

Meta-Analysis of Thyroid Cancer Expression Profiling Studies

Identification of Most Promising Biomarkers For Tissue Microarray Analysis

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1. Abstract

Objective and Design

It is estimated that 5-10% of the population will develop a clinically significant thyroid nodule during their lifetime. In one third or more of these patients, pre-operative diagnoses by needle biopsy are inconclusive. In many cases, a patient will undergo unnecessary surgery for what ultimately proves to be a benign lesion. Thus, there is a clear need for improved diagnostic tests to distinguish malignant from benign samples. The recent development of tissue microarray techniques should allow the rapid evaluation of potential new markers. However, researchers are faced with an overwhelming number of potential markers from numerous thyroid cancer profiling and classification studies. We present a systematic and comprehensive selection of potential thyroid cancer biomarkers from published studies by meta-analysis for use in tissue microarray analysis (TMA).

Materials & Methods

A total of 21 published studies were identified from the literature. Each study reported differentially expressed genes for at least one comparison type (e.g. Normal versus PTC). The following information was recorded wherever possible: unique identifier (probe/ID/accession), gene name, gene description, gene symbol, tissue types compared, fold change (magnitude and/or direction), p-value, validation (RT-PCR, IHC, Western), and PubMed ID. Wherever possible, the mapping of clone accession, probe ID or SAGE tag was updated using NCBI mapping files, Affymetrix annotation files, and the DiscoverSpace SAGE tag mapping tool respectively. A heuristic system was devised to identify the most promising markers, taking into consideration the number of studies reporting the potential marker, sample sizes and fold-change.

Results

In total, 1,785 potential biomarkers were identified (not considering overlap) from 21 gene expression studies considering 34 different tumour or tissue type comparisons. This resource allows for identification of markers that consistently differentiate one tumour/tissue type (e.g. papillary thyroid cancer) from others (e.g. normal, benign, other thyroid cancer subtypes). From this analysis, an informed selection of markers will be made for tissue microarray analysis, optimizing the chance of finding clinically relevant markers.

Conclusion

Bioinformatics meta-analysis and tissue microarray analysis represents a powerful approach to identifying new thyroid cancer biomarkers. Such markers could prove invaluable for the diagnosis and prognosis of tumours in clinical testing. This meta-analysis of published thyroid studies should prove a useful resource for many thyroid cancer researchers.

2. Methods

Figure 1. Analysis methods

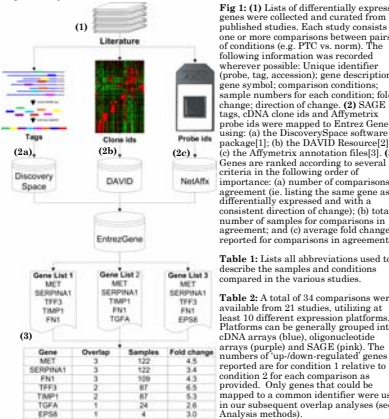


Fig 1: (1) Lists of differentially expressed genes were collected and curated from published studies. Each study consisted of one or more comparisons between pairs of conditions (e.g. PTC vs. norm). The following information was recorded wherever possible: Unique identifier (probe, tag, accession); gene description; gene symbol; comparison conditions; sample numbers for each condition; fold change; direction of change; (2) SAGE tags, cDNA clone IDs and Affymetrix probe IDs were mapped to Entrez Gene using: (a) the DiscoverSpace software package[1]; (b) the DAVID Resource[2]; (3) Genes are ranked according to several criteria in the following order of importance: (a) number of comparisons in agreement (ie. listing the same gene as differentially expressed with a consistent direction of change); (b) total number of samples for comparisons in agreement; and (c) average fold change reported for comparisons in agreement.

Table 1: Lists all abbreviations used to describe the samples and conditions compared in the various studies.

Table 2: A total of 34 comparisons were available from 21 studies, utilizing at least 10 different expression platforms. Platforms can be generally grouped into cDNA array (blue), oligonucleotide arrays (purple) and SAGE (pink). The numbers of 'up/down-regulated' genes reported are for condition 1 relative to condition 2 for each comparison as provided. Only genes that could be mapped to a common identifier were used in the meta-analysis (see Analysis methods).

Table 1. Abbreviations for sample descriptions

ANCL	Anaplastic thyroid cancer cell line
APFN	Autonomously functioning thyroid nodules
CTN	Cancer
CTN	Cold thyroid nodule
FA	Follicular adenoma
PCL	Follicular carcinoma cell line
PTO	Follicular thyroid carcinoma
PVPTC	Papillary variant papillary carcinoma
CTN	Cancer
HCC	Hurthle cell carcinoma
HN	Hypertrophic nodule
M	Metastatic
MACL	Anaplastic thyroid cancer cell line with metastatic capacity
Norm	Normal
PCL	Papillary carcinoma cell line
PCL	Papillary thyroid carcinoma
PVPTC	Papillary variant papillary carcinoma
UCL	Undifferentiated carcinoma cell line

3. Thyroid cancer expression data

Table 2. Thyroid cancer profiling studies included in analysis

Study	Platform	Genes/features	Comparison		Up/down
			Condition 1 (No. samples)	Condition 2 (No. samples)	
Chen <i>et al.</i> 2001	Atlas cDNA (Clontech)	588	M (1)	PTC (1)	18/40
Araldi <i>et al.</i> 2005	Custom cDNA	1807	BU(L) (1) PCL(L) (1) CTN(L) (1)	Norm (1) Norm (1) Norm (1)	9/20 1/8 1/7
Huang <i>et al.</i> 2001	Affymetrix HG-U95A	12558	PTC (8)	Norm (8)	24/27
Alfred <i>et al.</i> 2004	Affymetrix HG-U95A	12558	PTC (9)	PTC (9), Norm (13)	14/20
Ceratti <i>et al.</i> 2004	SAGE	N/A	FA(L)	PTC(L), Norm (1)	0/65
Bozinger <i>et al.</i> 2001	Atlas cDNA (Clontech)	588	APFN(C), CTN(C)	Norm (6)	0/16
Bozinger <i>et al.</i> 2001	Affymetrix HG-U95A	12558	PTC(7), PVPTC(7)	FA(L), HN(7)	48/85
Zou <i>et al.</i> 2004	Atlas cancer array	1176	MACL(4)	AC(L)	4/21
Weber <i>et al.</i> 2005	Affymetrix HG-U133A	22283	FA(L)	PTC(L)	12/84
Hawthorne <i>et al.</i> 2004	Affymetrix HG-U95A	12558	PTC(8)	CTN(8)	10/28
Onda <i>et al.</i> 2004	Ambrosam custom cDNA	27648	AC(L), ATC(L)	Norm (10)	31/56
Wasenius <i>et al.</i> 2003	Atlas cancer cDNA	1176	PTC(18)	Norm (3)	12/9
Burden <i>et al.</i> 2003	Affymetrix HG-U95A	12558	PTC(9)	FA(10)	59/45
Yano <i>et al.</i> 2004	Ambrosam custom cDNA	3068	PTC(7)	Norm (7)	5/40
Chevillard <i>et al.</i> 2004	Custom cDNA	5760	PTC(3)	FA(4)	12/31
Mazzanti <i>et al.</i> 2004	He-Uniformed cDNA	10000	PTC(10), PVPTC(15)	FA(L), HN(15)	5/41
Takano <i>et al.</i> 2000	SAGE	N/A	PTC(1)	ATC(1)	3/10
Finley <i>et al.</i> 2004	Affymetrix HG-U95A	12558	PTC(1)	FA(1)	4/1
Pawes <i>et al.</i> 2004	SAGE	N/A	PTC(1)	FA(1)	2/11
Jarabak <i>et al.</i> 2005	Affymetrix HG-U133A	22283	PTC(1)	FA(1)	7/6
Giordano <i>et al.</i> 2005	Affymetrix HG-U133A	22283	PTC(51)	Norm (4)	90/51
21 studies	10 platforms		34 comparisons (473 samples)		1785

4. Overlap analysis results

Table 3. Comparison groups analyzed for overlap

Overlap analysis group	condition set 1	Condition set 2	comps	# genes (multi-study)
All	Any	Any	34	755 (107)
Cancer vs. non-cancer	AC(L), ATC, PCL, PTC, PVPTC, HCC, M, MACL, PCL, PTC, TC(TC), UCL	APFN, CTN, FA, GT, HN, Norm	21	755 (107)
Cancer vs. normal	AC(L), ATC, PCL, PTC, PVPTC, HCC, M, MACL, PCL, PTC, TC(TC), UCL	Norm	12	478 (53)
Cancer vs. benign	AC(L), ATC, PCL, PTC, PVPTC, HCC, M, MACL, PCL, PTC, TC(TC), UCL	APFN, CTN, FA, GT, HN	8	332 (38)
Normal vs. benign	Norm	APFN, CTN, FA, GT, HN	3	19 (1)
Papillary cancer vs. non-cancer	PVPTC, PCL, PTC, TC(TC)	APFN, CTN, FA, GT, HN, Norm	12	563 (82)
Papillary cancer vs. normal	PVPTC, PCL, PTC, TC(TC)	Norm	6	309 (49)
Papillary cancer vs. benign	PVPTC, PCL, PTC, TC(TC)	APFN, CTN, FA, GT, HN	8	181 (13)
Papillary cancer vs. other	PVPTC, PCL, PTC, TC(TC)	Any other	15	528 (107)
PVPTC vs. other	PVPTC	Any other	10	319 (39)
PTC vs. FA	PTC	FA	6	222 (3)
Papillary cancer vs. other	PTC, PCL	Any other	10	65 (16)
Anaplastic cancer vs. other	AC(L), ATC, M, MACL	Any other	4	145 (14)
Anaplastic cancer vs. other	AC(L), ATC, MACL	Any other	3	91 (6)

Figure 2. Gene overlap for cancer vs. non-cancer analysis

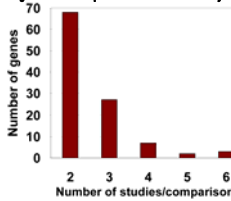


Table 3: Each overlap analysis group defines an artificial group of comparisons for which gene overlap was analyzed. In all groups considered except for one, we identified one or more genes that were reported in two or more studies. For example, the 'cancer vs. non-cancer' group (highlighted) includes all comparisons between what we would consider 'cancer' (as in condition set 1) and 'non-cancer' (as in condition set 2). In this case, 21 comparisons met the criteria and produced a list of 755 potential cancer markers, 107 of which were identified in multiple studies. These 'multi-study cancer versus non-cancer markers' are summarized further in figures 2-3 and tables 4-6.

Fig 2: A breakdown of the 107 genes found in multiple studies for the cancer versus non-cancer analysis. Some genes were observed in as many as six studies.

4. Overlap analysis results (cont'd)

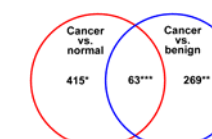


Fig 3: Of the 478 genes in the cancer/normal comparison and 332 genes of the cancer/benign group, a total of 63 genes were found in both. Table 4: shows a partial list (genes identified in 4 or more comparisons) from the cancer vs. non-cancer analysis. A complete table of all genes and all others are available as supplementary data (www.bcgsc.ca/bioinfo/getthyroid).

Table 5: Of the 107 genes with multi-study confirmation from the cancer versus non-cancer overlap analysis group, 102 were present in the Gene Ontology set of 15240 human genes. From this list, a total of 12 GO terms were found to be statistically over-represented: 3 biological process (B); 3 cellular component (C); and 6 molecular function (F). The 'obsoletal' column shows the number of genes from the list found associated with each GO term over the total number of genes annotated to that term in GO. The p-value was calculated using the BINGO[4] plugin for Cytoscape[5]. P-values are for a hypergeometric test, corrected with a Benjamini & Hochberg False Discovery Rate (FDR) correction, and a cut off of 0.05 applied to the result.

Table 4. Cancer versus non-cancer genes identified in 4 or more independent studies

Gene	Description	Comp's (Up/Down)	N	Fold Change
MITF	mit proto-oncogene (transcription factor)	60	522	3.03
TPF3	transcription factor 3 (testis)	60	196	-14.70
SERPINH1	serpin (or cysteine) proteinase inhibitor, clade A (alpha-1 antitrypsin, alpha-1)	60	192	15.84
EPAS1	epidermal growth factor receptor pathway substrate 8	50	186	15.15
TMEM1	tissue inhibitor of metalloproteinase 1 (kathryn potentiating activity, collagenase inhibitor)	50	142	5.38
TCF4	transforming growth factor, alpha	40	165	4.64
PCYT	glutamyl-proline cyclitolactonase (glutaminyl cyclase)	40	153	7.31
PROS1	protein S (alpha)	40	149	4.32
CRABP1	cellular retinoic acid binding protein 1	40	149	11.35
FN1	fibronectin 1	40	128	7.68
FCGBP	Fc fragment of IgG binding protein	04	108	-2.41
TPPO	thymol peptidase	04	91	-1.69

Table 5. GO analysis of multi-study genes from the cancer versus non-cancer overlap analysis group

GO term	Ontology	p-value	Genes in test set	Genes in background
extracellular region	GO:0005576	1.0E-04	107	15240
cytoplasm	GO:0005737	1.0E-04	107	15240
thyroid hormone generation	GO:0005737	1.0E-04	107	15240
thyroid hormone metabolism	GO:0005737	1.0E-04	107	15240
selenium binding	GO:0005737	1.0E-04	107	15240
thyroid hormone metabolism	GO:0005737	1.0E-04	107	15240
MAP kinase phosphatase activity	GO:0005737	1.0E-04	107	15240
thyroxine 5'-deiodinase activity	GO:0005737	1.0E-04	107	15240
cooper ion binding	GO:0005737	1.0E-04	107	15240
extracellular matrix	GO:0005737	1.0E-04	107	15240
retinoic acid receptor activity	GO:0005737	1.0E-04	107	15240

5. Conclusions and Future work

Conclusions:

> A significant number of genes are consistently identified in the literature as differentially expressed between different thyroid tissue and tumour subtypes > These consistent genes represent a useful starting point for a large-scale tissue microarray analysis to identify useful prognostic and diagnostic markers in a clinical setting

Future work:

> Meta-analysis starting from raw expression data > Selection of final candidates for tissue microarray (TMA) analysis > Development of classifier for thyroid tissue based on results of TMA

6. Acknowledgments

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references | 1. Vahedi et al. unpublished; <http://www.bcgsc.ca/bioinfo/getthyroid>; 2. Dennis et al. 2003; <http://www.david.abcc.ncifcrf.gov/>; 3. Affymetrix; <http://www.affymetrix.com/support/index affy>; 4. Maere et al. 2005; <http://www.pub.utd.edu/biochem/papers/BINGO/>; 5. Shannon et al. 2003; <http://www.cytoscape.org/>