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Integration of metabolomics and transcriptomics in nanotoxicity studies

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ABSTRACT

Biomedical research involving nanoparticles has produced useful products with medical applications. However, the potential toxicity of nanoparticles in biofluids, cells, tissues, and organisms is a major challenge. The ‘-omics’ analyses provide molecular profiles of multifactorial biological systems instead of focusing on a single molecule. The ‘omics’ approaches are necessary to evaluate nanotoxicity because classical methods for the detection of nanotoxicity have limited ability in detecting miniscule variations within a cell and do not accurately reflect the actual levels of nanotoxicity. In addition, the ‘omics’ approaches allow analyses of in-depth changes and compensate for the differences associated with high-throughput technologies between actual nanotoxicity and results from traditional cytotoxic evaluations. However, compared with a single omics approach, integrated omics provides precise and sensitive information by integrating complex biological conditions. Thus, these technologies contribute to extended safety evaluations of nanotoxicity and allow the accurate diagnoses of diseases far earlier than was once possible in the nanotechnology era. Here, we review a novel approach for evaluating nanotoxicity by integrating metabolomics with metabolomic profiling and transcriptomics, which is termed “metabotranscriptomics.”

INTRODUCTION

Several recent reports suggest breakthrough applications of nanoparticles (NPs) in biomedical and clinical fields (1-4). NPs have unique physiochemical properties because of their size and large surface area-to-volume ratio, rendering them more reactive and thermodynamically unstable than bulk materials (5, 6). Moreover, NPs are easily absorbed and readily interact with the human body when delivered through inhalation, penetration, and ingestion (7, 8). However, issues regarding their toxicity and safety due to adverse biological effects have led to widespread concerns about the possible negative effects of NPs (9, 10). Specifically, NP-induced oxidative stress is difficult to evaluate with classical methods (11, 12). The practical application of NPs requires studies of biological toxicity. However, there are limitations associated with the evaluation of nanotoxicity using traditional methods of analysis because of the complexities at the nanolevel, underscoring the need for advanced omics approaches.

Omics facilitates the collective characterization and quantification of many different molecules, such as DNA, RNA, proteins, lipids, and metabolites, in cells, tissues, organs, and organisms. The last few decades have produced developments in high-throughput technologies for omics, which enabled comprehensive understanding of processes and novel findings in biomedical studies (13-15). Thus, omics approaches address the complexity of biological systems via interpretations using bioinformatics analyses. Rapid developments in nanotechnology and the production of NPs, which are defined as engineered materials measuring less than 100 nm in one dimension, stress the importance of the potential toxicity of NPs (6, 7, 10, 16). The omics approaches including genomics, transcriptomics, proteomics, and metabolomics are used to evaluate nanotoxicity (17-20). However, a single omics approach provides limited insight into the intricate molecular pathways and the complex biological events in cells and organisms (21-23).

The concept of integrated omics was introduced by Dr. Hood, who suggested a systems

biology approach based on the combination of different omics data to provide a comprehensive understanding of the multifactorial origins of biological research (24, 25). Palsson *et al* suggested many approaches to generate multi-omic data sets and reduce the possibility of resource allocation for data generation versus data curation and integration (26). Integrated omics has been applied to a wide range of complex and intractable problems in biological studies. In particular, integrated omics was shown to facilitate the determination of cytotoxicity, especially nanotoxicity (27, 28) because cytotoxicity induced by NP treatment in cells could not be detected with traditional methods (29-31). Integrated omics provides a more comprehensive overview of the complexities associated with nanotoxicity compared with a single omics approach. Thus, these technologies contribute to expanding the safety evaluation of nanotoxicity and provide accurate diagnoses of diseases compared with the simple and fragmented interpretations using single omics approaches. Here, we divide this review into three sections as follows: (i) omics approaches for nanotoxicity; (ii) recent approaches for metabolomics and transcriptomics in nanotoxicity; and (iii) integration of omics for the analysis of nanotoxicity.

OMICS APPROACHES FOR NANOTOXICITY

Omics approaches provide a better understanding of cellular events by using large-scale data. In particular, high-throughput technologies in omics have enabled the use of large-scale data to generate novel findings related to NP toxicity and mechanisms of action (17-20). Thus, omics tools prevent the use of fragmented data that could lead to inappropriate conclusions about nanotoxicity. Although the molecular technologies for understanding nanotoxicity, stress responses, molecular damage, and varying responses to NPs have advanced in parallel with molecular cell biology and *in vivo* assessments, the traditional approaches for the safety evaluation of new NPs have limitations regarding their potential toxicity. Thus, omics

techniques are well suited to evaluate nanotoxicity both *in vitro* and *in vivo* by providing a more comprehensive view than was previously possible.

A systematic understanding of molecular responses in biological systems has been emphasized following the growth in analytic technologies and bioinformatics. Developments in sequencing technologies have allowed researchers to gather genomic and transcriptomic data (genotypic features) with much higher coverage and cost-effectively. In proteomics and metabolomics, advances in NMR and mass spectroscopy enable the analysis of a broader range of the proteome or metabolome (phenotypic features) with high precision and sensitivity (32). However, despite these improvements, single omics approaches have a fundamental “blind spot” in unraveling complex biological responses. For example, even though transcriptomics allows detection of extensive genotypic changes, it may not facilitate the interpretation of nucleic acid modifications in the genome or address issues concerning coverage of repeat-rich regions and low abundance genes, and is thoroughly inadequate for the determination of the actual phenotype (33, 34). In metabolomics, for which one endpoint is the biological phenotype (35), amplification methods are unavailable for minor metabolites, and a quantitative analysis of a targeted process can only provide a partial representation of an entire metabolic pathway (22). In addition, it is frequently associated with errors and limitations involving the interpretation of causal mechanisms in biological processes (36). The integration of two or more omics methods is highly recommended for a more comprehensive and holistic understanding of biological systems than is possible with a single omics approach.

RECENT APPROACHES FOR METABOLOMICS AND TRANSCRIPTOMICS IN NANOTOXICITY

In this section, we introduce the main omics approaches: metabolomics and transcriptomics

and their application to nanotoxicity studies.

Metabolomics

Metabolomics is the comprehensive analysis of chemical processes involving metabolites that drive cellular functions, such as cellular signaling cascades, homeostatic control, energy metabolism, and cell damage (37). Specifically, the metabolome represents the complete set of small-molecule chemicals found in biological fluids, cells, tissues, organisms, and biological samples; the metabolome directly links genotype with phenotype and is most related to the phenotype (35, 38). In contrast to other omics methods, metabolomics has great potential for the analysis and understanding of cellular biological mechanisms affected by NPs because metabolic changes accurately reflect the characteristic changes in biological fluids, cells, and tissues based on the quantitation of metabolome (27, 39-41).

Metabolomic profiling is necessary to evaluate potential toxicity using either nuclear magnetic resonance (NMR) or mass-spectrometry (MS). NMR is an effective tool for the determination of the structure of organic compounds *ab initio* and the quantitative analysis of a broad range of molecules (such as metabolic fingerprinting) in a crude extract without authentic standards (32, 42). In addition, NMR does not depend on hydrophobicity or metabolite dissociation value, and the results are comparatively more reproducible than those derived from MS (43). However, NMR has a relatively low sensitivity (> 1 nmol) and resolution and cannot detect NMR-inactive molecules (32). Thus, there are limitations for the comprehensive analysis of individual constituents within a sample (44). MS ionizes chemical species and sorts the ions based on their mass-to-charge ratio. It is one of the most widely used methods for the ultrasensitive and simultaneous detection of metabolites by coupling with gas or liquid chromatography (45, 46). Although different types of MS have a high sensitivity of detection, the sample preparation process is tedious, and the selectivity for different classes of

metabolites has both advantages and complications (43). In particular, metabolomic profiling of the cellular components, and target tissue metabolic reactions with gas chromatography-mass chromatography (GC/MS), without targeting a single metabolite, provides a better understanding of the biofluids, cells, and clinical conditions (28, 47-51). Thus, metabolomics has been used in nanotoxicity investigations utilizing high-throughput quantitation methods (11, 52). However, a limitation of metabolomics is that it provides consequential data without identifying the pathways of cellular mechanism. However, by integrating it with transcriptomics, a better understanding of subtle effects such as nanotoxicity can be obtained (27, 28).

Transcriptomics

Transcriptomics refers to the set of all RNA molecules in a cell and involves techniques such as microarray analysis and next-generation sequencing (NGS), called RNA-Seq. This technique is widely used to screen the toxicity of related RNA molecules and to elucidate the toxicity mechanisms (53, 54). Biological analytes from environmental, industrial, and drug-induced toxic exposures have been analyzed using transcriptomics (54-61). Moreover, transcriptomics contributes to the comprehensive investigation of cellular responses induced by NPs using bioinformatics software (22). Even though transcriptomics can provide large data from NP-treated cells, they are qualitative and do not establish a direct relationship with the pathology. Nanotoxicologists are working to overcome these shortcomings of transcriptomics by dovetailing the molecular mechanisms using other omics methods such as proteomics and metabolomics (27, 62). Moreover, metabolomics facilitates the identification of a phenotype based on cellular response and provides quantitative data, to compensate for the limitations of transcriptomics (27).

INTEGRATION OF OMICS FOR ANALYSIS OF NANOTOXICITY

Integration of omics

Integration of genomics, transcriptomics, proteomics, and metabolomics, facilitates a better understanding of the cellular biology because biological systems are dynamic and heterogeneous (21). Dr. Hood was a pioneer in terms of integrating the different data types and comparing them against a model with a focus on interdisciplinary and systems biology (25, 63). Compared with single omics approaches, integrated approaches provide a large volume of accurate information related to pathophysiology (64-68). Here, we describe the integration of transcriptomics with metabolomics in a process called “metabotranscriptomics” for the analysis of nanotoxicity, to allow a comprehensive analysis of the treatment outcomes with NPs.

Metabotranscriptomics for nanotoxicity analysis

The use of NPs in biomedical research, such as in the diagnosis and treatment of diseases, has gained tremendous momentum (1, 7, 69). Recent research has shown that internalized NPs can cause cytotoxicity by inducing reactive oxygen species (ROS) and increasing endoplasmic reticulum (ER) stress (7, 27, 70, 71). Despite increasing research efforts, the underlying mechanisms of NP toxicity are not clearly understood because of limited studies and preliminary stages of research.

Magnetic NPs (MNPs) and MNPs coated with biocompatible compounds, which are defined as single-dimensional particles with magnetic properties, have been investigated in the context of novel applications in biochemistry, biology, medicine, antibody engineering, cell tracking, and imaging tools (72-75). MNPs@SiO₂(RITC) are synthesized MNPs that consist of a cobalt ferrite core, CoFe₂O₄, and a silica shell containing chemically-bound Rhodamine B

isothiocyanate (RITC) for cell staining, separation and MRI contrast (76). The cobalt ferrite core and RITC contained within the silica shell contribute to the stability of $\text{MNPs}@SiO_2(\text{RITC})$ and prolonged red fluorescence at 540 nm without photobleaching. A study into the tissue distribution of $\text{MNPs}@SiO_2(\text{RITC})$ in mice demonstrated their ability to cross the blood–brain barrier (BBB) without inducing functional deficits. Analyses using hematoxylin and eosin (H&E) staining found no abnormal histopathological lesions in organs after the intraperitoneal injection (IP) of $\text{MNPs}@SiO_2(\text{RITC})$ into mice (29) (Fig. 1). Moreover, the injection did not induce any clinical changes (growth, body weight, behavior) or alterations in serum biochemical parameters (glucose, cholesterol, creatinine, and the ratio between the concentrations of the enzymes aspartate transaminase and alanine transaminase). In addition, a few *in vitro* studies such as FACS analysis, MTT assay, chromosome aberration assay, and cell cycle assay, failed to detect toxicity induced by NPs (27, 29-31) (Fig. 1).

Previous studies were confined to the pathophysiological effects of $\text{MNPs}@SiO_2(\text{RITC})$. Treatment with $\text{MNPs}@SiO_2(\text{RITC})$ yielded 24 metabolites, nine of which were considered to be significantly altered. However, assessment using only metabolomics does not yield convincing data and has low reliability. Integrating the metabolomic profiling with transcriptomics will allow a more sensitive and detailed toxicological evaluation of cellular responses to NPs and identify novel nanotoxicological biomarkers (27). Metabolic profiling of $\text{MNPs}@SiO_2(\text{RITC})$ -treated human embryonic kidney 293 (HEK293) cells revealed that changes in amino acids (AAs), organic acids (OAs) and a few metabolites were related to ROS generation (27), which triggered mitochondrial damage (Fig. 1).

The percent compositions and normalized values of AAs and OAs clearly revealed a marked increase in glutamic acid and pyruvate levels, and decrease in other AAs, such as alanine, valine, leucine, isoleucine, proline, and tyrosine, and other OAs, such as α -

ketoglutarate, oxaloacetate, fumarate, and malate, in the group treated with MNPs@SiO₂(RITC) (27). However, there is a limitation associated with linking ROS generation with metabolic changes (Fig. 2A). For the transcriptome, the expression levels of 45 ROS-generation-related genes were altered. Specifically, 26 genes were upregulated and 19 genes were downregulated, and these genes were found to be connected with direct relationships (Fig. 2B). Thus, we combined transcriptomics and metabolomics for these data using Ingenuity Pathway Analysis (IPA Ver. 8.5, Ingenuity Systems, <http://www.ingenuity.com>), which is a web-based bioinformatics software for the identification of biological functions. The datasets of differentially expressed genes and metabolites were combined, to elucidate the interactions between differentially expressed genes and altered metabolites (Fig. 2C), and determine biological changes related to ROS generation. The integration of the metabolic profile and transcriptome revealed a direct correlation between the metabolites and genes related to ROS generation. These results demonstrate the importance of metabotranscriptomics for the detailed analyses of nanotoxicity. Although integrated omics approaches facilitate a comprehensive analysis of cellular pathways in biological systems, several new challenges need to be overcome before they can be used in nanotoxicity investigations.

CONCLUSION

Here, we reviewed the cutting-edge metabotranscriptomic approaches for nanotoxicity evaluation. The traditional methods of detection are limited by their ability to measure nanotoxicity. The introduction of advanced tools has led to the integration of omics, especially metabolomic profiling and transcriptomics, to provide extensive information on biological conditions. The technological progress in the molecular diagnostics can pave the way to the development of additional omics techniques. In addition to the integration of transcriptomics and metabolomics, the combination of transcriptomics, genomics, and proteomics in nanotoxicity studies can be used to facilitate the analyses of subtle changes in cellular physiology and molecular biology. Future studies in nanotoxicity will require the integration and multidisciplinary use of omics methods. This integration is expected to produce major advances in toxicity research and encourage the discovery of novel biomarkers for nanotoxicity for a more complete understanding of the effects of NPs in biomedical studies.

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CONFLICTS OF INTEREST

The authors have no conflicting financial interests to declare.

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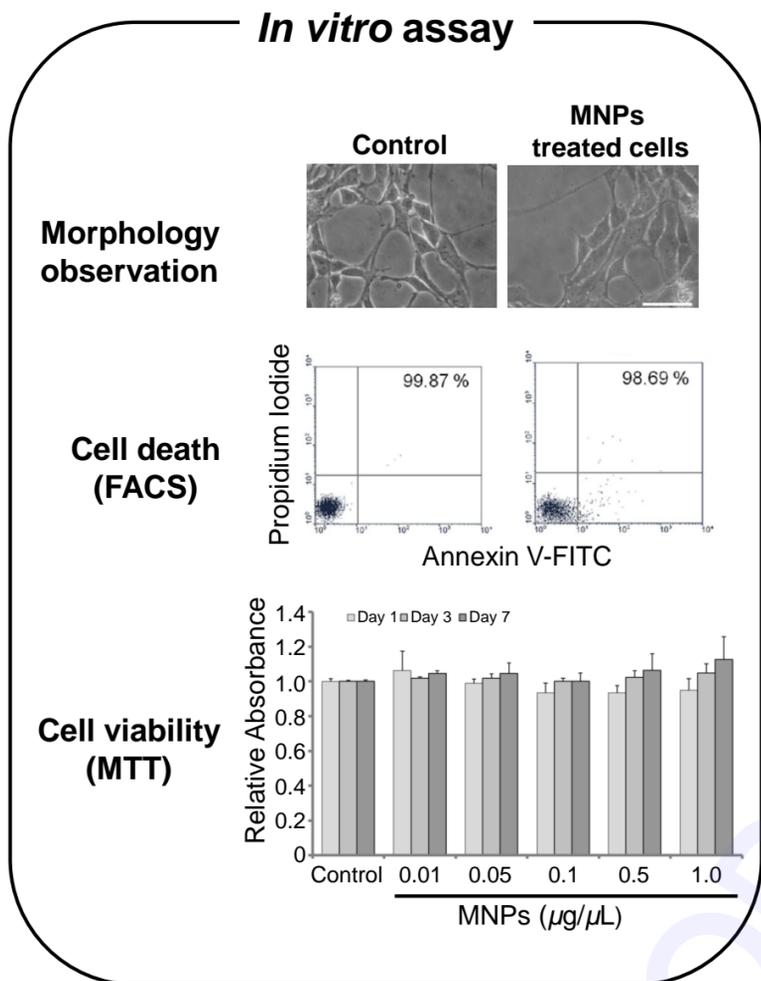
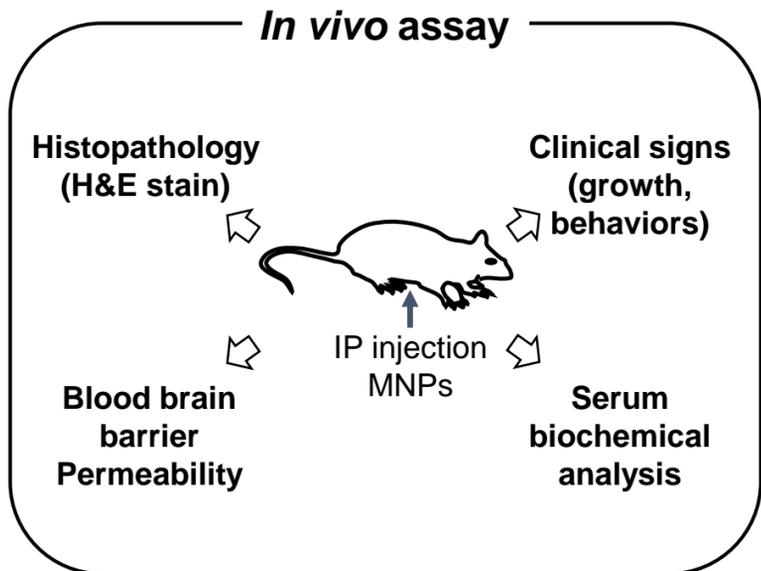
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FIGURE LEGENDS

Figure. 1. Summary diagram comparing conventional methods and metatranscriptomics approach for the assessment of MNPs@SiO₂(RITC)-induced nanotoxicity (27, 29-31). MNPs: MNPs@SiO₂(RITC), IP: Intraperitoneal, TEM: transmission electron microscopy.

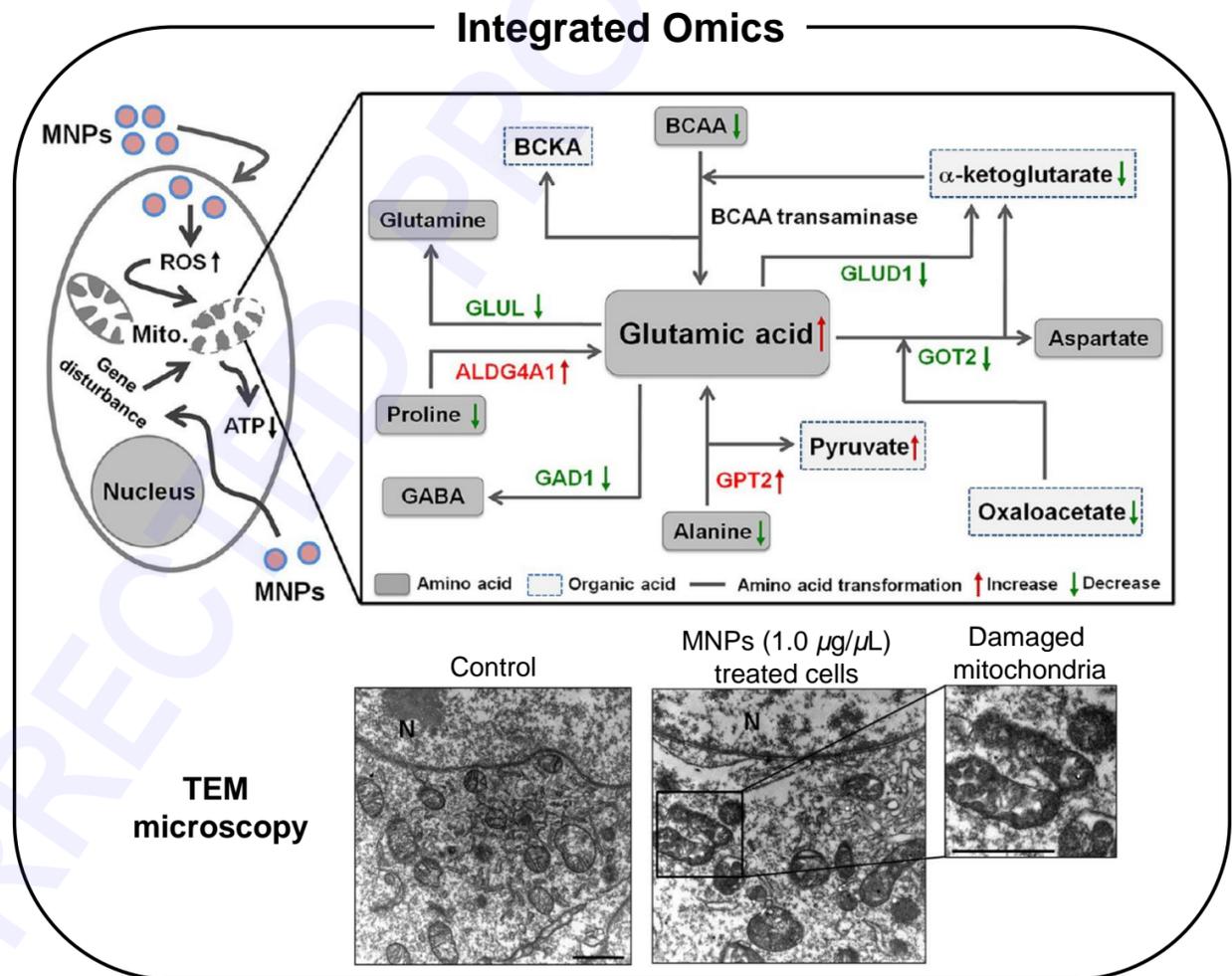
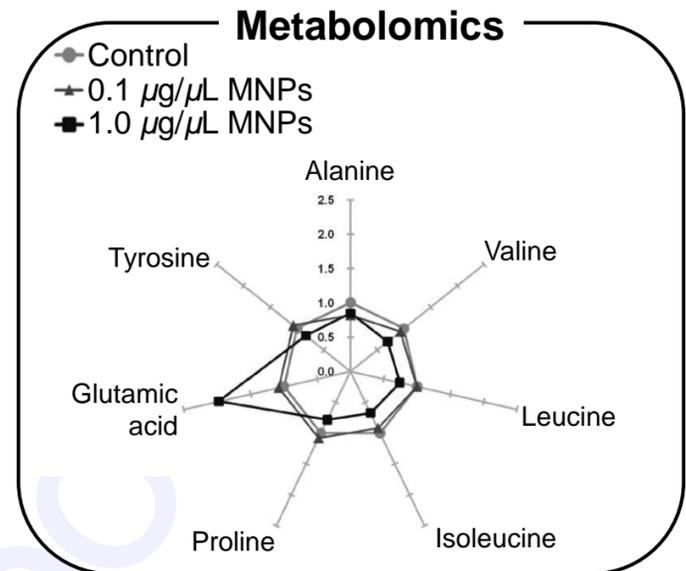
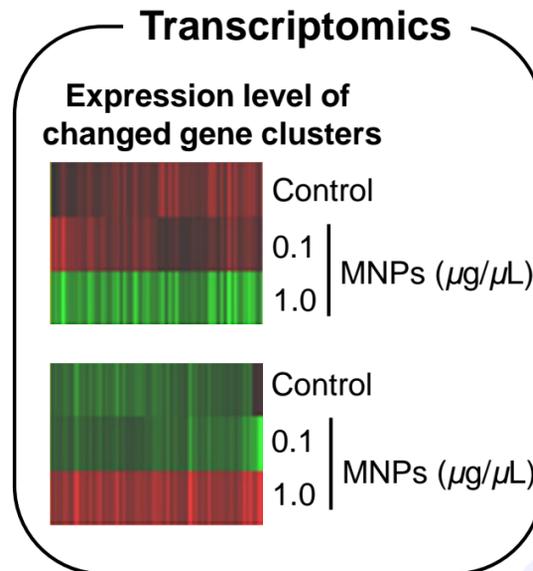
Figure. 2. Bioinformatics of ROS generation using ingenuity pathway analysis (IPA), (A) metabolomics, (B) transcriptomics, and (C) metatranscriptomics based on a previous report (27). Red and green areas indicate up- and downregulated metabolites, respectively, in cells treated with MNPs@SiO₂(RITC) compared with control cells. Differentially regulated metabolites obtained from the metabolic profile (more than a $\pm 20\%$ change) and microarray data (genes with a > 3 -fold change) are shown. In the representation of the genetic networks, the red and green colors indicate up- and down-regulated genes, respectively. Network shape indicates categorization of molecules and function. Information pertaining to the corresponding genes can be found in NCBI (<https://www.ncbi.nlm.nih.gov/>).

Conventional Method

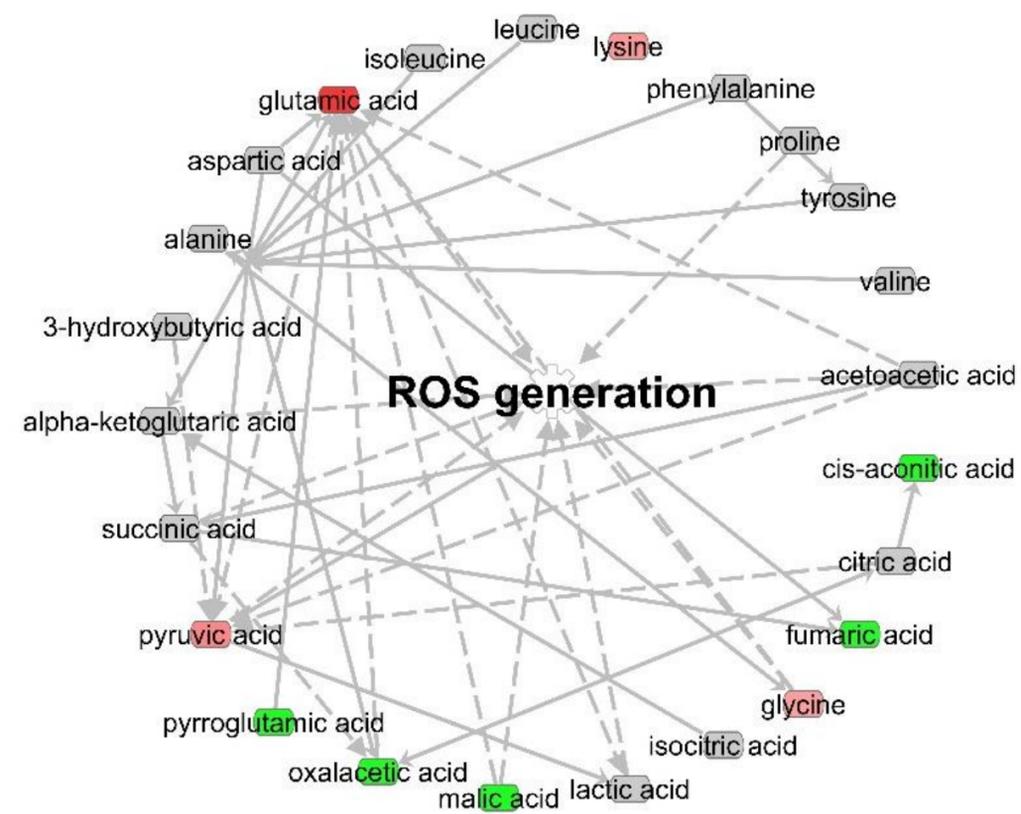
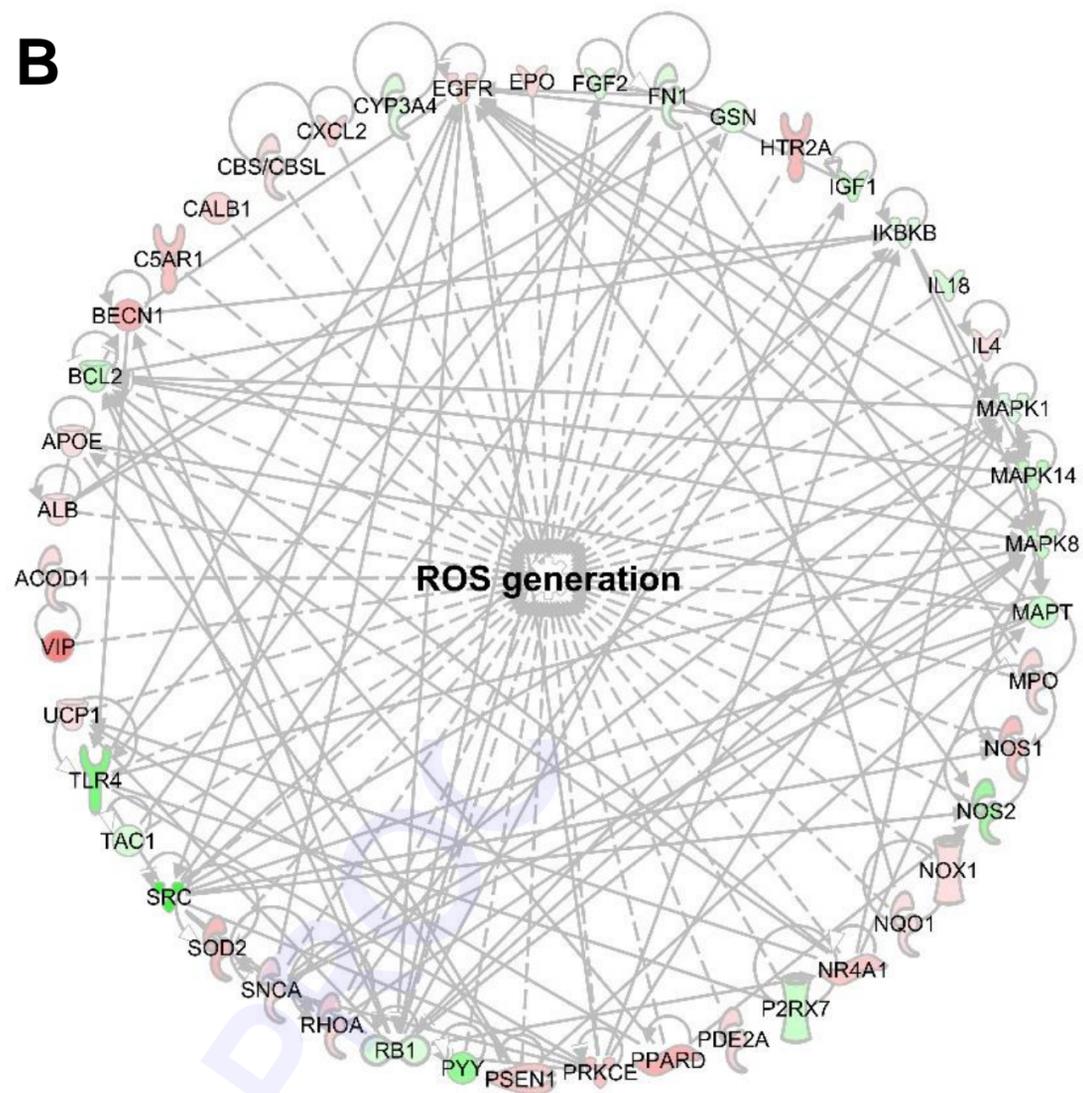


Undetected toxicity

Metabotranscriptomics



Detection of ROS-induced mitochondrial damage

A**B****C**