change >1.5 direction in both cell lines based on the following criteria: a false discovery rate (FDR)–adjusted

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2.48E-03

 -1.78 6.50E-05

 -2.82 1.78E-07 1.22E-05

52 BepQfvreD3XB1P_Ank 7272 TTK TTK protein kinase −2.82 1.78E-07 1.22E-05 −1.78 6.50E-05 2.48E-03

TTK protein kinase

TTK

7272

 $\operatorname{\mathsf{BepQf}vreD3XBIP_Ank}$

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 $MCF-7$

 $LMCaP$

Description

EntrezID Symbol

No. Probe.NuID

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Table 3

Classes of L-methionine-responsive gene signatures and their top functions

The genes were classified based on molecular networks (www.Ingenuity.com; see text). The downward arrows indicate genes that were downregulated by L-methionine exposure in both LNCaP and MCF-7 cells, and the upward arrows indicate genes that were up-regulated in both cell lines. The expression of genes indicated without arrows and not in bold-face in these networks was unchanged in response to L-methionine treatment; the expression of four of these genes was changed in only one of the two cell lines (NCAPD3, UBE2A, ADAM15, and C4ORF43). Ingenuity Pathways Analysis computes a score for each network according to the fit of that network to the user-defined set of Focus Genes. This score is derived from a *P*-value and indicates the likelihood of the Focus Genes in a particular network being found together due to random chance. A score of 2 indicates that there is a 1 in 100 chance that the Focus Genes are together in a network due to random chance. Therefore, scores of 2 or higher have at least a 99% confidence of not being generated by random chance alone. This score is given in the third column of this table, and the number of focus genes that were changed in expression is given in the fourth column