

# UniVIO: A Multiple Omics Database with Hormonome and Transcriptome Data from Rice

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Plant hormones play important roles as signaling molecules in the regulation of growth and development by controlling the expression of downstream genes. Since the hormone signaling system represents a complex network involving functional cross-talk through the mutual regulation of signaling and metabolism, a comprehensive and integrative analysis of plant hormone concentrations and gene expression is important for a deeper understanding of hormone actions. We have developed a database named Uniformed Viewer for Integrated Omics (UniVIO: http://univio.psc. riken.jp/), which displays hormone-metabolome (hormonome) and transcriptome data in a single formatted (uniformed) heat map. At the present time, hormonome and transcriptome data obtained from 14 organ parts of rice plants at the reproductive stage and seedling shoots of three gibberellin signaling mutants are included in the database. The hormone concentration and gene expression data can be searched by substance name, probe ID, gene locus ID or gene description. A correlation search function has been implemented to enable users to obtain information of correlated substance accumulation and gene expression. In the correlation search, calculation method, range of correlation coefficient and plant samples can be selected freely.

Keywords: Database • Multiomics • Oryza sativa • Plant hormones • Transcriptome.

Abbreviations: tRNA-IPT. tRNA-isopentenyltransferase; UniVIO, Uniformed Viewer for Integrative Omics.

# Introduction

Plant hormones are signaling molecules that regulate plant growth and development. Plant hormones function, at least partially, through the regulation of the expression of downstream genes, and the underlying molecular mechanisms are being revealed (Shan et al. 2011, Garg et al. 2012). For instance,

cytokinin is perceived by membrane-localized histidine kinases, and the signal is transduced by a His-Asp phosphorelay in a two-component system (Kieber and Schaller 2010, Hwang et al. 2012). Auxin and gibberellin are perceived by nuclear receptors and signal transduction is mediated by the 26S proteasome pathway (Hartweck 2008, Mockaitis and Estelle 2008, Lumba et al. 2010 Hayashi 2012). ABA is perceived by a regulator of type 2C protein phosphatases localized in the nucleocytoplasm; the signal is transmitted by the phosphatase activity and the phosphorylation status of downstream components (Cutler et al. 2010, Umezawa et al. 2010, Qin et al. 2011). Plasma membrane-localized G protein-coupled receptor-type G proteins are also suggested as ABA receptors (Pandey et al. 2009). The spatiotemporal distribution of the signaling systems specifies the situation in which a hormone response can occur. In addition to the signaling systems, hormone concentration and distribution are determinants of hormone action. Hormone metabolism, including biosynthesis, conjugation and degradation, is intricately regulated by feedback loops in space and time to finely control hormone activities.

It has long been known that there are functional interactions between plant hormones, especially between cytokinin and auxin (Loomis and Torrey 1964, Sachs and Thimann 1967), cytokinin and ethylene (Cary et al. 1995), auxin and gibberellin (Weijer 1959), auxin and ABA (Saftner and Wyse 1984), and gibberellin and ABA (Pearce et al. 1987). Current research aims to explain molecular mechanisms of the interactions between plant hormones via the regulation of their signaling and metabolic genes. For instance, the negative regulation of key genes of cytokinin biosynthesis by auxin has been demonstrated to underlie apical dominance in stems (Tanaka et al. 2006, Kitazawa et al. 2008). The cytokinin signaling factor ARR1 directly regulates the expression of the auxin signaling repressor gene SHY2/IAA3 (Taniguchi et al. 2007, Dello Ioio et al. 2008), and auxin regulates a subset of the type-A response regulator genes in Arabidopsis and rice (Müller and Sheen 2008, Zhao et al. 2010, Tsai et al. 2012). Frigerio et al. (2006) showed that auxin up-regulated the expression of gibberellin metabolic

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Table 1 Plant hormones and related compounds analyzed in this study

Hormone	Class	Abbreviation	Physiological property
Abscisic acid	Abscisic acid	ABA	Active compound
Indole-3-acetic acid	Auxin	IAA	Active compound
Indole-3-acetyl-L-alanine	Auxin	IAAla	Deactivated compound
Indole-3-acetyl-L-isoleucine	Auxin	IAlle	Deactivated compound
Indole-3-acetyl-L-leucine	Auxin	IALeu	Deactivated compound
Indole-3-acetyl-L-aspartic acid	Auxin	IAAsp	Deactivated compound
Indole-3-acetyl-L-tryptophan	Auxin	IATrp	Deactivated compound
Indole-3-acetyl-L-phenylalanine	Auxin	IAPhe	Deactivated compound
$N^6$ - $(\Delta^2$ -Isopentenyl)adenine	Cytokinin	iP	Active compound
$N^6$ - $(\Delta^2$ -Isopentenyl)adenine riboside	Cytokinin	iPR	Intermediate
$N^6$ - $(\Delta^2$ -Isopentenyl)adenine ribotides	Cytokinin	iPRPs	Intermediate
$N^6$ - $(\Delta^2$ -Isopentenyl)adenine-N7-glucoside	Cytokinin	iP7G	Deactivated compound
$N^6$ - $(\Delta^2$ -Isopentenyl)adenine-N9-glucoside	Cytokinin	iP9G	Deactivated compound
trans-Zeatin	Cytokinin	tZ	Active compound
trans-Zeatin riboside	Cytokinin	tZR	Intermediate
trans-Zeatin ribotides	Cytokinin	tZRPs	Intermediate
trans-Zeatin-N7-glucoside	Cytokinin	tZ7G	Deactivated compound
trans-Zeatin-N9-glucoside	Cytokinin	tZ9G	Deactivated compound
trans-Zeatin-O-glucoside	Cytokinin	tZOG	Deactivated compound
trans-Zeatin riboside-O-glucoside	Cytokinin	tZROG	Deactivated compound
trans-Zeatin ribotide-O-glucosides	Cytokinin	tZRPsOG	Deactivated compound
cis-Zeatin	Cytokinin	cZ	Active compound
cis-Zeatin riboside	Cytokinin	cZR	Intermediate
cis-Zeatin ribotides	Cytokinin	cZRPs	Intermediate
cis-Zeatin-O-glucoside	Cytokinin	cZOG	Deactivated compound
cis-Zeatin riboside-O-glucoside	Cytokinin	cZROG	Deactivated compound
cis-Zeatin ribotide-O-glucosides	Cytokinin	cZRPsOG	Deactivated compound
Dihydro-zeatin	Cytokinin	DZ	Active compound
Dihydro-zeatin riboside	Cytokinin	DZR	Intermediate
Dihydro-zeatin ribotides	Cytokinin	DZRPs	Intermediate
Dihydro-zeatin-N9-glucoside	Cytokinin	DZ9G	Deactivated compound
Gibberellin A1	Gibberellin	$GA_1$	Active compound
Gibberellin A3	Gibberellin	$GA_3$	Active compound
Gibberellin A4	Gibberellin	$GA_4$	Active compound
Gibberellin A7	Gibberellin	GA <sub>7</sub>	Intermediate
Gibberellin A8	Gibberellin	$GA_8$	Deactivated compound
Gibberellin A9	Gibberellin	$GA_9$	Intermediate
Gibberellin A12	Gibberellin	GA <sub>12</sub>	Intermediate
Gibberellin A19	Gibberellin	GA <sub>19</sub>	Intermediate
Gibberellin A20	Gibberellin	GA <sub>20</sub>	Intermediate
Gibberellin A24	Gibberellin	GA <sub>24</sub>	Intermediate
Gibberellin A44	Gibberellin	GA <sub>44</sub>	Intermediate
Gibberellin A53	Gibberellin	GA <sub>53</sub>	Intermediate



genes in Arabidopsis, and several studies highlighted the cross-talk between auxin and gibberellin signaling and metabolism at the molecular level (Ozga et al. 2009, Weston et al. 2009, O'Neill et al. 2010).

Recent omics analyses further illuminate interactions among plant hormones. Comprehensive transcriptome analysis in Arabidopsis outlined the interplay between major plant hormones (Nemhauser et al. 2006, Goda et al. 2008). Transcriptome analysis of cytokinin responses and metaanalysis of public transcriptome data showed that cytokinin regulates various hormone-related genes including auxin signaling AXR3/IAA17, gibberellin synthesis GA4, gibberellin signaling GAI, ethylene synthesis ACC genes and ethylene signaling ERF/AP2 genes (Brenner et al. 2005, Brenner et al. 2012). Transcriptome and hormone-metabolome (hormonome) analyses in rice gibberellin signaling mutants, such as gid1, gid2 and slr1, showed that both hormone contents and expression of gibberellin, cytokinin, ABA and IAA metabolic genes are affected in at least one of the mutants (Kojima et al. 2009). Therefore, to characterize the global interactive network of plant hormone actions, comparative analyses of the hormonome and transcriptome are essential. Although the comprehensive analysis of plant hormones remains difficult because of their varied chemical and structural properties and low concentrations, recent developments in mass spectrometry and solid-phase extraction enabled us to quantify multiple hormone species in small amounts of plant material (Kojima et al. 2009, Kanno et al. 2010, Chen et al. 2011).

In addition to general transcriptome databases such as the co-expression database ATTED-II (Obayashi et al. 2007, Obayashi et al. 2009, Obayashi et al. 2011), several specialist databases provide information related to the hormone network in plants (Mochida and Shinozaki 2011). For example, RiceFOX provides plant hormone profiles of transgenic Arabidopsis plants overexpressing full-length rice cDNAs (T. Sakurai et al. 2011), while the Arabidopsis Hormone Database contains a collection of hormone-related genes (Jiang et al. 2011). Correlation analysis using the web-based tool AtCAST demonstrated correlations between transcriptome data related to hormone responses (Sasaki et al. 2011). However, there is no public platform that provides integrated endogenous plant hormone concentrations and transcriptome data. Here we describe Uniformed Viewer for Integrative Omics (UniVIO), which houses comparative hormonome and transcriptome data analyzed in the same plant samples, and offers useful search functions including correlation search.

## **Database Contents**

UniVIO currently houses contents of four major plant hormones (cytokinin, auxin, gibberellin and ABA) including active forms and their precursors and conjugates (43 compounds in total; Table 1), transcriptome data analyzed in 14 organ parts of rice plants (Fig. 1 and Table 2) and shoots

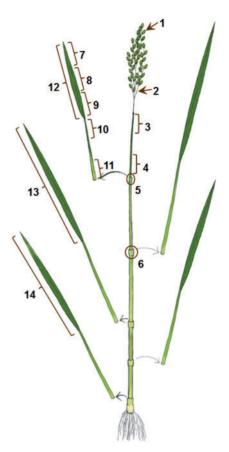


Fig. 1 Illustration of rice plant organs used for hormone and transcriptome analyses. 1, flowers before anthesis; 2, panicle branches; 3, top part of internode I; 4, basal part of internode I; 5, node I; 6, node II; 7, tip of the blade of the flag leaf; 8, middle part of the blade of the flag leaf; 9, basal part of the blade of the flag leaf; 10, top part of the sheath of the flag leaf; 11, basal part of the sheath of the flag leaf; 12, whole blade of the flag leaf; 13, whole blade of leaf 2 counted down from the flag leaf; 14, whole blade of leaf 4 counted down from the flag leaf. The numbers correspond to the numbers in Table 2.

of three gibberellin signaling mutant seedlings (Kojima et al. 2009). Abbreviations of names of substances listed in Table 1 are used hereafter.

## Hormonome and transcriptome analysis

Oryza sativa L. cv. Nipponbare was grown on soil in a greenhouse with irrigation and supplemental artificial light. At the heading stage, tissues were harvested and immediately frozen in liquid nitrogen after measurement of fresh weight. Harvested tissues were stored at -80°C until extraction of plant hormones or total RNA. Plant hormones were extracted, purified and quantified as described previously (Kojima et al. 2009). Microarray analysis was performed using a GeneChip® Rice Genome Array (Affymetrix). Total RNA was extracted from the plant samples using the RNeasy® Mini Kit (QIAGEN). Preparation of labeled target-complementary RNA, subsequent purification and fragmentation were carried out using



Table 2 Sample information of rice organs and mutants from which plant hormone and transcriptome data housed in UniVIO were obtained

Nos.	Plant organs	<b>Abbreviations</b>	Biological replicate			
			GeneChip	Hormone		
Organs at	the heading stage, analyzed in this study					
1	Flowers before anthesis	Flw	3	3		
2	Panicle branches	PBr	3	3		
3	Top part of internode I	InN I, top	3	3		
4	Basal part of internode I	InN I, bsl	3	3		
5	Node I	Nod I	3	3		
6	Node II	Nod II	3	3		
7	Tip of the blade of the flag leaf	FLB, tip	3	3		
8	Middle part of the blade of the flag leaf	FLB, mid	3	3		
9	Basal part of the blade of the flag leaf	FLB, bsl	3	3		
10	Top part of the sheath of the flag leaf	FLS, top	3	3		
11	Basal part of the sheath of the flag leaf	FLS, bsl	2	3		
12	Whole blade of the flag leaf	FLB	1	3		
13	Whole blade of leaf 2 counted down from the flag leaf	LB-2	1	3		
14	Whole blade of leaf 4 counted down from the flag leaf	LB-4	1	3		
Gibberellir	n-related mutants, analyzed by Kojima et al. (2009)					
	Shoot of Taichung 65 (control)	T65	3	3		
	Shoot of gid1-3 mutant	gid1	3	3		
	Shoot of gid2-1 mutant	gid2	3	3		
	Shoot of slr1 mutant	slr1	3	3		

Numbers shown in the first column for organs correspond to the numbers shown in Fig. 1. The abbreviations are those used in UniVIO.

One-Cycle Target Labeling and Control Reagents (Affymetrix). Double-stranded cDNA was prepared from 5  $\mu$ g of total RNA. Hybridization, washing, staining and scanning were performed as described in the supplier's protocol. A 5  $\mu$ g aliquot of fragmented complementary RNA was used for hybridization. These experiments were conducted according to the manufacturers' guidelines.

## Data processing

Plant hormone contents were normalized by fresh weight and expressed as pmol g FW<sup>-1</sup>. The microarray data were extracted as CEL files and imported into GeneSpringGX version 11 (Agilent Technologies), followed by summarization using the MAS5 algorithm but no baseline transformation. The summarized microarray data of all probe sets were extracted as raw signal intensities. The hormone contents and signal intensities were averaged over biological replicates (**Table 2**), and the mean values were included in UniVIO.

## **Data sources**

Microarray data of wild-type organs and gibberellin mutants (Kojima et al. 2009) have been deposited in The National Center for Biotechnology Information Gene Expression Omnibus (NCBI GEO) database under accession numbers GSE41556 and GSE15046, respectively.

Gene descriptions were obtained from the Rice Annotation Project Databases (http://rapdb.dna.affrc.go.jp/; Itoh et al. 2007, Tanaka et al. 2008) and the MSU Rice Genome Annotation Project (http://rice.plantbiology.msu.edu/; Ouyang et al. 2006). Gene ontologies were obtained from the GO Ontology consortium (http://geneontology.org; Ashburner et al. 2000). A matrix table assigning probe IDs and gene locus IDs was obtained from the Rice Oligonucleotide Array Database (http://www.ricearray.org/index.shtml; Jung et al. 2008).

# **Database Construction**

# Concepts and workflow

UniVIO was constructed to present the data in a manner as simple as possible. Since a large number of entries have to be shown in the data viewer, a graphic visualization of the data appeared desirable. On the other hand, a numerical data representation would help users to identify the significance of the data. Therefore, UniVIO presents search results as a heat map table showing hormone contents and gene expression levels.

To construct and organize UniVIO, plant hormonome data and transcriptome data were acquired from plant samples as described above. Then, gene annotations were obtained from



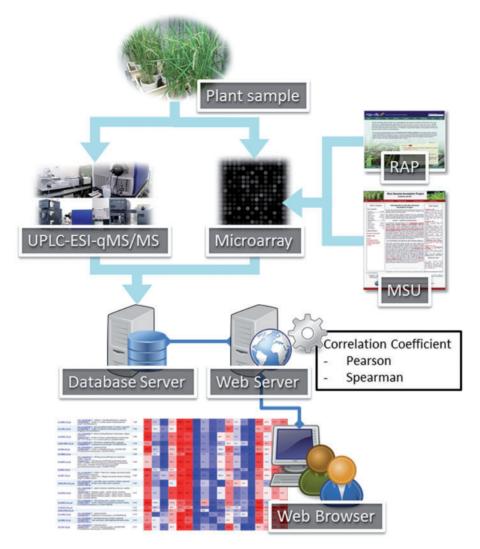


Fig. 2 Schematic representation of the workflow to organize UniVIO. All plant samples were collected under the same conditions, and then were used to analyze the concentration of plant hormones and gene expression levels by mass spectrometry and microarray analysis, respectively. The gene annotation information was obtained from the Rice Annotation Project (RAP; http://rapdb.dna.affrc.go.jp/) and the Rice Genome Annotation Project (MSU; http://rice.plantbiology.msu.edu/), and was combined with the transcriptome data. Users retrieve data from the database server through the Web server using Web browsers. Correlation coefficients are computed on the Web server. Data tables are generated in the Web browser only. UPLC, ultra-performance liquid chromatography; ESI, electrospray ionization; qMS/MS, tandem quadrupole mass spectrometry.

the two public rice genome databases and combined with the transcriptome data. To let users retrieve the hormonome and transcriptome data using a Web browser, the data were housed in a database server connected to a Web server. Correlation coefficients are computed on the Web server. The data table is controlled on the Web browser only. This workflow is illustrated in Fig. 2.

# Database implementation and statistical treatment

Since UniVIO is implemented as a typical Web server-client system using general open source software (Linux, Apache,

MySQL and Perl), users can guery the database via the Internet using a Web browser. The system is compliant with common Web browsers regardless of the operating system, but not with older browsers such as Internet Explorer 7. There is no limit to use; anyone can access all of the data without sign in. To calculate Pearson product-moment correlation coefficients and Spearman's rank correlation coefficients, the GNU Statistic Library (http://www.gnu.org/software/gsl/) and Statistics::RankCorrelation (http://search.cpan.org/dist/ Statistics-RankCorrelation/), respectively, are used. Correlation coefficients are calculated in the Web server on each request (Fig. 2).

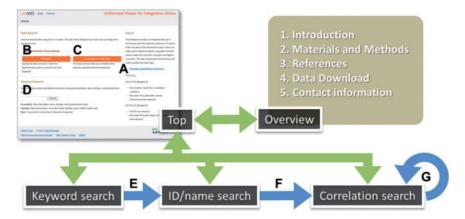


Fig. 3 Site map of UniVIO. Green arrows indicate mutual accessibility between the top page and other pages comprising the UniVIO website. On the top page, there are links to move to the overview (A), ID/name search (B), correlation search (C) and keyword search (D) pages. Blue arrows indicate functional connections between the keyword search and ID/name search (E), between the ID/name search and correlation search (F), and between corelation searches (G), in a a typical workflow in UniVIO.

## **Instructions**

The UniVIO website is comprised of five pages (Fig. 3). The top page provides access to other pages: overview, ID/name search, correlation search and keyword search pages (Fig. 3A, B, C, and D, respectively). The overview page provides data information including materials, methods, raw data and contact information. The search pages support corresponding search functions. When users do not have a list of gene(s) and/or hormone(s) to be queries, they may start with a keyword search. A result of a keyword search is shown in the ID/ name search page (Fig. 3E). Then, when the users find an interesting ID (gene or plant hormone) in the result, they may go directly on to the correlation search using the interesting ID as a query on the ID/name search page (Fig. 3F). This direct correlation search function is also available on the correlation search page (Fig. 3G). The expected workflow is drawn in Fig. 3. Detailed instructions for the search functions are described in the following sections.

# ID/name search

To execute an ID/name search, users move to the search page by clicking a button on the top page (Fig. 3B). UniVIO accepts a single or multiple (up to 1,000) hormone names, gene locus IDs and probe IDs as queries (process 1 in Fig. 4A). Hormone names may be added to a query input form by selecting on a checkbox tree (process 2 in Fig. 4A). Each query can be separated by a space, comma or line break. To focus on particular samples, the plant samples to be shown in a heat map table may be selected using checkboxes (process 3 in Fig. 4A). Color schemes for the heat map can be selected as percentile based or fold change based (process 5 in Fig. 4A). In the former option, red and blue are saturated at 100 and 0 percentiles, respectively; in the latter, the colors are saturated at 3-fold and at one-third, respectively, of the mean values of selected samples. In the current data set of rice organs,

microarray analysis was not or was less replicated in some samples (Table 2). It possibly affects the reliability of gene expression levels and correlations for expression patterns. The sample selection function also enable the exclusion of less-replicated samples when users concern this point.

In a heat map table, data can be sorted by IDs (probe IDs and hormone names) shown in the first column (process 6 in Fig. 4A). To find plant hormones easily in a heat map table, check the box labeled 'Show only phytohormone' located just above the heat map table (process 7 in Fig. 4A). Each ID shown in the resulting table is linked to a correlation search using the ID as a query (process 8 in Fig. 4A).

#### Correlation search

To execute a correlation search, users move to a corresponding search page by clicking a button on the top page (Fig. 3B). UniVIO accepts a single probe ID or hormone name as a query (process 9 in Fig. 4B). Users also may be directed to the results of correlation searches from heat map tables in ID/name searches (Figs. 3F, 4A) and correlation searches (Fig. 3G). To allow free modifications of search conditions, UniVIO provides several options: calculation methods (parametric Pearson product-moment correlation or nonparametric Spearman's rank correlation), range of correlation coefficients (from -1 to 1) and plant samples (processes 10, 11 and 12 in Fig. 4B). All of these options affect the correlation coefficients. If the number of results is >1,000 (the upper limit in UniVIO), no result is shown. To see results, one has to narrow the range of correlation coefficients. Instructions concerning the resulting heat maps are basically the same as for ID/name searches (see above and Fig. 4A).

# Keyword search

Keyword searches can be started from the top page (Fig. 3C). UniVIO accepts a single word of three or more characters, or multiple words as a query, and will search gene descriptions,



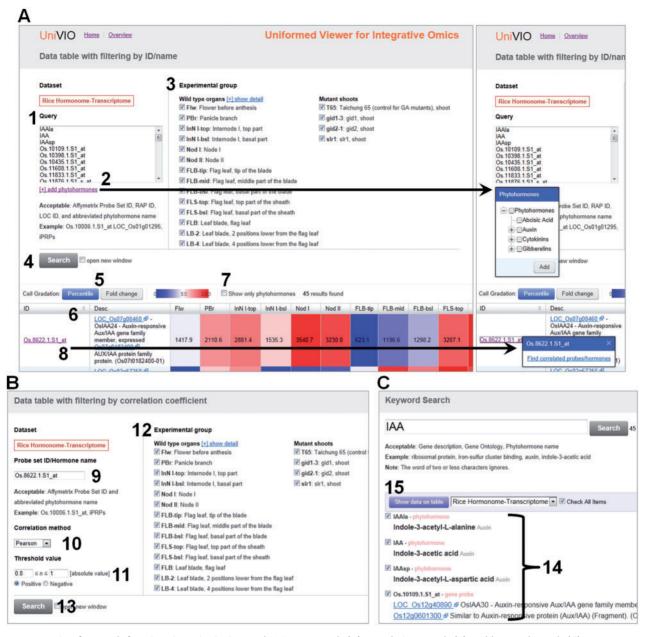


Fig. 4 Instruction for search functions in UniVIO. Cropped ID/name search (A), correlation search (B) and keyword search (C) pages. 1, query input box for ID/name search; 2, checkbox tree to manage plant hormones; 3, selection of samples to be shown in a heat map table; 4, execution of ID/name search; 5, selection of a coloring method; 6, sorting function by IDs; 7, extraction of plant hormones from the heat map table; 8, direct link to a correlation search result; 9, query input box for correlation search; 10, selection of a method to calculate correlation coefficients; 11, setting a range of correlation coefficients; 12, selection of samples to be used for calculation of correlation coefficients and to be shown in a heat map table; 13, execution of the correlation search; 14, selection of genes and plant hormones shown in a heat map table; 15, execution of ID/ name search using probe IDs and plant hormones selected in the keyword search page.

gene ontologies and hormone names. When a query is submitted from the top page (Fig. 3D), the keyword search page will be displayed on which candidate genes and hormones can be selected (process 14 in Fig. 4C). If the number of candidates exceeds 1,000, they are not shown. To see the candidates, further specify the search conditions by adding or changing keywords. To show a resulting heat map, click 'show data on

table' (process 15 in Fig. 4C). The results page shown resembles that returned by an ID/name search (see above and Fig. 4A).

## Discussion

Among the 43 plant hormone-related compounds analyzed, 13 (ABA, IAA, iPRPs, cZRPs, iPR, tZR, cZR, cZ, cZRPsOG, cZROG,

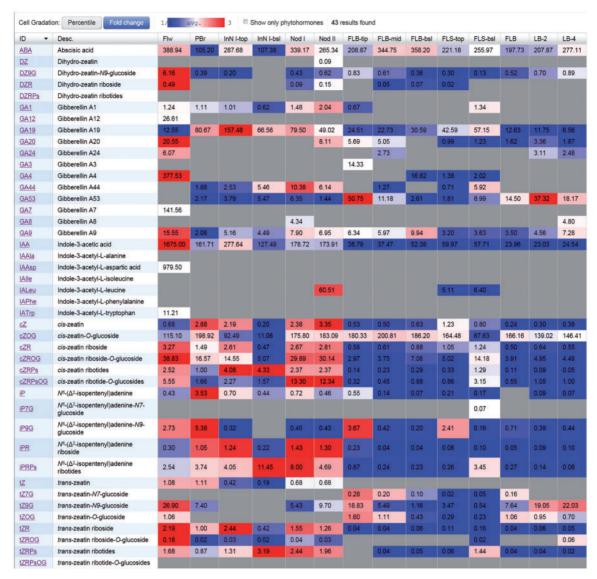


Fig. 5 Contents of plant hormone-related compounds in various rice organs. The heat map is drawn using UniVIO with fold change coloring. Hormone contents are indicated as means of three biological replicates in units of pmol  $g^{-1}$  FW.

cZOG, GA<sub>9</sub> and GA<sub>19</sub>) were detected in the 14 organ parts analyzed (**Fig. 5**). Correlative relationships between the hormones and between hormones and hormone-related genes are summarized in **Figs. 6**, **7** and **Supplementary Fig. S1**, respectively. Some aspects of these correlations are discussed below. The hormonome and transcriptome data in the gibberellin signaling mutants have been discussed in our previous study (Kojima et al. 2009).

## Interhormone correlation in wild-type organs

In a current model of cytokinin biosynthesis that is mostly based on studies of Arabidopsis, cZ is thought to originate from the degradation of prenylated tRNA that has been modified by tRNA-isopentenyltransferase (tRNA-IPT). This pathway is distinct from iP and tZ biosynthesis initialized

by prenylation of adenosine phosphates (Kakimoto 2001, Takei et al. 2001, Kasahara et al. 2004, Miyawaki et al. 2006). However, cZRPs, cZR, cZ, cZRPsOG and cZROG were positively correlated with iPRPs, iPR and/or tZR (Fig. 6). This result suggests two possibilities. First, there could be a novel cZ biosynthesis pathway linked to iP and tZ biosynthesis in rice, and, secondly, the cZ levels in organs could be controlled by mechanisms other than de novo synthesis. However, cZ was strongly negatively correlated with OsIPT9 which encodes a tRNA-IPT (Supplementary Fig. S1). Regarding the latter possibility, reversible inactivation by O-glucosylation could be a candidate pathway controlling the cZ level. In rice, cZOG was the most abundant cytokinin derivative and it was not correlated with other cytokinins (Figs. 5, 6). Putative cZ-O-glucosyltransferases have been



		ABA	IAA	iPRPs	cZRPs	iPR	tZR	cZR	cZ	cZRPsOG	cZROG	cZOG	GA9	GA19
A.D.A	r	1.000		A P. P. S.			31550.55			-			*********	100000000000000000000000000000000000000
ABA	ρ	1.000												
	r	n.s.	1,000											
IAA	ρ	n.s.	1.000											
inno.	r	n.s.	р	1.000										
iPRPs	ρ	n.s.	0.749	1.000		,				-				
cZRPs	r	n.s.	n.s.	0.837	1.000									
CZRPS	ρ	n.s.	0.893	0.858	1.000									
iPR	r	n.s.	n.s.	n.s.	n.s.	1.000			-				7	
IFK	ρ	n.s.	0.747	0.753	n.s.	1.000								
tZR	r	n.s.	n.s.	n.s.	0.730	0.746	1.000							
IZR	ρ	n.s.	0.915	0.694	0.845	0.763	1.000							
cZR	r	n.s.	n.s.	n.s.	n.s.	0.736	0.917	1.000						
CZK	ρ	n.s.	0.771	n.s.	n.s.	n.s.	0.787	1.000						
cZ	r	n.s.	n.s.	n.s.	n.s.	0.914	n.s.	0.678	1.000					
CZ	ρ	n.s.	0.697	n.s.	n.s.	n.s.	n.s.	0.859	1.000					
cZRPsOG	r	n.s.	n.s.	n.s.	n.s.	0.736	n.s.	0.759	0.667	1.000				
CZRFSOG	ρ	n.s.	0.732	n.s.	0.678	0.751	0.882	0.754	n.s.	1.000				
cZROG	r	n.s.	0.715	n.s.	n.s.	n.s.	0.796	0.939	n.s.	0.816	1.000		* * * * * * * * * * * * * * * * * * *	
CZROG	ρ	n.s.	0.851	n.s.	0.735	0.674	0.919	0.859	0.705	0.916	1.000			
c70G	r	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	1.000		
cZOG	ρ	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	1.000		
GA9	r	0.762	0.772	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	1.000	
GNa	ρ	0.807	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	1.000	
GA19	r	n.s.	n.s.	n.s.	0.700	0.707	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	1.000
GMIB	ρ	n.s.	n.s.	0.793	0.717	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	1.000

Fig. 6 Correlations between hormones. Plant hormone-related compounds detected in all organs tested are selected. Pearson product-moment correlation coefficients (upper, r) and Spearman's rank correlation coefficients (lower, ρ) are indicated and highlighted in red (positive) or blue (negative) where they are significant ( $\alpha = 0.01$ ). n.s., not significant.

Probe ID	Symbol	Correlations												
Prooe ID		IPRPs	cZRPs	iPR	tZR	tZR cZR	cZ	cZRPsOG	cZROG	cZOG	IAA	GA9	GA19	ABA
Abscisic acid metabolism		20.000	00000000		F-100 T-100 T			PROPERTY AND		er moreon	27707-00	27.70.70	170700	
Os.1605.1.S1_at	OsZEP1	-0.741	-0.771	n.s.	n.s.	-0.690	n.s.	n.s.	-0.698	n.s.	n.s.	n.s.	n.s.	n.s.
Os.27021.1.S1_x_at	OsAAO3;2	0.737	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	-0.734	n.s.	n.s.	n.s.	n.s.
Os.44431.1.S1_at	OsAAO3;3	-0.721	-0.691	n.s.	n.s.	n.s.	n.s.	-0.674	-0.717	n.s.	n.s.	n.s.	n.s.	n.s.
Os.44431.2.S1_x_at	OsAAO3;3	-0.727	-0.682	n.s.	n.s.	n.s.	n.s.	-0.709	-0.678	n.s.	n.s.	n.s.	n.s.	n.s.
Os.12145.1.S1_at	OsABA8OX1	n.s.	n.s.	n.s.	0.666	n.s.	0.667	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.
Auxin metabolism	Section (Control of the Control of t									au oncurrent				
Os.7114.1.S1_s_at	OsASA2	-0.767	-0.761	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	0.665	n.s.	n.s.	n.s.	n.s
Os.11814.1.S1_at	OsASB2	0.864	0.805	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	-0.818	n.s.	n.s.	n.s.	n.s
Os.5347.1.S1_x_at	OsTAA1;3	n.s.	-0.705	n.s.	-0.687	-0.681	n.s.	n.s.	-0.709	n.s.	n.s.	n.s.	n.s.	n.s
Os.15575.1.S1_at	OsAAO2	n.s.	n.s.	-0.768	-0.744	-0.753	-0.768	n.s.	-0.726	n.s.	n.s.	n.s.	n.s.	n.s
Os.52733.1.S1_at	OsYUCCA7	n.s.	0.712	0.793	0.708	0.811	0.731	0.871	0.760	n.s.	n.s.	n.s.	n.s.	n.s
Cytokinin metabolism	1011-1202			184676	5.7			a .						
Os.8794.3.S1_x_at	OsIPT9	n.s.	n.s.	0.893	n.s.	n.s.	0.850	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s
OsAffx.5165.1.S1_at	OsIPT10	n.s.	-0.668	n.s.	n.s.	n.s.	n.s.	-0.713	-0.692	n.s.	n.s.	n.s.	n.s.	n.s
Os.11222.2.S1_x_at	LOG	-0.705	-0.766	n.s.	n.s.	-0.666	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.
Os.11146.1.S1_at	LOGL7	-0.739	-0.744	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	-0.827	n.s
Os.46093.1.S1_at	LOGL10	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	0.788	n.s.	n.s.	n.s.	n.s.	n.s.	n.s
Gibberellin metabolism														
Os.16326.1.S1_at	CPS	n.s.	n.s.	n.s.	0.733	0.857	n.s.	0.776	0.936	n.s.	0.770	0.680	n.s.	n.s.
OsAffx.27508.22.S1_s_at	KS	n.s.	0.681	0.769	0.730	0.691	0.744	n.s.	n.s.	n.s.	n.s.	n.s.	0.811	n.s
Os.10689.1.S1_at	KAO	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	n.s.	0.979	0.790	n.s.	n.s
Os.19823.1.S1_at	GA20ox2	n.s.	n.s.	0.856	n.s.	n.s.	0.780	0.802	n.s.	n.s.	n.s.	n.s.	n.s.	n.s
Os.21634.1.S1_at	GA2ox3	n.s.	n.s.	0.922	n.s.	n.s.	0.846	0.835	n.s.	n.s.	n.s.	n.s.	n.s.	n.s
Os.47955.1.S1 at	EUIL3	n.s.	n.s.	0.768	n.s.	n.s.	0.725	n.s.	n.s.	n.s.	n.s.	n.s.	0.837	n.s

Fig. 7 Correlations between hormone-related compounds and hormone metabolic genes. Plant hormone-related compounds detected in all organs tested are selected. Microarray probes representing 'Present' as flag information in all samples tested are selected. Pearson productmoment correlation coefficients are indicated and highlighted in red (positive) or blue (negative) where they are significant ( $\alpha = 0.01$ ). n.s., not significant.

found in cZ-rich plants such as maize and rice, but not in Arabidopsis which does not accumulate significant amounts of cZ derivatives (Veach et al. 2003, Hou et al. 2004, Wang et al. 2011, Kudo et al. 2012). Positive correlation of cytokinins was found with IAA (Fig. 6).

GA9 and GA19 were positively correlated with other hormones: GA<sub>9</sub> with ABA and IAA, and GA<sub>19</sub> with iPRPs, cZRPs and iPR (Fig. 6). Since GA9 and GA19 are precursors of the bioactive GA<sub>4</sub> and GA<sub>1</sub>, respectively, this may imply different physiological roles for GA<sub>4</sub> and GA<sub>1</sub> in rice. The synthetic pathway of GA<sub>1</sub> is thought to branch from the GA<sub>4</sub> synthesis pathway by 13-hydroxylation of GA<sub>12</sub> (Kobayashi et al. 2000, Hedden and Thomas 2012). The enzyme responsible for the 13-hydroxylation has not been identified.

Negative correlation was not found in any combinations of the 13 hormone-related compounds when data of all organs were used to calculate correlation coefficients (Fig. 6). However, modification of search conditions, such as selection of samples, allowed us to find negative correlations. For example, negative correlations were detected between tZ and iPRPs (r = -0.750), tZRPs (r = -0.681) or cZRPs (r = -0.881), when flower, panicle branch, and the top and basal part of internode I, node I and node II were selected.



# Correlation between hormones and hormone-related genes in wild-type organs

Among cytokinin precursors including ribotides and ribosides, ribotides such as iPRPs and cZRPs showed negative correlation with cytokinin-activating LOG (Os01g0588900) and LOGL7 (Os05g0541200) (Fig. 7). Since iPRPs and cZRPs are substrates for the enzymes LOG, it was not surprising that iPRP and cZRP levels correlated negatively with the activities of the enzyme genes LOG and LOGL7. However, the roles of many LOG family members in rice have not been determined yet, whereas metabolic profiling of multiple log mutants showed that LOG7 makes the largest contribution to the total cytokinin activation on the whole seedling level in Arabidopsis (Tokunaga et al. 2012). The iPRPs and cZRPs showed strongly positive correlation with the gibberellin signaling repressor gene SLR1 (Os03g0707600) (Supplementary Fig. S1). Although Arabidopsis GAI, a homolog of SLR1, was up-regulated by cytokinin (Brenner et al. 2005), SLR1 was not listed as a cytokinininducible gene in a recent transcriptome analysis (Garg et al. 2012). While cytokinin derivatives, IAA, GA9 and GA19 showed positive or negative correlations with several IAA, gibbellin and/ or ABA metabolic genes, ABA did not show any correlation with any hormone metabolic genes we tested (Fig. 7). Neither did ABA correlate significantly with any hormone signaling genes (Supplementary Fig. S1).

Consistent with our previous analysis (Kojima et al 2009), IAA was present at high levels in flower (Fig. 5). Expression of several IAA biosynthetic genes also appeared high in flowers: OsASA1 (Os.4176.1.S1 at, Os03g0826500), OsASB1 (Os.6876. 1.S1\_at, Os04g0463500), OsTAA1;4 (Os.13941.1.S1\_x\_at and Os.13941.1.S1 s at, Os05g0169300), OsYUCCA4 (Os.22585. 1.S1\_at, Os01g0224700) and OsYUCCA5 (OsAffx.31989.1. S1 at, Os12g0512000) (Supplementary Fig. S2A). These results are in line with the reported abundance of IAA in anthers and the expression of OsTAA1;4 and OsYUCCA4 at later stages of pollen development (Hirano et al. 2008). IAA-amino acid conjugates were detected in specific organs: IAAsp and IATrp in flowers, and IALeu in the top and basal parts of the sheath of the flag leaf as well as in node II (Fig. 5). Among the IAA-amino acid synthase (GH3) family, OsMGH/OsGH3-8 (Os.11798.1.S1\_ at, Os07g0592600), which plays an important role in rice floret fertility (Yadav et al. 2011), showed high expression in flowers (Supplementary Fig. S2B). This result is consistent with a previous study (Jain et al. 2006).

## **Perspective**

Obviously, the regulative network of plant hormone metabolism and gene expression plays a pivotal role in the orchestration of physiological and morphological functions in plants. However, because of its complexity, we have only begun to understand this network. Integrative analyses of the plant metabolome and transcriptome have already contributed to studies on metabolic regulation, such as enviromental response (Yamakawa and Hakata 2010, Schlüter et al. 2012) and spatial

control (Matsuda et al. 2010, Pick et al. 2011), and on secondary metabolism, such as glucosinolate (Hirai et al. 2004, 2005) and terpenoid (Rischer et al. 2006) matabolism. Thus, integrative omics is a valuable method to figure out the complex network.

There are several web-based platforms to analyze and visualize correlations in rice omics data. For instance, OryzaExpress (http://riceball.lab.nig.ac.jp/oryzaexpress/) integrating public rice omics data provides a tool to visualize the gene co-expression network (Hamada et al. 2011). KaPPA-View4 (http://kpv.kazusa.or.jp/kpv4/) can visualize gene expression, metabolite accumulation and gene-gene and metabolite-metabolite correlation in a matabolic pathway map based on transciptome and metabolome data uploaded by users (N. Sakurai et al. 2011). Our UniVIO provides both a comparative hormonome-transcriptome data set and a web application supporting a correlation search function, enabling exploration of positive or negative correlations between the concentration of a plant hormone-related compound and gene expression level. However, further accumulation of comparative hormonome-transcriptome data concerning organ/tissue development and environmental response, and additional mutants in rice and other model plants is needed to increase the utility of UniVIO. We are going to add new data annually, and Arabidopsis data will be installed within 2 years.

# Supplementary data

Supplementary data are available at PCP online.

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