**In silico** identification of transcriptional regulators associated with c-Myc

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Received July 22, 2004; Revised and Accepted August 21, 2004

**ABSTRACT**

The development of powerful experimental strategies for functional genomics and accompanying computational tools has brought major advances in the delineation of transcriptional networks in organisms ranging from yeast to human. Regulation of transcription of eukaryotic genes is to a large extent combinatorial. Here, we used an *in silico* approach to identify transcription factors (TFs) that form recurring regulatory modules with c-Myc, a protein encoded by an oncogene that is frequently disregulated in human malignancies. A recent study identified, on a genomic scale, human genes whose promoters are bound by c-Myc and its heterodimer partner Max in Burkitt’s lymphoma cells. Using computational methods, we identified nine TFs whose binding-site signatures are highly overrepresented in this promoter set of c-Myc targets, pointing to possible functional links between these TFs and c-Myc. Binding sites of most of these TFs are also enriched on the set of mouse homolog promoters, suggesting functional conservation. Among the enriched TFs, there are several regulators known to control cell cycle progression. Another TF in this set, EGR-1, is rapidly activated by numerous stress challenges and plays a central role in angiogenesis. Experimental investigation confirmed that c-Myc and EGR-1 bind together on several target promoters. The approach applied here is general and demonstrates how computational analysis of functional genomics experiments can identify novel modules in complex networks of transcriptional regulation.

**INTRODUCTION**

With the completion of the sequencing of the human genome and genomes of many other organisms, research efforts are shifting to functional genomics, i.e. deciphering how components of physiological networks interact and regulate each other in cellular systems. The most notable achievements to date have been the delineation of transcriptional regulatory networks, which became feasible with the advent of genome-wide experimental techniques that specifically shed light on this layer of cellular systems (1–4). These techniques include gene expression microarrays and the combined strategy of chromatin immunoprecipitation (ChIP) and promoter microarrays (also termed ‘ChIP-on-chip’). The former enables the simultaneous recording of expression levels of thousands of genes (5), and the latter enables identification on a genomic scale of promoters that are bound by specific transcription factors (TFs) under certain conditions, in a single experimental assay (6,7). These novel techniques generate huge volumes of biological data, the exploitation of which is largely dependent on the development of appropriate computational tools.

The ChIP-on-chip approach was recently used to map global TF-promoter binding relationships in yeast under standard growth conditions (6). It was also used in mammalian cells to identify genome-wide direct targets of multiple TFs (8–10). Seeking direct targets of c-Myc, Li *et al.* (8) identified 776 human genes in Burkitt’s lymphoma cells whose promoters are bound by the oncoprotein c-Myc and its heterodimer partner Max. c-Myc regulates cell-cycle proliferation, apoptosis and differentiation. Overexpression of c-Myc is one of the most common alterations in human cancer, yet it is not clear how it promotes malignant transformation (11–14). It is widely accepted that the transcriptional regulation activity of c-Myc is critical for the development of malignancy associated with it, but the target genes that mediate this process remain elusive.

Regulation of transcription of eukaryotic genes is to a large extent combinatorial, i.e. the conditions under which a gene is expressed are determined using an intricate interplay of multiple positive and negative transcriptional regulators that recognize and bind to *cis*-regulatory elements within and beyond the gene’s promoter region. Therefore, a major goal in deciphering transcriptional regulation networks is to identify combinations of TFs that functionally cooperate to form regulatory modules. Such modules can be recognized by the co-occurrence of the corresponding TF-binding sites in the same promoters (1,15,16). Our aim in this work is to identify TFs that form recurrent *cis*-regulatory modules with c-Myc. To this end, we computationally analyzed the promoters...
reported by Li et al. (8) to be bound by c-Myc. We report on nine TFs whose binding signatures are significantly over-represented in this set of promoters, including several TFs known to play important roles in the regulation of cell cycle progression. The results point to possible functional links between these TFs and c-Myc, and offer an additional explanation for the positive effect of c-Myc on cellular proliferation. Another TF whose binding signature was highly enriched in the c-Myc target promoters is EGR-1, which is rapidly activated by many types of stress, including hypoxia, DNA damage and vascular injury, and has a central role in angiogenesis (17,18). We demonstrate experimentally that c-Myc and EGR-1 bind together several common target promoters. The approach applied here is general and demonstrates the potential of computational promoter analysis of ChIP-on-chip data for the identification of novel transcriptional modules.

MATERIALS AND METHODS

Extraction of putative promoter sequences from human and mouse genomes

Putative promoter sequences corresponding to all known human and mouse genes were extracted from the human and mouse genome sequences (Ensembl, version 13, December 2002) using a Perl script based on the application programming interface provided by the Ensembl project (19). For each gene flagged as a ‘known gene’, a genomic sequence was extracted that spanned 1000 bp upstream to 200 bp downstream of the gene’s putative transcription start site (TSS). The extracted sequences were masked for repetitive elements. A total of 19351 and 18748 putative promoters were extracted from the human and mouse genome, respectively. The sequences of these promoter sets can be downloaded from http://www.cs.tau.ac.il/~rshamir/prima/PRIMA.htm.

To avoid biases due to highly similar promoters, we constructed for each organism a non-redundant set of promoters by running all-against-all BLAST comparisons. For every promoter pair with BLAST E-score <10⁻⁵⁰, we excluded one of the members from the non-redundant set. The non-redundant human and mouse promoter sets contain 17390 and 15521 promoters, respectively.

Computational identification of enriched TFs

The computational analysis used PRIMA software that is described in detail in (16) and is available at http://www.cs.tau.ac.il/~rshamir/prima/PRIMA.htm. In short, given target and background sets of promoters, PRIMA performs statistical tests aimed at identifying TFs whose binding sites are significantly more abundant in the target set than in the background set. PRIMA uses position weight matrices (PWMs) as models for regulatory sites that are bound by TFs. A total of 300 PWMs that represent human or mouse TF-binding sites were obtained from the TRANSFAC database (20). The entire collection of non-redundant human promoters was used as the background set. Human–mouse homolog genes were determined using the EnsMart utility provided by Ensembl (21). Default parameters were used in all PRIMA runs. The program scanned both strands of each promoter for putative binding sites, i.e. for sites with high similarity score to the PWM. Throughout the text, we term such high scoring sites as ‘hits’ of the corresponding TF. Thresholds of similarity scores for declaring hits correspond to an average rate of one hit per 10,000 bp of random promoter sequences (16).

ChIP assays

The cell line used for all ChIP analyses was the human myelogenous leukemia K562 cells. The cells were grown in RPMI 1640 medium supplemented with 10% fetal bovine serum and antibiotics. The cells growing in log phase were formaldehyde cross-linked and chromatin was immunoprecipitated as described previously (22). An aliquot of 2 µg of rabbit polyclonal c-Myc and HGF antibodies (anti-Myc sc-764, anti-HGF sc-7949; Santa Cruz Biotechnology, Inc., Santa Cruz, CA) and 5 µg of rabbit polyclonal Egr-1 antibody (sc-189; Santa Cruz Biotechnology, Inc.) were used to precipitate chromatin from 2 × 10⁷ cells.

Real-time PCR

For real-time PCR, a SYBR green core reagents kit (Applied Biosystems) was utilized. Known quantities of total input DNA were used to generate a standard curve for determining the percentage of total input for each ChIP sample. All amplifications were carried out in the linear range of standard curves.

RESULTS

Enriched TF signatures in promoters bound by c-Myc/Max

The data set recently published by Li et al. (8) contained 876 human promoters that were bound by c-Myc, 931 promoters that were bound by Max and 776 promoters that were bound by both TFs. To reduce false positive discoveries, we focused on the set of promoters that were bound by both c-Myc and Max, hereafter referred to as the c-Myc/Max target set. As a first step, we extracted from the human genome sequences that correspond to promoters of all known genes (for details see Materials and Methods), limiting the analysis to sequences spanning 1000 bp upstream to 200 bp downstream of the putative TSSs of the corresponding genes. We determined the region of sequence around the putative TSS in which to search for transcriptional regulatory elements by examining the location distribution of 1075 empirically validated TF-binding sites in human promoters [data from TRANSFAC database (20)]. Eighty percentage of these elements were located within the region of our analysis (data not shown). Clearly, present experimental data on TF-binding sites is biased toward binding sites located at short distances from the TSS. Although certain regulatory elements were demonstrated to act over great distances, up to several kilobases from the TSS, it is clear that ample information on mechanisms of transcriptional regulation resides in sequences in close proximity to the TSS.
Our collection of human promoter sequences contains a total of 19351 promoters, of which a subset of 17390 promoters is non-redundant (see Materials and Methods). The full and non-redundant promoter sets include sequence data for 615 and 519 genes, respectively, out of the 776 genes in the Myc/Max target set. We applied PRIMA, a promoter analysis software that we developed recently (16), to scan the 519 non-redundant promoters of the Myc/Max target set for overrepresented TF-binding-site signatures. This means that we searched for TFs whose binding-site signatures are significantly more abundant on this promoter set than expected by chance, given their abundance on the entire collection of non-redundant human promoters (for details on PRIMA see Materials and Methods). Such overrepresentation suggests the existence of functional links between the over-abundant TFs and c-Myc/Max. PRIMA uses PWMs as models of regulatory sites that are bound by TFs. A total of 300 PWMs that represent mammalian TF-binding sites (obtained from the TRANSFAC database) were tested, and nine of them were identified by PRIMA to be significantly enriched in the c-Myc/Max target set \( P < 10^{-5} \), and after conservative adjustment for multiple testing \( P < 0.05 \); Table 1, (A).

Previously, we used PRIMA to analyze human genes whose expression is cell-cycle-dependent (16). Interestingly, most of the TFs whose signatures were enriched in the cell-cycle-dependent promoter set were also enriched in the c-Myc/Max target set (E2F, NF-Y, Sp1, Nrf1, ETF, CREB and AhR/Arnt). Therefore, it was suggested that the c-Myc/Max binds its target promoters directly. As noted above, a canonical E-box motif (CAGTG), which is directly recognized and bound by the c-Myc/Max heterodimer. In accordance with this observation, PRIMA found that PWMs with canonical E-box core motifs are only slightly enriched on the Myc/Max target set, and they did not pass our significance threshold. However, another PWM that models c-Myc/Max binding sites, whose core motif is a variant of the E-box \((\text{CAYGYY,} \ Y = \{C,T\})\), was significantly enriched on this target set (PWM M00322 in TRANSFAC DB) [Table 1, (A)]. It cannot be ruled out that we failed to detect the E-box overrepresentation as a result of the fact that a fraction of these regulatory elements are located outside the analyzed promoter region.

### Table 1. TF-binding site signature enrichments in c-Myc/Max target sets

<table>
<thead>
<tr>
<th>TF [PWM accession ID in TRANSFAC DB]</th>
<th>(A) ( P )-value for PWM hits' abundance on c-Myc/Max target set</th>
<th>(B) ( P )-value for PWM hits' abundance on mouse ortholog set</th>
<th>(C) ( P )-value for PWM hits' abundance on mode 1 subset</th>
</tr>
</thead>
<tbody>
<tr>
<td>ETF [M00695]</td>
<td>(1.2 \times 10^{-15})</td>
<td>(6.8 \times 10^{-13})</td>
<td>(3.2 \times 10^{-7})</td>
</tr>
<tr>
<td>Sp1 [M00196]</td>
<td>(1.7 \times 10^{-14})</td>
<td>(8.9 \times 10^{-12})</td>
<td>(6.3 \times 10^{-10})</td>
</tr>
<tr>
<td>Nrf1 [M00652]</td>
<td>(6.5 \times 10^{-14})</td>
<td>(7.9 \times 10^{-11})</td>
<td>(3.2 \times 10^{-7})</td>
</tr>
<tr>
<td>NF-Y [M00185]</td>
<td>(3.2 \times 10^{-12})</td>
<td>(1.4 \times 10^{-7})</td>
<td>(1.5 \times 10^{-5})</td>
</tr>
<tr>
<td>CREB [M00177]</td>
<td>(4.7 \times 10^{-8})</td>
<td>Not enriched</td>
<td>Not enriched</td>
</tr>
<tr>
<td>c-Myc/Max [M00322]</td>
<td>(1.5 \times 10^{-7})</td>
<td>(2.8 \times 10^{-6})</td>
<td>(5.7 \times 10^{-6})</td>
</tr>
<tr>
<td>Egr-1 [M00243]</td>
<td>(2.4 \times 10^{-7})</td>
<td>(6.7 \times 10^{-5})</td>
<td>(3.4 \times 10^{-8})</td>
</tr>
<tr>
<td>Elk-1 [M00025]</td>
<td>(3.9 \times 10^{-7})</td>
<td>(5.5 \times 10^{-11})</td>
<td>Not enriched</td>
</tr>
<tr>
<td>E2F [M00516]</td>
<td>(5.8 \times 10^{-7})</td>
<td>(6.6 \times 10^{-5})</td>
<td>Not enriched</td>
</tr>
<tr>
<td>Ahr/Arnt [M00237]</td>
<td>(6.4 \times 10^{-7})</td>
<td>Not enriched</td>
<td>Not enriched</td>
</tr>
<tr>
<td>E-box</td>
<td>(8.4 \times 10^{-4})</td>
<td>(4.1 \times 10^{-5})</td>
<td>(6.5 \times 10^{-7})</td>
</tr>
</tbody>
</table>

\( ^{c}\text{-Myc/Max signature is markedly enriched on ‘mode 1’ subset, since by definition this set comprises the genes whose promoters were found to have a hit for c-Myc/Max. Not enriched indicates } P > 0.0001 (\text{P-values are before correction for multiple testing}). \)

### Conservation of TF-binding-site enrichment on mouse ortholog promoters

The enrichment of binding-site signatures of specific TFs in the c-Myc/Max target set raises the possibility that c-Myc/Max and the TFs maintain functional relationships and together form recurrent transcriptional regulation modules that control the expression of numerous genes. In order to strengthen this \textit{in silico}-derived hypothesis, we repeated the tests for TF-binding site enrichment, this time on the set of promoters comprising the mouse orthologs of the human c-Myc/Max target set. Extracting promoter sequences for all known mouse genes, we collected 18478 mouse promoter sequences, of which 15521 were non-redundant. The non-redundant set includes sequence data for 407 mouse orthologs of the human Myc/Max target set. Applying PRIMA to this set, we found that seven out of the nine PWMs that were enriched in the human set were also enriched in the mouse set [Table 1, (B)]. The enrichment of the same TF-binding sites on the ortholog sets suggests that the functional relationships between these TFs are conserved between mice and humans.

### TFs associated with c-Myc/Max through direct DNA binding

As noted above, a canonical E-box is found in only \sim 25% of the promoters identified by Li \textit{et al} (8) as direct targets of c-Myc/Max. Therefore, it was suggested that the c-Myc/Max heterodimer controls its target by two different modes: in the first one, it directly binds its target promoters through its classical E-box element or a variant thereof; and in the second mode, it participates in the regulation of target promoters...
without binding directly to the DNA, but by physically interacting with other sequence-specific TFs or with components of the general transcription machinery.

We attempted to identify TFs that form cis-regulatory modules with c-Myc/Max via the first mode. To this end, we identified a subset of the c-Myc/Max target set, comprising genes whose promoters contain high-scoring putative binding sites (hits) for c-Myc/Max. As the c-Myc/Max variant binding element represented by PWM M00322 was more enriched than the canonical E-box, we scanned the c-Myc/Max target set for promoters with hits for this PWM. We identified hits for M00322 in 134 out of 615 promoters, which were good candidates for being regulated by c-Myc/Max through its direct DNA binding (Supplementary Table A). We refer to this subset as 'c-Myc/Max mode1 subset'. Since it is smaller than the complete c-Myc/Max target set, we expected statistical phenomena associated with it to be less significant. However, we observed that one TF, EGR-1, was even more significantly enriched in the mode1 subset than in the complete set [Table 1, (C)]. This makes EGR-1 a strong candidate for being c-Myc/Max’s partner in the regulation of mode1 target promoters. Of note, E2F, a pivotal regulator of cell cycle progression, was not enriched in the mode1 subset, suggesting that the c-Myc/Max cooperation with E2F is maintained mainly through mode 2.

### EGR-1 binding site enrichment in the GC-rich c-Myc/Max mode1 subset

As noted above, the EGR-1 binding site signature was more enriched in the c-Myc/Max mode1 subset than in the complete c-Myc/Max set. Inspection of EGR-1 PWM (M00243) showed that it has a high-GC-content with a GCGTGGG core. This led us to compare the GC-content of the mode1 subset, the c-Myc/Max target set and the full collection of non-redundant human promoters (Table 2). We observed that the c-Myc/Max target set is more GC-rich than the full set of human promoters (57.1% vs. 57.0%, Z-score = 9.4), and that the c-Myc/Max mode1 subset is even more GC-rich (60.1%, Z-score = 5.8) compared to GC-content of the c-Myc/Max target set). This raised the question whether the abundance of hits for EGR-1 on the mode1 subset could be explained merely by its high-GC-content. To address this concern, we generated random PWMs based on the EGR-1 PWM in a way that preserved its GC-content. We generated the random PWMs by permuting the columns of the original PWM, and randomly interchanging A with T and G with C. We then compared the enrichment of five permuted and the original EGR-1 PWM on the c-Myc/Max mode1 subset. Significantly, the original EGR-1 PWM yielded an enrichment score that was far more significant than the scores obtained by the random PWMs generated based on it (Table 3). For further examination, we sorted all TRANSFAC mammalian PWMs according to their GC-content and recorded their enrichment in the mode1 subset. Supplementary Table B lists the results for the top 25 GC-rich PWMs. This list shows that the over-representation of EGR-1 in the mode1 subset is not a mere reflection of its high-GC-content, as there are many PWMs at least as GC-rich as EGR-1 that are not at all enriched on this subset.

### Experimental confirmation for co-binding of Myc and EGR-1 in selected promoters

A total of 54 promoters of the c-Myc/Max mode1 subset contained strong hits for EGR-1. To experimentally test our in silico derived hypothesis that these two TFs together form a recurrent transcriptional cis-regulatory module, we selected six of these genes and examined the binding of both c-Myc and EGR-1 to their promoters using ChIP assays. For all six genes examined (RAP2B, KHSRP, PolH, PTPN1, PP and KPNNA3), the signals obtained for both c-Myc and EGR-1 were above the background level (P < 0.025, one-tailed t-test; Figure 2). For a negative control, we chose from the c-Myc/Max target set the MCC2 gene, whose promoter did not contain any hit for EGR-1. In agreement, the signal obtained by the ChIP assay for EGR-1 binding to this promoter was very close to background level. These results demonstrate that c-Myc and EGR-1 co-binding occurs on multiple promoters. Further experiments will be required to establish direct functional link between these two transcriptional regulators.

### Tables

#### Table 2. High-GC-content of the c-Myc/max mode1 subset

<table>
<thead>
<tr>
<th>Promoter set</th>
<th>Number of promoters</th>
<th>%A</th>
<th>%C</th>
<th>%G</th>
<th>%T</th>
</tr>
</thead>
<tbody>
<tr>
<td>Non-redundant human promoters</td>
<td>17 390</td>
<td>23.1</td>
<td>26.4</td>
<td>27.0</td>
<td>23.5</td>
</tr>
<tr>
<td>c-Myc/Max target set</td>
<td>615</td>
<td>21.5</td>
<td>28.3</td>
<td>28.7</td>
<td>21.6</td>
</tr>
<tr>
<td>c-Myc/Max mode1 subset</td>
<td>134</td>
<td>19.5</td>
<td>30.4</td>
<td>30.5</td>
<td>19.7</td>
</tr>
</tbody>
</table>

#### Table 3. Enrichment of the original and permuted EGR-1 PWM on the c-Myc/Max mode1 subset

| PWM | EGR-1 | Rand1 | EGR-1 | Rand2 | EGR-1 | Rand3 | EGR-1 | Rand4 | EGR-1 | Rand5 |
|-----|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| Enrichment score | $4.5 \times 10^{-7}$ | 0.17 | 0.28 | 0.25 | 0.088 | 0.0034 |

High-scoring hits for c-Myc/Max (M00322) and EGR-1 (M00243) were found in 134 and 167 promoters, respectively, out of a total of 615 promoters in the c-Myc/Max target set. A total of 54 promoters contained strong hits for both c-Myc/Max and EGR-1 (Supplementary Table C). Finally, we examined the location distribution of c-Myc/Max and EGR-1 hits on the promoters of the c-Myc/Max target set. For both TFs, the computationally identified binding sites are significantly concentrated in the proximity of the TSS and their density drops downstream of it (Figure 1). In contrast, the distribution of hits identified for a random PWM generated by permuting the EGR-1 PWM was quite uniform between −650 and +200 bp with respect to the TSS, as expected for random hits. The fact that the hit distributions for c-Myc/Max and EGR-1 show a prominent peak in the anticipated position is an additional indication of the quality of the human genome TSS annotations (such peaks would have not been obtained if significant deviations had existed between the locations of the annotated and real TSS in a large proportion of the genes), and for the specificity of the hits identified by PRIMA (high false positive rates for hits identified by PRIMA would have obscured the peak of true hits).
DISCUSSION

The mechanisms by which c-Myc promotes cell growth and transformation are poorly understood, but several lines of evidence suggest that its function as transcriptional regulator is required for its ability to promote malignancy. Since transcriptional regulation is often the result of combinatorial interplay among various TFs, we attempted to identify TFs that form recurring cis-regulatory modules with c-Myc. To this end, we analyzed a publicly available dataset of human promoters that are bound by c-Myc/Max in Burkitt’s lymphoma cells. Our computational analysis identified nine TFs whose binding-site signatures are significantly overrepresented in this set of promoters. Most of these TFs were also enriched in the set of mouse orthologs of c-Myc target promoters, suggesting a conserved biological function mediated by the cooperation of c-Myc and these TFs.

Among the TFs enriched in the c-Myc/Max target promoter set were the pivotal regulators of the transcriptional program associated with cell cycle progression, E2F and NF-Y. The enrichment of the binding-site signatures of these factors in promoters that are bound by c-Myc/Max suggests functional links between them and c-Myc. Functional links between c-Myc and E2F are well documented (23–26). Myc promotes the cell cycle progression by coordinate activation of cell cycle driving genes (e.g. Cdc25A, Cdk4, and Cyclins D2, E and A), and by the suppression of cell cycle arrest genes (such as p15, p21, p27 and GADDs) (24). Some of the cell cycle promoting genes are common targets of c-Myc and members of the E2F family TFs (27). Recently, a physical link between c-Myc and E2F-6 was reported, with a role in maintaining cellular quiescence (28). As for the link between c-Myc and NF-Y, physical interaction between c-Myc and the NF-YB and NF-YC subunits of the NF-Y trimer has been demonstrated previously (29), and in particular, these interactions are important for the down-regulation of a well-established c-Myc target, PDGFRb (30). Computational identification of cis-modules cannot decipher the regulatory effect of the module. The finding on the negative effect of the c-Myc/NF-Y module on PDGFRb transcription underscores the importance of further studies, which are beyond the scope of our current study, to establish whether the functional interplay between predicted TFs with c-Myc results in negative or positive regulation of target genes.

In addition to the examples outlined above, we sought to determine whether EGR-1 binds promoters, which are predicted in silico to contain EGR-1 consensus sites along with c-Myc binding sites. Given that promoter-reporter transient transfection assays, with specific reference to c-Myc, are inherently noisy and dependent on the cell types as well as cell density and other factors, we chose ChIP assays to determine whether our in silico prediction of EGR1 binding could be validated. The experimental data substantiate the prediction that these two TFs together form a transcriptional cis-regulatory module that recurs on multiple promoters.
Figure 2. ChIP of c-Myc and EGR-1 targets in K562 cells. Each graph represents real-time PCR amplification of the promoter and control regions of each gene using anti-Myc, anti-EGR-1, anti-HGF and no antibody precipitated chromatin as template. Bars represent the percentage of total input DNA for each ChIP sample averaged over three PCRs. Error bars represent 1 SD. The horizontal solid line represents 0.02% total input DNA, the background signal for this assay. The signals obtained for the binding of c-Myc and EGR-1 to the promoter regions of all six genes examined (RAP2B, KHSRP, PolH, PTPN1, PP and KPNA3), but not for the negative control, MCCC2, were above the background level ($P < 0.025$, one-tailed $t$-test). Primer sequences and positions, and locations of the putative hits for c-Myc and EGR-1 on the examined promoters are given in Supplementary Table D.
The possibility for functional links between c-Myc and EGR-1 is intriguing because of the pivotal role of EGR-1 in angiogenesis, the formation of new blood vessels from pre-existing vasculature. Uncontrolled angiogenesis plays an important role in tumor growth, and the sprouting of new blood vessels into tumors suggests that angiogenesis is necessary for the progression of malignancy. Recent reports underscored the critical roles of both EGR-1 and c-Myc in angiogenesis, Fahmy et al. (17) reported that the inhibition of EGR-1 expression repressed neovascularization and blocked angiogenesis and tumor growth in mouse and rat models. Baudino et al. (31) reported that c-Myc is also required for the proper expression of several major angiogenic factors, and that c-Myc(−/−) ES cells are dramatically impaired in their ability to form tumors in immune-compromised mice, and the small tumors that do develop are poorly vascularized. Here, we proposed a possible synergistic action between c-Myc and EGR-1 in transcriptional regulation of the target genes, and experimentally demonstrated the binding of both c-Myc and EGR-1 to several target promoters.

The approach applied here is general, and demonstrates the power of computational analysis to elucidate novel functional links in transcriptional networks, based on functional genomics data.

SUPPLEMENTARY MATERIAL

Supplementary Material is available at NAR Online.

ACKNOWLEDGEMENTS

R.E. is a Joseph Sasson Fellow. This work was partially supported by a research grant from the Ministry of Science and Technology, Israel (Y.S. and R.S.) and NIH grants 2 R01CA057341 (C.V.D.) and T32HL07525 (K.I.Z.). This work was carried out in partial fulfillment of the requirements for the PhD degree of R.E.

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