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Review article

DNA microarray analysis as a tool to investigate the therapeutic mechanisms and drug development of Chinese medicinal herbs

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ABSTRACT

Chinese herbal medicines have been used for the treatment of various diseases for centuries. Although several herbal formulas and herbal components have shown therapeutic potential, the active components and the molecular mechanisms mediating the effects of said formulas remain to be discovered. Microarray analysis has become a widely used tool for the generation of gene expression data on a genome-wide scale. This paper discusses the application of whole genome expression profiling as a tool to investigate the molecular mechanisms governing the therapeutic effects of traditional Chinese medicine. This review also highlights how data derived from DNA microarray analysis can be used to screen for drug targets of various herbal drugs, to predict the therapeutic potential of herbal drugs, to analyze the safety of drugs in the preclinical stage of drug development, and to establish a modern definition of traditional Chinese medicine.

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1. Introduction

Systems biology serves as a translational platform between traditional Chinese medicine and modern science. In this study, we review the technology behind whole genome expression profiling and discuss the biomedical application of the technique to the study of Chinese medicinal herbs.

2. Technology behind genome expression profiling

2.1. Development of whole genome expression profiling

In 1995, Schena and colleagues[1] at Stanford University in Palo Alto, CA, USA, published the first paper on the use of

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131 Q1 cDNA microarray probes printed in a two-dimensional grid
 132 onto glass slides. They showed that their high-capacity
 133 system could simultaneously monitor the expression of
 134 many genes. Microarrays prepared by high-speed robotic
 135 printing of complementary DNAs on glass are used for
 136 measurements of quantitative expression of corresponding
 137 genes. Because of the small format and the high density of
 138 arrays, hybridization volumes of less than two microliters can
 139 be used, enabling the detection of rare transcripts in probe
 140 mixtures derived from two micrograms of total cellular
 141 mRNA. Two-color fluorescence hybridization is then used to
 142 Q2 simultaneously visualize differentially expressed genes. In
 143 1996, Affymetrix began to market commercially available DNA
 144 chips. Various microarray experimental platforms have been
 145 developed since then.

2.2. Commonly used microarray platforms

150 Three different types of microarray platforms are commonly
 151 used: spotted cDNAs, spotted oligonucleotides, and Affyme-
 152 trix arrays [2].

153 Q3 Spotted cDNA arrays typically use sets of specific cDNA
 154 plasmids in gridded liquid. The inserts of each clone are
 155 typically amplified by polymerase chain reaction, and a few
 156 pico liters are physically spotted onto glass slides by liquid-
 157 handling robots. Spotted cDNA arrays are only used in
 158 academic centers because of their flexibility and relatively low
 159 cost.

160 Spotted oligonucleotide arrays are also built on glass slides
 161 by liquid-handling robots; however, the input solution
 162 comprises synthetic oligonucleotide (often 60–70 mer) rather
 163 than plasmids. Most of the process is automated, leading to
 164 less sample mix-up and less sample dropout. Disadvantages
 165 of spotted oligonucleotides include the relatively high cost of
 166 synthesizing large numbers of large oligonucleotides and the
 167 nonrenewable nature of the resource. Nonetheless, spotted
 168 oligonucleotide arrays are still widely used.

169 Q4 Affymetrix GeneChips are factory designed and synthe-
 170 sized. Design is done using software to choose a series of 11-
 171 to 25-mer probes from the 3-foot end of each transcript or pre-
 172 dicted transcript in the genome. Synthesis of arrays is done
 173 using light-activated chemistry and photolithography
 174 methods. Spotted oligonucleotides and Affymetrix arrays have
 175 superseded the use of spotted cDNAs. The manufacturers of

commonly used DNA microarray platforms are listed in
 Table 1.

2.3. Limitations and standardization of microarray platforms

DNA microarrays enable researchers to simultaneously
 monitor the expression of thousands of genes. However, the
 current technology has several limitations. The major prob-
 lems are sensitivity, accuracy, specificity, and reproducibility
 of microarray results. Studies have shown that, for relatively
 abundant transcripts, the existence and direction, but not the
 magnitude, of expression changes can be reliably detected. Q5
 However, accurate measurements of absolute expression
 levels and the reliable detection of low abundance genes are
 difficult to achieve. The main problems seem to be the
 suboptimal design or choice of probes and some incorrect
 probe annotations. Marshall [3] compared the reliability of
 numerous array platforms, including the Affymetrix Gene-
 Chip, the Agilent array and the Amersham array systems, Q6
 and found that more than one-half of the variability observed
 in the results was attributable to differences in the microarray
 platforms themselves. Efforts to standardize microarray data
 have been underway for some time and include the stan-
 dardization of sample preparation, RNA isolation, cDNA
 synthesis, hybridization analysis, and quality control check-
 points to ensure reproducibility of data. For example, quality
 control criteria for RNA isolation include yield, purity, and
 integrity. An RNA integrity number greater than eight indi-
 cates that the RNA sample is suitable for cDNA synthesis. The
 criteria for cDNA labeling include concentration and incor-
 poration efficiency. An incorporation efficiency of 15 labeled
 nucleotides per 1000 cDNA nucleotides indicates that cDNA
 labeling is suitable for hybridization. The gene expression
 profile obtained using standardized protocols can yield data
 that are consistent between laboratories and are intrinsically
 comparable [4].

Use of identical microarray chips and identical protocols
 would minimize the efforts made by researchers to integrate
 expression data, thereby allowing for the information
 embedded in these data to be maximally explored. In 2004, the
 Microarray and Gene Expression Data (MGED) society wrote an
 open letter to scientific journals proposing standards for
 publication. The MGED society suggested that journals require

Table 1 – Manufacturers of DNA microarray platforms.

Manufacturer	Location	Website
Affymetrix	Santa Clara, CA, USA	www.affymetrix.com
Agilent Technologies Expression Analysis	Santa Clara, CA, USA Durham, NC, USA	www.agilent.com www.expressionanalysis.com
Jivan Biologics	Larkspur, CA, USA	www.jivanbio.com
Marligen Biosciences	Ijamsville, MD, USA	www.marligen.com
NanoString Technologies	Seattle, WA, USA	www.nanosttring.com
NimbleGen	Madison, WI, USA	www.nimblegen.com
Oxford Gene Technology	Oxford, UK	www.ogt.uk
PerkinElmer	Waltham, MA, USA	www.perkinelmer.com
Phalanx Biotech Group	Hsinchu, Taiwan	www.phalanxbiotech.com

261 submission of microarray data to one of two public repositories: Gene Expression Omnibus (GEO) or ArrayExpress. More-
262 over, they stated that authors should provide a checklist of
263 variables and supply the checklist as supplementary infor-
264 mation at the time of submission. Other members of the
265 microarray community welcomed these steps, in particular
266 Brazma and colleagues [5], who proposed the Minimum
267 Information About a Microarray Experiment (MIAME),
268 a guideline that describes the minimum information required
269 to ensure that microarray data can be easily interpreted. The
270 standardization of global gene expression data will make
271 microarray data much more useful and accessible.

272 In summary, DNA microarray technology has evolved
273 rapidly since its introduction in 1995. Although certain limi-
274 tations of the current technology exist and have become more
275 apparent during the past couple of years, the ability of
276 microarrays to monitor the expression of thousands of genes
277 simultaneously is unsurpassed [6].

282 3. Application of whole genome expression 283 profiling to traditional Chinese medicine studies

284 Whole genome expression profiling can be applied to study
285 the biomedical effects of Chinese medicinal herbs. Extracts
286 prepared from medicinal plants and other natural sources
287 contain a variety of molecules with potent biological activities.
288 Unfortunately, it is often difficult to analyze the biologic
289 activities of these extracts because of their complex nature
290 and the possible interaction of their components. Genome-
291 wide expression monitoring with high-density microarrays
292 provides a simple way to test the biochemical effects of herbs,
293 thereby gaining insight into their potential beneficial effects
294 and negative side effects. DNA microarray has been used to
295 evaluate the toxicity of novel drug candidates and to identify
296 disease targets for drug development. Additionally, the ther-
297 apeutic efficacy of a given drug can be predicted on the basis of
298 gene expression patterns *in vitro*.

303 3.1. Evaluation of biologic activity and mechanisms of 304 Chinese herbs

305 Microarray data have been used to characterize the biologic
306 activities and mechanisms of action of herbal formulae or
307 herbal compounds. For example, PC-SPES is a dietary
308 supplement comprised of extracts from eight different herbs:
309 *Scutellaria baicalensis*, *Glycyrrhiza glabra*, *Ganoderma lucidum*,
310 *Isatis indigotica*, *Panax pseudo-ginseng*, *Dendranthema mor-
311 ifolium*, *Rabdosia rebescens*, and *Serenoa repens*. PC-SPES is also
312 used as an alternative therapy by patients with prostate
313 carcinoma [7–9]. The gene expression profile in cultured cells
314 that have been exposed to PC-SPES shows differential
315 expression of genes involved in modulating cell cycle, cell
316 structure, and androgen response, indicating that alteration
317 of some of those genes may be responsible for PC-SPES-
318 mediated cytotoxicity [10]. Yukmijihwang-tang (YMJ), also
319 known as LiuWei Dihuang Wang, is composed of six different
320 medicinal herbs, including *Rehmannis radix*, *Radix dioscoreae*,
321 *Fructus corni*, *Poria*, *Cortex moutan*, and *Radix alismatis*. YMJ has
322 been widely used for centuries as an antiaging herbal formula

326 in Asian countries [11]. Microarray data indicate that YMJ
327 enhances memory retention by inducing several genes that
328 are involved in protecting neuronal cells, enhancing cell
329 proliferation, and stimulating neurite growth [12]. *Pinelliae*
330 *Rhizoma* extract (PRe) is used to treat cough and asthma.
331 However, the mechanism by which PRe exerts its effect on
332 psychological disorders has not been studied. Kim and
333 coworkers [13] used microarray to analyze the effect of PRe in
334 mice exposed to psychological stress. They found that the
335 expression of most genes that are altered in response to
336 psychological stress is restored to normal levels in PRe-treated
337 mice, with recovery rate of 81.5% for up-regulated genes and
338 85.2% for down-regulated genes. When the interaction
339 network was analyzed, the recovery rate of the core node
340 genes (46 up- and 29 down-regulated genes) in PRe-treated
341 mice was over 95%, indicating that those genes may be the
342 effective targets of PRe. Curcumin, a major chemical compo-
343 nent of *Curcuma longa*, is used as a spice to give a specific flavor
344 and yellow color to curry. It is also used as a cosmetic agent
345 and in some medical preparations [14]. Curcumin displays
346 anticarcinogenic properties in animals [15,16]. Microarray-
347 based gene expression patterns indicate that, in addition to
348 anticarcinogenic effects, curcumin may be an effective anti-
349 metastatic agent via the regulation of expression of certain
350 genes [17]. Aristolochic acid (AA), the major constituent of
351 *Aristolochia* species, is associated with nephritis and renal
352 cancer [18–20]. Microarray and network analysis have shown
353 that most AA-altered genes are connected with nuclear factor-
354 κ B (NF- κ B), suggesting that NF- κ B plays a critical role in the
355 pathogenesis of AA-induced renal diseases [21]. Extracts
356 prepared from medicinal plants and other natural sources
357 contain a variety of molecules with potent biological activities;
358 the aforementioned studies suggest that genome-wide
359 expression monitoring with high-density microarrays is an
360 effective method for analyzing the biologic activities of those
361 extracts.

362 3.2. Establishing a modern definition of traditional 363 Chinese medicine

364 Chinese herbal formulas consist of several herbal compo-
365 nents. However, the mechanisms of action of most Chinese
366 herbal formulas and the relationship between formulae and
367 their components remain to be elucidated. The putative
368 mechanism of San-Huang-Xie-Xin-Tang (SHXXT) and the
369 relationship between SHXXT and its herbal components were
370 analyzed in our laboratory using a microarray technique [22].
371 Gene-set enrichment analysis indicated that SHXXT and its
372 components displayed a unique anti-proliferation pattern
373 involving p53 and DNA damage signaling pathways in HepG2
374 cells. Network analysis showed that SHXXT-affected genes
375 were regulated by p53. In addition, clustering analysis
376 showed that *Rhizoma coptis*, the principal herb in SHXXT,
377 shared a similar gene expression profile with SHXXT. These
378 findings indicate that *R coptis* is the principal herb in the
379 herbal combination SHXXT (Fig. 1). To the best of our
380 knowledge, this was the first study to reveal the relationship
381 between a traditional Chinese medicine formula and its
382 herbal components using microarray and bioinformatics
383 approaches.

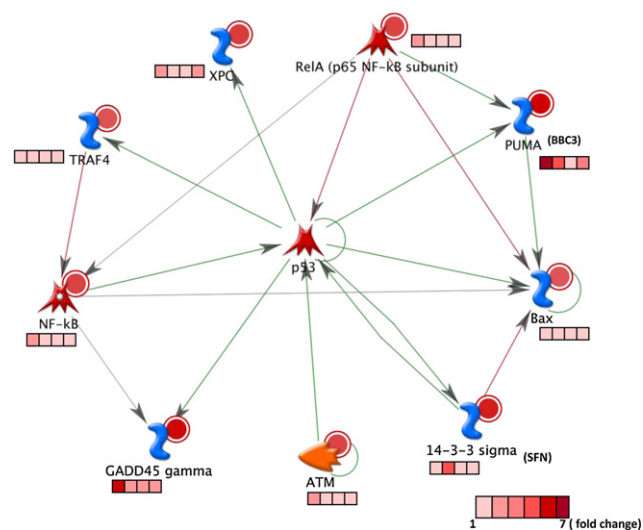


Fig. 1 – Network analysis of SHXXT-regulated genes. We selected the target genes that are regulated by p53 from <http://rulai.cshl.edu/cgi-bin/TRED/tred.cgi?process=searchTFGeneForm>. To estimate the overall regulatory effect of SHXXT on these target genes, we used the ‘geneSetTest’ function implemented in the R program of the Limma package to compare the absolute t-statistic values for these target genes with those for all genes. These target genes were then combined with the differentially expressed genes, which belonged to the Gene Ontology (GO) category ‘regulation of biological process,’ to investigate their relationship with p53. We used the MetaCore Analytical suite to construct the interaction networks between p53-downstream genes and part of the differentially expressed genes. The fold changes in gene expression in SHXXT-, *Rheum officinale*-, *Coptis chinensis*-, and *Radix scutellariae*-treated cells, respectively, are shown at the bottom. SHXXT = San-Huang-Xie-Xin-Tang.

3.3. Evaluation of drug safety

Many natural products, including polyphenols, terpenes, alkaloids, flavonoids, and phenolics, are potential therapeutic agents [23]. Previous studies have shown that phytochemicals affect the expression levels of genes involved in drug metabolism [24]. To evaluate whether phytochemicals affect drug metabolism, we analyzed the expression levels of genes encoding phase I and II drug metabolism enzymes in cells exposed to anthraquinone compounds. Phase I drug metabolism genes encode alcohol dehydrogenases, aldehyde dehydrogenases, and cytochrome P450 families, while phase II drug metabolism genes encode glutathione S-transferases, sulfotransferase, and UDP glucuronosyltransferase (UGT) families. We found that genes involved in phase II drug metabolism were down regulated during anthraquinone compound treatment (Table 2). These data suggest that anthraquinone compounds may slow down the excretion of drugs, thereby increasing the half-life of drugs [25].

Table 2 – Analysis of expression levels of genes associated with drug metabolism.^a

Gene symbol	log ₂ ratio	Standard deviation
UGT1A10	−0.23	1.89
UGT2A1	−0.28	1.15
UGT2B11	−1.70	5.42
UGT2B15	−0.85	0.69
UGT2B4	−0.34	0.29
UGT2B7	−0.94	0.79

^a Results were obtained from three independent assays. A total of 219 genes associated with drug metabolism were selected from ‘The Pharmacogenetics and Pharmacogenomics Knowledge Base’ website (<https://www.pharmgkb.org/index.jsp>). Among these genes, we analyzed the expression levels of phase I drug metabolism genes, including alcohol dehydrogenases, aldehyde dehydrogenases, and cytochrome P450 genes, and phase II drug metabolism genes, including glutathione S-transferases, sulfotransferase, and UGT genes. The log₂ ratio and standard deviation of UGT genes are shown.

3.4. Prediction of the therapeutic potential of medicinal herbs

Vanillin has been shown to inhibit mutagenesis and to suppress the invasion and migration of cancer cells [26]. In our previous studies, microarray data and gene ontology investigation indicated that vanillin affected clusters of genes involved in the cell cycle and apoptosis. Network analysis indicated that Fos might play a central role in the regulation of the gene expression network. Results from reporter assay and Western blot further indicated that vanillin inhibited Fos-related transcription factor activator protein 1 (AP-1) activity via an extracellular signal-regulated kinase pathway. Our data suggest that vanillin exhibits anticancer potential by regulating cell cycle and apoptosis and that its regulation may involve the suppression of AP-1 (Fig. 2) [27,28].

AA belongs to a family of compounds found in the Aristolochiaceae family of plants. *Aristolochia* species in particular have been used for centuries in Asia for medicinal purposes. Although AA is bioactivated in both the kidney and liver, it only induces diseases and tumors in kidney and urinary tract in human and rodents [18]. To elucidate why AA displays such tissue-specific carcinogenicity, Chen and colleagues [29] examined gene expression profiles in kidney and liver of rats treated with carcinogenic doses of AA. They found that the biologic processes related to defense response, apoptosis, and immune responses were significantly altered by AA exposure in kidney but not in liver. These findings may explain why AA induces tumors in the kidney but not in the liver [29].

Ginkgo biloba extract EGb 761 is widely used to treat neurologic disorders [30,31]. In a previous study, we tested the effects of EGb761 on the transcriptional profile of mouse genes. A KEGG pathway analysis showed that EGb761 affected the neuroactive ligand-receptor interaction pathway in brain. A total of 53 genes were significantly affected, and EGb761 up-regulated a subgroup of dopamine receptors, especially dopamine receptor 1a. Immunohistochemical staining confirmed the microarray data. The finding that *G biloba*

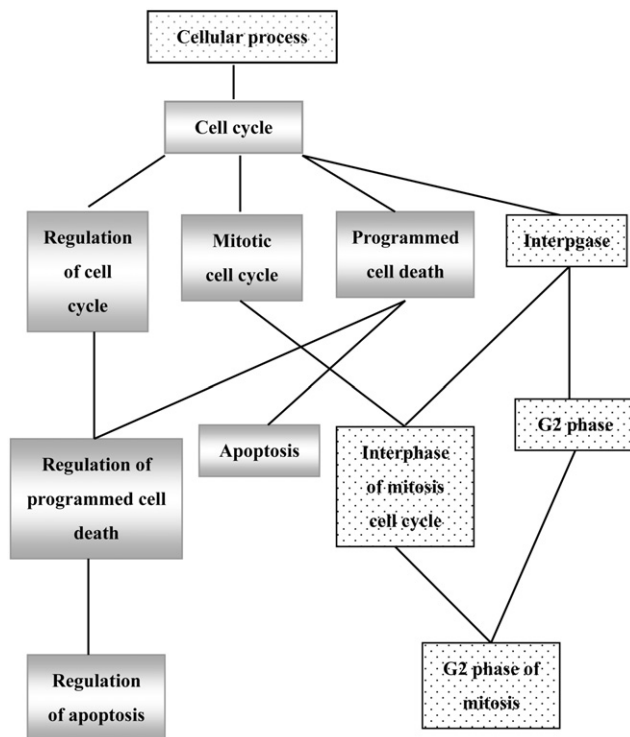


Fig. 2 – Ontology analysis of vanillin-affected genes. vanillin-affected genes were analyzed by GO on the Gene Ontology Tree Machine website (<http://bioinfo.vanderbilt.edu/gotm/>), a web-based and tree-based data-mining environment for gene sets. We used the WebGestalt tool to test significant GO terms, and the significant GO terms are shown.

treatment resulted in increased expression of dopamine receptor 1 in brain may explain why EGb761 is an effective treatment of neurologic disorders such as Parkinson disease (Table 3) [32].

3.5. New drug development

Whole genome expression profiling has also been used for the development of new drug [33–36]. Large-scale gene expression analyses of toxin-treated cells and animals have yielded information on the toxic potential of novel drug candidates [37–41]. In addition, gene expression profiles have been applied to identify the disease targets for drug development [42]. Moreover, the therapeutic efficacy of drugs can be predicted on the basis of gene expression signatures *in vitro* [43,44].

A number of studies have shown that DNA microarray data have potential utility in drug discovery and drug target validation [44,45]. For example, Lamb and others [46] analyzed the expression profiles of 164 small molecules with DNA microarray. By comparing the genomic signatures of drug candidates or the disease state to this resource, the authors found that it was possible to identify potential mechanisms of action, confirm previous applications of known drugs, and identify additional potential uses for known drugs [46]. Their

results demonstrate that the establishment of a huge gene expression database would be useful for finding connections among small molecules that share similar mechanisms of action and that are involved in similar physiologic processes, thereby allowing for the development of disease-fighting drugs.

Several studies have indicated similarities between gene expression profiles and therapeutic activities [46–48]. In addition, genome-wide expression monitoring with high-density microarrays provides a simple way to test biochemical effects of herbs, thereby gaining insights into their potential beneficial effects and negative adverse events [30]. In a recent study, we applied DNA microarray to analyze biologic events, predict the therapeutic potential of drugs, and evaluate the safety of herbal formulas [49]. For seven consecutive days, mice were administered orally with 15 of the most widely used Chinese herbal formulae listed in the Taiwan National Health Insurance Database, and the gene expression profiles in liver or kidney were analyzed by DNA microarray. Our data showed that most formulas altered metabolic pathways, such as the pathways governing glutathione metabolism and oxidative phosphorylation, and regulatory pathways, such as that regulate antigen processing and presentation and insulin-like growth factor signaling. By comparing the gene expression signatures of formulas with those of disease states or drugs, we found that response of mice to formula might be associated with disease state in said mice, such as metabolic or cardiovascular diseases. Moreover, most formulas altered the expression levels of cytochrome P450, glutathione S-transferase, and UGT genes, suggesting that caution should be paid to possible drug interactions of these formulas. Furthermore, the similarities in gene expression profiles between formulas and toxic chemicals were low in kidney, suggesting

Table 3 – Neuroactive ligand-receptor interaction of EGb761-affected genes in the brain and kidney.^a

	Observed (total)	<i>p</i> value
Brain	53 (237)	4.13×10^6
Kidney	0 (237)	0.536

a Fluorescent RNA targets were prepared from 5 μ g of total RNA using a MessageAMP aRNA kit and Cy5 dye. Fluorescent targets were hybridized to Mouse OneArray Whole Genome DNA microarray. After an overnight hybridization at 50 °C, nonspecific binding targets were washed away and the slides were scanned with an Axon 4000 scanner. The Cy5 fluorescent intensity of each spot was analyzed by genepix 4.1 software. The signal intensity of each spot was corrected by subtracting background signals. Spots with a signal-to-noise ratio of less than 0 as well as those of control probes were filtered. Spots that passed these criteria were normalized by the R program in the Limma package using quantile normalization [50]. The *p* value of each gene was calculated by *t*-statistics using the Differential Expression (T-Rex) tool in the Gene Expression Pattern Analysis Suite [51]. These differentially expressed genes (*p* < 0.01) were further analyzed by the KEGG pathway [52]. Pathway enrichment analysis was performed on the WebGestalt website (<http://bioinfo.vanderbilt.edu/webgestalt/login.php>) by the hypergeometric test, which is used to evaluate the *p* value of the over-represented pathways. The neuroactive ligand-receptor interaction pathway is shown.

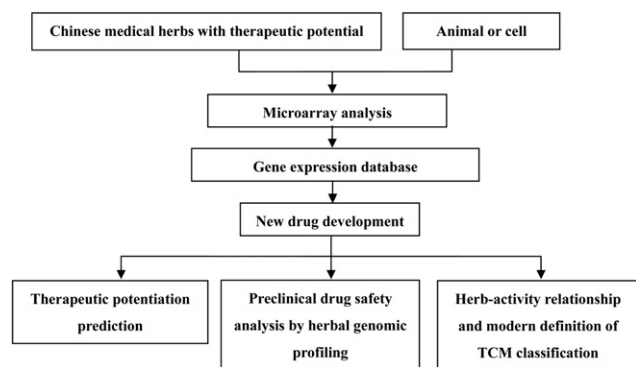


Fig. 3 – Paradigm for the application of whole genome expression profiling as a tool for therapeutic prediction, drug development, and safety evaluation of Chinese herbal medicines.

that these formulas might not induce nephrotoxicity in mice. This transcriptomic platform will not only help researchers understand the therapeutic mechanisms associated with herbal formulas and gene interactions, but will also help researchers develop novel disease-fighting drugs (Fig. 3).

4. Conclusion

Whole genome expression profiling can provide a basis for investigating the molecular mechanisms governing the therapeutic effects of Chinese herbal medicines and can be used to elucidate the biology of disease progression, identify potential therapeutic targets, and facilitate the development of traditional Chinese medicine-derived biopharmaceutical products.

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