Combined analysis of the time-resolved transcriptome and proteome of plant

pathogen *Xanthomonas oryzae* pv. *oryzae*

Seunghwan Kim[1], Wooyoung Eric Jang[2], Min-Sik Kim[2, 3§], Jeong-Gu Kim[1§], Lin-Woo Kang[4§]

[1]Genomics Division, National Academy of Agricultural Sciences (NAAS), Rural Development

Administration (RDA), Jeonju 54874, Republic of Korea; [2]Department of Applied Chemistry,

Kyung Hee University, Yongin 17104, Republic of Korea; [3]Department of New Biology,

DGIST, Daegu 42988, Republic of Korea; [4]Department of Biological Sciences, Konkuk

University, 1 Hwayang dong, Gwangjin-gu, Seoul 05029, Republic of Korea

[§]Corresponding authors:

Dr. Lin-Woo Kang

Department of Biological Sciences, Konkuk University, Seoul 05029, Republic of Korea; Tel.

82-2-450-4090; Fax: 82-2-444-6707; E-mail: lkang@konkuk.ac.kr

Dr. Jeong-Gu Kim

Genomics Division, National Academy of Agricultural Science (NAAS), Rural Development

Administration (RDA), Jeonju 54874, Republic of Korea; Tel. 82-63-238-4566; E-mail:

jkim5aug@korea.kr

Dr. Min-Sik Kim, Department of New Biology, DGIST, Daegu 42988, Republic of Korea; E-

mail: mkim@dgist.ac.kr

Running title: analysis of transcriptome and proteome in P-activated Xoo

Abstract

Xanthomonas oryzae pv. *oryzae* (Xoo) is a plant pathogen responsible for causing bacterial

blight in rice. The immediate alterations in Xoo upon initial contact with rice are essential for

pathogenesis. We studied time-resolved genome-wide gene expression in pathogenicity-

activated Xoo cells at the transcriptome and proteome levels. The early response genes of Xoo

include genes related to cell motility, inorganic ion transport, and effectors. The alteration of

gene expression is initiated as early as few minutes after the initial interaction and changes with

time. The time-resolved comparison of the transcriptome and proteome shows the differences

between transcriptional and translational expression peaks in many genes, although the overall

expression pattern of mRNAs and proteins is conserved. The discrepancy suggests an important

role of translational regulation in Xoo at the early stages of pathogenesis. The gene expression

analysis using time-resolved transcriptome and proteome provides unprecedented valuable

information regarding Xoo pathogenesis.

Keywords: Xoo-rice interactions, proteome and transcriptome, time-resolved gene expression,

Xanthomonas oryzae pv. *oryzae*, pathogenicity, translational regulation

Introduction

Rice (*Oryza sativa* L.) is the most widely consumed staple food, sustaining two-thirds of the world's population (Jackson, 2016). Bacterial blight of rice caused by *Xanthomonas oryzae* pv. *oryzae* (Xoo) is a devastating disease for which an effective pesticide has not been developed yet; it is known to cause severe yield losses of up to 50% in several rice-growing countries (Oliva et al., 2019). The demand for rice is expected to increase by at least 25% by 2030, owing to the rapidly growing world population, environmental stress arising in response to climate change, and pathogen pressure (Li, Wang, & Zeigler, 2014). Further, the Green Revolution has resulted in a shift in rice cultivation, from varied traditional landraces to limited high-yielding varieties, via artificial selection. This has resulted in the co-evolution of crop pathogens including Xoo with the selected host races in the modern agricultural ecosystem (Quibod et al., 2020).

The pathogen-host system of Xoo and rice serves as an ideal agricultural model to study crop diseases in a field setting at the molecular level; this is facilitated by the early elucidation of the whole genome structure of both Xoo and rice (Jackson, 2016; Lee et al., 2005). Xoo typically invades rice leaves through the wounds or hydathodes and replicates in the xylem vessels to cause disease (Mew, Alvarez, Leach, & Swings, 1993). Rice contains a two-tiered innate immune system, consisting of pathogen-associated molecular pattern- and effector-triggered immunity, which protects against Xoo and initiates the hypersensitivity response at the infection site (Jones & Dangl, 2006). In Xoo-rice interactions, Xoo injects effectors into rice cells to modulate the cellular activities of the host to promote pathogenesis (Tsuge, Furutani, & Ikawa, 2014). The early interactions between Xoo and rice at the infection site determine the fate of infection, i.e., occurrence of disease or initiation of the immune response. The environmental conditions prevalent at the site of infection are varied and complex, and our

understanding of alterations in the Xoo cells in response to the initial interactions with rice is

limited.

Transcription and translation are tightly coupled in bacteria and can occur simultaneously in

the cytosol (Hershey, Sonenberg, & Mathews, 2019). Although proteins are the final functional

products of genes, the quantity of specific mRNA molecules often represents the expression

level of a gene at a given time point with the well-established RNA sequencing (RNA-Seq)

technology. In comparison, high resolution mass spectrometry-based quantitative proteomics

is a more recent analytical technique, and still has a lower coverage of protein products,

requires greater sample quantity, and is more expensive (Schubert, Rost, Collins, Rosenberger,

& Aebersold, 2017).

We had previously developed an *in vitro* pathogenicity assay to recapitulate Xoo-rice

interactions at the site of infection by treating Xoo cells with the rice homogenate (RLX), and

assessed the time-resolved changes in the transcriptome (S. Kim et al., 2016; S. H. Kim et al.,

2011; S. Kim et al., 2013). The *in vitro* pathogenicity assay provides high signal to noise data

with Xoo cells synchronized with respect to the timing of pathogenicity activation.

Transcriptome data from RNA-Seq experiments revealed that most virulence genes of Xoo

were upregulated within an hour of the initial interaction with RLX, and these upregulated

genes were related to bacterial motility, inorganic ion transport, hypersensitive response and

pathogenicity (*hrp*), bacterial toxins and effectors of avirulence (*avr*), plant cell wall

degradation, and extracellular polysaccharide synthesis and secretion (S. Kim et al., 2016).

Here, we expand the study of gene expression in pathogenicity-activated (P-activated) Xoo

from the transcriptome to the proteome. The time-dependent expression of specific mRNAs

and proteins represents the kinetics of transcriptional and translational expression of certain

genes and allow the visualization of the immediate alterations in gene expression in P-activated

Xoo.

Results

Time-resolved proteome data

We coupled LC-MS/MS technology with an *in vitro* assay system to obtain the time-resolved proteome data for P-activated Xoo cells (Scheme 1). The *in vitro* assay system recapitulated the initial interaction between Xoo cells and damaged rice leaf tissues at the site of infection; this was achieved by adding fresh RLX—prepared by grinding the leaves of a Xoo-susceptible rice cultivar (Milyang 23) in liquid nitrogen—to a Xoo cell culture in the mid-exponential phase. Samples for proteome analysis were collected from RLX-treated (P-activated) and untreated (control) Xoo cells at 0, 30, 60, 90, and 120 min after RLX treatment (Table S1).

Analysis on UniProt revealed that the total 4,956 predicted open reading frames in the Xoo genome (KACC10331) corresponded to 4,382 proteins in the proteome, of which 2,589 proteins (59%) were identified for at least one time point and 2,296 proteins (52%) were detected in both replicates (Figure S1A). Median sequence coverages for total identified proteins were 24% and 23% for each of the independent duplicate experiments (Figure S1B). A total of 20,963 and 19,684 non-redundant peptides were identified in both replicates, with 47,025 and 41,428 peptide-spectrum matches, respectively (Figure S1C).

Protein abundance values obtained after quantile-normalization (Figure S2A) were used for pairwise comparisons; the Pearson's correlation coefficients corresponding to the abundance values showed close correlations (0.98-0.99), indicating comparable cellular concentration of most proteins (Figure S2B). The smallest correlations were observed for the RLX-treated samples at 30 min, indicating greater changes in the proteome during the initial 30 min; this was consistent with the transcriptome data. We further performed a principal component analysis to determine the relationships between the assessed samples. Figure S2C shows that the P-activated sample at 0 min clustered closely with all control samples, whereas the P-

activated samples at other time points were more spread out. The proteome of the P-activated

Xoo cells at 30 min was considerably different from that of P-activated Xoo at other time points;

this was consistent with the results of the pairwise multi-scatter plot (Figure S2B).

Time-supervised hierarchical clustering was performed to determine temporal and

synchronized changes in protein abundance. Two distinct synchronized patterns—one

decreasing and the other increasing in response to RLX treatment—were observed (Figure 1).

The two lists including proteins exhibiting the two patterns were used for STRING analysis,

which produced two interaction networks (Figure S3A-B). Several genes related to cell motility,

inorganic ion transport, and transcriptional regulators were immediately responsive to the

pathogenicity signal (Table S2); these have been described in detail in the discussion section.

Up and down regulated proteins

Quantitative proteome data obtained from P-activated and control Xoo cells at every 30 min

allowed the visualization of the three-dimensional protein expression data in terms of the genes,

time intervals, and expression levels (Figure 2 and Table S3). The expressed protein level of

each gene from P-activated Xoo cells are divided by that of the same gene from control at each

time point to calculate the fold change of time-dependent protein expression level of the

specific gene. For all the open reading frames, approximately 93 (2.0%), 213 (4.5%), and 468

(9.9%) proteins were upregulated by more than 200% (2-fold), 50%, and 20% at 30 min,

respectively, and approximately 7 (0.1%), 93 (2.0%), and 561 (11.9%) proteins were

downregulated to less than 25% (2-fold), 50%, and 80% at 30 min, respectively (Table S4).

In case of the duration of expressed proteins in P-activated Xoo, only 8 (0.2%), 32 (0.7%),

and 75 (1.6%) proteins were upregulated for entire 120 min by more than 200% (2-fold), 50%,

and 20%, respectively and 0 (0%), 1 (0.02%), and 87 (1.8%) proteins were downregulated for

the same 120 min to less than 25% (2-fold), 50%, and 80% , respectively (Table S5), indicating

that most proteins were temporarily up- or downregulated in the *in vitro* assay.

The proteins at 30 min presented the highest change at the expression level within 120 min.

More than 90 proteins (1.9%) were upregulated by more than 2-fold at 30 min in each dataset,

whereas 27 proteins were upregulated in both datasets from the duplicate experiments (Table

S6). Almost half of the 27 upregulated proteins were related to cell motility and ion uptake,

seven were related to chemotaxis and motility, and five were related to transporters and pumps.

At the same 30 min, more than 90 proteins were downregulated by more than 50%, 17 proteins

were downregulated in both datasets (Table S7), including transcription-related proteins—such

as sigma-54 modulation protein and MetE/MetH family transcriptional regulator—and cell

division- and cell cycle-related proteins.

Clusters of Orthologous Groups of proteins (COGs)

To study the global gene expression pattern based on gene function, we superimposed time-

dependent protein expression levels as per the functional categories in the Clusters of

Orthologous Groups of proteins (COGs) database (Figure 2), which were grouped into three

classes (red, yellow, and blue) depending on the observed pattern. To facilitate the comparison

of time-dependent protein expression levels, the expression level at 0 min was set as the

reference level (=1) for each gene, and the fold change in the protein expression level was

calculated at each time point, as in case of the RNA-Seq data analysis (S. Kim et al., 2016).

The most prominent changes in protein expression in Xoo cells were observed for proteins

associated with cell motility (N) and inorganic ion transport and metabolism (P), which were

placed in the red class. In category N, two major expression peaks, indicating more than 2-fold

up-regulation, were detected at 30 and 90 min. In category P, upregulated proteins peaked at

around 30 min. Proteins grouped in the yellow class exhibited moderate changes in their

expression level, and were divided into nine functional categories, including signal

transduction (T), intracellular trafficking, secretion, and vesicular transport (U), energy

production and conversion (C), transcription (K), replication (L), cell wall/membrane/envelop

biogenesis (M), and protein turnover (O). The blue class included proteins associated with other

nine categories, the expression levels of which exhibited little change when compared with

those at 0 min, albeit except for a few proteins.

Comparison of mRNA and protein levels

We compared mRNA and protein levels of each gene in the transcriptome and proteome

(Figure 3; thin dashed red lines of mRNAs and thick dashed and solid black lines of proteins).

Both transcriptome and proteome data were obtained for up to 120 min after RLX treatment.

The experimental procedure used to prepare protein samples for LS-MS/MS required at least

30 min, whereas that for mRNA samples required at least 5 min; the time interval for the

proteome data was set to 30 min, whereas that for transcriptome data was set to 5-30 min. Due

to the different sampling times, the transcriptome data included mRNA levels at additional time

points of 5, 10, 15, and 45 min, whereas the proteome data did not include protein levels

between 0 and 30 min, in which case the protein levels (thick dashed black lines) were

extrapolated based on the protein levels at other time points using non-linear regression.

Flagella and chemotaxis-related genes

Flagella and chemotaxis-related genes encode more than 40 proteins, including structural

components and assembly factors of flagellar hook-basal body and filament and chemotaxis

proteins (Mukherjee & Kearns, 2014). Two dominant hierarchical clusters, with distinct

synchronized patterns of increasing and decreasing proteome concentration (indicated by red

and blue, respectively), were identified (Figure 1). The red cluster of 359 proteins included 21

(53%) of the 40 genes related to the flagellar assembly pathway, of which 20 genes were

included in the red cluster (Table S2).

We grouped the flagella and chemotaxis-related genes into three gene clusters, i.e., groups

I-A, I-B, and II, based on their positions in the genome (Figure 4A). Genes in cluster I-A and

I-B were associated with flagellar machinery, such as flagellar basal body hook and type III

secretion system (T3SS), whereas cluster II included chemotaxis-related genes.

Superimposition of transcriptome and proteome data of cell motility related genes revealed

clearly superimposed upregulated peaks at 30 min (Figure 3A). The transcriptome data

revealed that most flagella and chemotaxis-related genes were regulated in a similar pattern,

i.e., genes in all clusters of I-A, I-B, and II were downregulated at 5 min and upregulated at 30

min. However, the proteome data presented a different expression pattern between clusters I-A

and I-B and cluster II. Proteins in clusters I-A and I-B were upregulated at 30 min, consistent

with the transcriptome data. But proteins in cluster II were upregulated at 90 min, exhibiting a

delay of 1 h when compared with the transcriptome data (Figure S4).

Inorganic ion transport and metabolism genes

Iron uptake genes play a key role in pathogenicity during the early stages of host-pathogen

interactions (Garau et al., 2004). TonB-dependent receptors (TBDRs) are bacterial outer

membrane proteins that bind and transport ferric chelates of siderophores. Several annotated

TBDR genes have been identified in the Xoo KACC10331 genome, including *IroN*, *FyuA*,

FecA, *BtuB*, *FhuA*, *CirA*, and *FepA* (S. Kim et al., 2016). Of these, *FecA* (*Xoo0901*) and *CirA*

(*Xoo3793*) were upregulated and *IroN* genes (*Xoo0394* and *Xoo1784*) were downregulated in

both the transcriptome and proteome data. In the proteome data, *FecA* (*Xoo0901*) and *CirA* (*Xoo3793*) upregulation peaked at 30 min (Figure 5A). *IroN* genes that were downregulated at the transcript level were also downregulated at the protein level (Figure 5B). The expression levels of other paralogs of *FecA* and *CirA* genes were comparable with those in the control in the transcriptome as well as the proteome data (Figure S5), indicating that these genes could respond to different pathogenic signals—which were missing in the *in vitro* system—or be pseudogenes.

Phosphate uptake genes, i.e., *OprO*, *PhoX*, *PstSCAB*, and *PhoU*, were upregulated in the transcriptome (up to 16-fold) at 5-10 min. The proteome data of OprO, PhoX, PstSCAB, and PhoU revealed upregulation up to 3-fold at 30 min (Figure S6).

Expression and secretion of effectors

In *Xanthomonas oryzae* pv. *oryzicola*, AvrBs2 suppresses host immunity and promotes disease development (Ullah et al., 1998). In *Xanthomonas campestris* pv. *vesicatoria*, the role of AvrBs3 is that of a transcription activator-like (TAL) effector that activates the expression of plant immunity genes (Kay, Hahn, Marois, Wieduwild, & Bonas, 2009). The secretion of effectors XoAvrBs2 (*Xoo0168*)—a Xoo ortholog of AvrBs2—and XoAvrBs3 (*Xoo2276*)—a Xoo ortholog of AvrBs3—through T3SS was confirmed upon interaction with RLX (S. H. Kim et al., 2011; S. Kim et al., 2013). In addition to the transcriptome and proteome, the expression and secretion of effectors were assessed using dot blots of Xoo cells transformed with the TAP-tagged effector genes (Figure 6).

The expression of *XoAvrBs2* transcript was upregulated from 5-30 min, while that of the XoAvrBs2 protein was upregulated at 30 min, as is evident from the transcriptome and proteome data. In the dot blots, the levels of TAP-tagged XoAvrBs2 were upregulated by 4-

239   fold at 5 min and the secretion of XoAvrBs2 was detected as early as 15 min. The expression

240   of *XoAvrBs3* transcript was upregulated and peaked from 5-30 min, whereas that of XoAvrBs3

241   protein was maintained at almost the same—or slightly lower levels—at 60 min in the

242   transcriptome and proteome data. Dot blots revealed the secretion of XoAvrBs3 from 15 min.

Discussion

Plant pathogens could exhibit complex responses to the initial interactions with the host under varying biotic and abiotic conditions at the site of infection. The varied immediate response of the plant pathogen is important for successful infection. In this study, we analyzed the immediate time-resolved response of Xoo cells from the initial interaction with rice in terms of gene expression at both the mRNA and protein levels using an *in vitro* assay system, wherein RLX mimicked the damaged rice leaf tissue (Scheme 1).

Of all the predicted open reading frames in Xoo, we quantified approximately 83% and 47% of the mRNAs and proteins, respectively, in a time-dependent manner. A good correspondence of overall expression pattern was observed between the mRNA and protein levels of genes. The expression pattern was more synchronized for genes closely located in the genome (Figure 4 and S4); this could be attributed to the polycistronic gene structure in bacteria. With respect to the rapidity of pathogenic gene expression in response to the signals triggered upon interaction with the host, both transcriptional and translational machineries responded immediately to the interaction, at the earliest assessed time points of 5 and 30 min in the transcriptome and proteome data. The secretion of effectors was also found at 15 min in dot blots. Both mRNAs and proteins presented the greatest variation in their levels during the initial 30 min.

The Xoo genes exhibiting the most rapid responses to the initial interaction with RLX, included genes associated with cell motility and inorganic ion uptake, and genes coding for effector molecules. All these three functional categories of genes are closely related to the early stages of pathogenesis. The genes associated with cell motility could be responsible for the migration and accumulation of the Xoo cells at the site of infection via the damaged xylem tissues or exposed hydathodes in rice. As inorganic ions, such as iron and phosphate ions, function as essential cofactors in all living organisms, bacterial pathogens and host rice cells

compete to obtain and secure the limited resources available. Effector molecules are more

directly related to pathogenesis and are injected by the Xoo into the rice cells via pili-like T3SS;

these molecules modulate the immune system of rice.

The earliest available time-point for comparing the gene expressions in terms of mRNA and

protein levels was 30 min after the RLX treatment. At 30 min, 290 mRNAs were upregulated

by more than 2-fold in the transcriptome, compared with 93 proteins in the proteome (Table

S4), and the average fold change in the expression of the upregulated genes in the transcriptome

and proteome was similar, i.e., 3.3-fold and 3.4-fold, respectively (Table S6). The number of

upregulated transcripts is much higher than—even though we figure into our calculations the

higher coverage of quantified genes in transcriptome than in proteome—that of the proteins at

30 min, which indicates that not all the upregulated mRNAs are simultaneously translated to

proteins.

The difference in the gene expression in terms of the mRNA and protein levels at a given

time point suggests the existence of an additional translational regulation step—after the

transcriptional regulation—although transcription and translation can occur simultaneously in

bacteria. For some genes, the mRNA and protein expression peaks were observed at varying

time points. Bacterial pathogenesis appears to involve a fine-tuning mechanism after

transcription, which could help adjust the expression of the early-responsive genes under

varying biotic and abiotic environmental conditions.

Genes related to flagella and cell motility are clustered in the bacterial genome. In Xoo, three

clusters are observed, i.e., I-A, I-B, and II. On superimposing the mRNA and protein levels in

time, the expression of mRNAs and proteins peaked at 30 mins for genes in the clusters I-A

and I-B, the expression of mRNAs and proteins peaked at 30 and 90 min, respectively, for

genes in cluster II. A translational regulation step might be involved that determines the time

for the translation of specific mRNAs—coded by flagella and cell motility-related genes—into

proteins depending on the priority of each gene.

The hierarchy of the expression of flagella and cell motility-related genes has been

extensively studied in the transcriptome of *E. coli*, where sigma and anti-sigma factors are

known to be the key transcriptional regulators (Mukherjee & Kearns, 2014). The overall

organization of flagella and chemotaxis-related genes is different between *E. coli* and Xoo

(Figure S7). In Xoo, the cluster of *fliE-R* genes (cluster I-B) is positioned just downstream to

the *flgB-L* genes (cluster I-A), whereas in *E. coli,* chemotaxis genes are present between the

two. *flhDC* genes do not have any orthologs in Xoo, and a different transcriptional regulator,

i.e., the *fleQ* gene, is present. Based on the nomenclature used in *E. coli*, several Xoo genes

classified as Class III genes are located at different positions and their expression regulation is

also different. In Xoo, ribosomes or other translation factors might recognize certain unknown

priority signals in mRNA transcripts for the translational regulation.

The immediate upregulation of inorganic ion uptake genes, especially iron uptake genes,

may aid Xoo cells to obtain the essential cofactor ions, when they are competition with the host

cells (Liu, Kong, Wu, & Ling, 2020). The iron ion is essential for photosynthesis and respiration

and needed by many redox enzymes for most organisms on Earth. Pathogens commonly use

iron chelating molecules or siderophores to derive this scarce inorganic cofactor from the hosts.

The leakage of iron ions from damaged leaf tissues might serve as an important signal of

initiating infection and might provide an opportunity to secure essential iron ions for the Xoo

cells.

In comparison with the flagella and cell motility-related genes that are closely clustered in

the genome and exhibit coordinated expression levels, the ion uptake genes are dispersed across

the genome and are independently expressed (Figure S8). The separation of these genes

facilitates independent regulation. Compared with the flagella and motility-related genes, the inorganic ion uptake genes exhibited great variation in the mRNA levels but similar protein expression levels (Figure 3).

Effectors are key molecules for pathogenicity that modulate the host immune responses after infection. The protein expression and secretion of the Xoo effectors XoAvrBs2 and XoAvrBs3 were assessed using dot blots, which enabled the monitoring of effector proteins at the early stages of the initial interaction, i.e., within 30 min of RLX treatment. The secretion of effectors was observed from 15 min after application of the pathogenic stimulus. In the P-activated proteome data, the expression of effector proteins was maintained at a similar level to that at 0 min; this may be attributed to the similar protein synthesis and secretion rates.

The effector genes are also dispersed across the genome, like the ion uptake genes (Figure S9). Interestingly, transposase genes are located close to the effector genes; these may aid the effector genes to transpose through the bacterial genome and plasmids. The expression of XoAvrBs2 was upregulated 5 min after RLX treatment, and its secretion was detected in the culture medium at 15 min. Although the cellular protein level was upregulated by approximately 20% in the proteome at 30 min, the dot blot showed an upregulation of 4-fold at 5 min. The secreted XoAvrBs2 exhibited a 16-fold increase at 120 min. In case of XoAvrBs3, the cellular protein level decreased by approximately 20% in the proteome at 60 min, whereas the dot blot revealed a 16-fold increase in XoAvrBs3 secretion at 120 min. The experimental methods for proteome and dot blots analysis are different (S. H. Kim et al., 2011; S. Kim et al., 2013). For the dot blots analysis, TAP-tagged *XoAvrBs2* and *XoAvrBs3* genes were introduced into Xoo cells via the plasmid having the endogenous promotor and expressed from the plasmid. In the proteome analysis, XoAvrBs2 and XoAvrBs3 proteins were expressed from the endogenous *XoAvrBs2* and *XoAvrBs3* genes in the Xoo genome.

Gene expression involves sequential transcription and translation. In bacteria, with respect

to post-transcriptional and post-translational modification, mRNAs without a cap at 5' end and

a poly A tail at 3' end have a short half-life—as short as few minutes—and proteins undergo

only limited post-translational modifications. The limited post-translational modifications in

bacterial proteins impose a pressure on a nascent protein from the ribosome to take a functional

form immediately. The present study revealed that genes related to cell motility and inorganic

ion uptake, and genes coding for effector molecules of Xoo are the first to respond to the initial

interactions with RLX, and play an essential role in the following processes (Figure 7): (1)

invasion of the Xoo cells into the rice leaf tissues, (2) securing the limited cofactors, and (3)

modulating the immune responses of the host to favor pathogenesis. This combined analysis of

the time-resolved transcriptome and proteome of Xoo during the initial interaction with rice

tissues provides valuable insights into the pathogenic mechanism of Xoo.

Materials and Methods

Bacterial strain and culture conditions

Xanthomonas oryzae pv. *oryzae* (Xoo) strain KACC10331, consisting of 4,941,439 nucleotides and 4,733 open reading frames, without any apparent autonomous plasmids, was obtained from the Korean Agricultural Collection Center (KACC) (Lee et al., 2005). The bacteria were cultured in nutrient broth (Difco, Detroit, MI, USA) or Yeast Glucose Cm Agar (YGC) (2.0% D-(+)-glucose, 2.0% $CaCO_3$, 1.0% yeast extract, and 1.5% agar) at 28°C.

Construction of expression vector and transformation of Xoo

The effector genes *XoAvrBs2* (*Xoo0168*) and *XoAvrBs3*—including the promoter region (from -149 and -750 bp to the start codon of the respective gene)—were amplified by PCR, and ligated into the pGEM®-T Easy Vector (Promega). The cloned sequences were verified and then digested with *Kpn*I and *Sac*I, and the products were ligated into the pHM1-XTAP-T_{gap} vector. The recombinant vectors were purified and introduced into Xoo strain KACC10331 by electroporation, using Gene Pulser II (Bio-Rad, Hercules, CA) with a 0.2 cm-gap cuvette at 2.5 kV cm^{-1}. Xoo cells were then diluted immediately with 1 mL Super Optimal Broth (SOC) medium and incubated at 28°C with agitation for 2 h. Cells were then recovered from the culture medium and plated on nutrient broth agar plates containing 50 µL mL^{-1} spectinomycin and incubated at 28°C for 4 d.

The transformants were cultured in 100 mL of nutrient broth up to the mid-exponential phase (A_{600} = 0.5). Cells were harvested by centrifuging 1 mL of the cell culture at 12,000 rpm and 4°C for 5 min. Harvested cells were washed once with phosphate-buffered saline (PBS) at pH 7.2, resuspended in 200 µL PBS, and then sonicated. Protein samples were serially diluted using 2 M urea or PBS in 96-well plates and then transferred to a polyvinylidene difluoride

membrane (PVDF; 0.2 μm, Bio-Rad) using a 96-well vacuum dot-blotter (Bio-Rad). The

membrane was then washed thrice with PBS, blocked with 5% skim milk for 30 min, and

subjected to a one-step immuno-affinity reaction using the rabbit peroxidase-anti-peroxidase

soluble complex antibody (Sigma-Aldrich, St. Louis, USA). The membrane was developed,

and bound antibodies were detected by chemiluminescence.

Treatment of rice leaf extract for proteome analysis

Oryza sativa L. cv. Milyang 23, a Xoo-susceptible rice cultivar, was used for performing

proteome analysis. Rice plants were grown in a paddy field at Jeonju in South Korea

(35°49'52.0"N 127°03'55.6"E) until panicle initiation (approximately 8 to 9 weeks). Forty

clumped rice leaves were harvested and homogenized with liquid nitrogen using a mortar and

pestle. One-gram aliquots of the resulting homogenate (RLX) were transferred to Eppendorf

tubes and stored at -80°C. Xoo was cultured (100 mL) in nutrient broth up to the mid-

exponential phase ($A_{600} = 0.5$) in a shaking incubator at 28°C and 200 rpm, and RLX (2 g) was

then added to the culture medium. The culture (100 mL) was subjected to sequential filtration

through a gauze, 40-μm nylon cell strainer (FALON, New York, USA), and 5-μm syringe filter

(Sartorius, Germany) to remove RLX (0, 30, 60, 90, and 120 min after RLX addition). The

filtered culture (100 mL) was centrifuged at 10,000 ×g and 4°C for 10 min. Duplicate samples

were obtained for each time point from two independent experiments.

Sample preparation for proteome analysis

The harvested samples were lysed in a lysis buffer (9 M urea prepared in 20 mM HEPES (pH

7.5), supplemented with protease inhibitor cocktail (Complete mini, Roche) and phosphatase

inhibitor (PhosSTOP, Sigma-Aldrich)) and sonicated on ice. The exact amount of proteins in

each sample was determined using the bicinchoninic acid assay. Protein integrity was

confirmed by SDS-PAGE and 200 µg of protein from each sample was used for analysis. The

disulfide bonds were reduced by treatment with 10 mM dithiothreitol for 1 h, and incubation

with 30 mM iodoacetamide (30 min in the dark) was performed to alkylate free sulfhydryl

functional groups. Samples were diluted with triethylammonium bicarbonate buffer (pH 8.0)

in a manner such that the final urea concentration was 1.5 M. Proteins were digested using MS

grade trypsin (Thermo Fisher Scientific) at a protein to enzyme ratio of 50:1 for 12 h at 37°C.

The reaction was quenched by lowering the sample pH ($<$3) using trifluoroacetic acid. The

obtained peptides were desalted using a C18 spin column (Harvard) to remove salts and other

contaminants, and the purified peptides were dried. Then, they were isotopically labeled using

the 10-plex tandem mass tag (TMT, Thermo Fisher Scientific), as per the manufacturer's

protocol. The reaction was allowed to continue for 2 h at room temperature and TMT-labeled

samples were subsequently dried in a SpeedVac concentrator. Chemical labeling with TMT

was confirmed by liquid chromatography-tandem mass spectrometry (LC-MS/MS), and the

samples were pooled and fractionated using a basic reverse phase liquid chromatography

(RPLC) system. The pooled TMT-labeled peptide mixture was resuspended in 10 mM

ammonium formate and fractionated into 12 fractions using a C18 column (C_{18}, 5 µm pore size,

4.6 mm $\times$ 250 mm, XBridge, Waters). The fractionated peptides were dried and stored at -80°C

until LC-MS/MS analysis.

LC-MS/MS and database search

Each fractionated peptide sample was analyzed using an Orbitrap Fusion™ Lumos™ Tribrid™

Mass Spectrometer coupled with the Easy-nLC 1200 nano-flow liquid chromatography system

(Thermo Fisher Scientific). The dried peptides were reconstituted using 0.1% formic acid and

loaded on a C18 trap column. Peptides were resolved using a linear gradient solvent B (0.1%

formic acid in 95% acetonitrile) and analyzed by high resolution mass spectrometry in the data-

dependent acquisition mode. MS1 and MS2 were acquired for the precursor and the peptide

fragmentation ions, respectively. MS1 scans were measured at a resolution of 120,000 and

an *m/z* of 200. MS2 scans were acquired following the fragmentation of precursor ions by high-

energy collisional dissociation (HCD) and were detected at a mass resolution of 50,000 and

an *m/z* of 200. Dynamic exclusion was used to reduce redundant fragmentation of the same

ions. The obtained mass spectrometry data were analyzed using the MaxQuant software

(Tyanova, Temu, & Cox, 2016). Raw MS data were searched against the Xoo proteome in

UniProt database. Carbamidomethylation of cysteine and 10-plex TMT modification of lysine

and N-terminals were set as static modifications, whereas oxidation of methionine was set as a

variable modification. False discovery rates at the levels of protein and peptide-spectrum

matches were set at 0.01. The raw MS data and MaxQuant search results have been submitted

to ProteomeXchange (project accession: PXD020135, reviewer access

with username: reviewer34070@ebi.ac.uk and password: mzn76I1O).

Proteomic data analysis

The contaminant and reverse identified proteins were removed from the MaxQuant data.

Proteins identified in both replicates were pooled for quantile normalization. The normalized

values for the replicates were subjected to supervised hierarchical clustering and principal

component analysis, using Perseus (Tyanova, Temu, Sinitcyn, et al., 2016), and the results were

depicted in the form of a multi-scatter plot.

STRING map analysis

The list of proteins in the selected patterns from the hierarchical clusters was uploaded on the

STRING database (https://string-db.org/) to analyze the protein interaction maps. The clusters

of proteins in the blue and red classes, which show the similar time-resolved expression

patterns, were analyzed, and the list including the gene names with the selected organism was

inputted in the multiple proteins search setting. The number of nodes and edges were

automatically calculated based on Xoo genes with a PPI enrichment p-value of 20.9E-10.

Figures were downloaded in the PNG file format for visualization.

RNA-Seq and data analysis

In addition to the previously obtained RNA-Seq data for P-activated Xoo cells, RNA-Seq data

at 90 and 120 min were obtained to correspond with the proteome data at these time points.

RNA-Seq and data analysis were performed as previously described (S. Kim et al., 2016).

Briefly, total RNA from samples was used to generate sequencing libraries, from which

ribosomal RNA was removed using MICROBExpress Bacterial mRNA Enrichment Kit

(Ambion, Austin, TX, USA), and enriched mRNA was prepared using Illumina TruSeq RNA

Sample Preparation Kit (Illumina, San Diego, CA, USA). The RNA obtained after

fragmentation was used to generate cDNA fragments, which were sequenced using Illumina

Genome Analyzer IIx and mapped to the reference genome sequence

(http://www.ncbi.nlm.nih.gov/nuccore/58579623?report=fasta) using CLC Genomics

Workbench 4.0 (CLC bio, Aarhus, Denmark). Relative transcript abundance was calculated

based on the number of reads per kilobase per million mapped sequence reads (RPKM).

Analysis of time-resolved continuous mRNA and protein expression

The RPKM values of the mRNAs in the transcriptome and TMT intensities of the proteins

in the proteome corresponded to the observed expression level of each gene at a given time

point. To facilitate the comparison of gene expression levels, the observed expression level at

each time point was converted to fold change in gene expression, by dividing the expression

level at a given time point by the initial expression level (0 min) of the same gene. The fold

change in the time-resolved expression levels of a given gene during the two hours following

the RLX treatment was fitted to a curve and analyzed using non-linear regression by GraphPad

Prism (version 3.02 for Windows, GraphPad Software, San Diego California USA,

www.graphpad.com), and the continuous time-dependent changes in the mRNA and protein

expression levels were determined using the fitted curve.

Analysis of the correspondence between mRNA and protein expression levels

For the accurate comparison of the mRNA and protein expression levels in P-activated Xoo

cells, the expression levels were corrected by comparing with those in the control cells at each

time point. The fold change in mRNA and protein expression in P-activated Xoo cells at a given

time point was divided by that of the control cells at the same time point. For analyzing the

correspondence between the mRNA and protein levels of each gene in the transcriptome and

proteome, the resulting control-corrected fold change values of mRNAs were compared with

those of the proteins for a given time point as the gene expression level of mRNAs and proteins.

Declarations

Competing interests

Seunghwan Kim, Wooyoung Eric Jang, Min-Sik Kim, Jeong-Gu Kim, and Lin-Woo Kang declare that they have no conflict of interest. The authors declare no competing financial interests. This article does not contain any studies with human or animal subjects performed by the any of the authors.

Available data and material

The raw MS data and MaxQuant search results are available at ProteomeXchange (Project accession: PXD020135, Reviewer access with username: reviewer34070@ebi.ac.uk and password: mzn76I1O)

Funding

This work was undertaken in association with the Cooperative Research Program for Agriculture Science & Technology Development (Project No. PJ01327002020), Rural Development Administration, Republic of Korea.

Author Contributions

Investigation, Seunghwan Kim, Wooyoung Eric Jang, Min-Sik Kim, Jeong-Gu Kim, and Lin-Woo Kang; writing, Seunghwan Kim, Wooyoung Eric Jang, Min-Sik Kim, Jeong-Gu Kim, and Lin-Woo Kang; methodology, Seunghwan Kim, Wooyoung Eric Jang, Min-Sik Kim, Jeong-Gu Kim, and Lin-Woo Kang; funding acquisition, Jeong-Gu Kim, and Lin-Woo Kang; supervision, Min-Sik Kim, Jeong-Gu Kim, and Lin-Woo Kang. All authors have read and

agreed to the published version of the manuscript.

Acknowledgments

We are grateful to the staff members of the Cooperative Research Program for Agriculture

Science & Technology Development (Project No. PJ01327002020), Rural Development

Administration,

Abbreviations

Xoo, *Xanthomonas oryzae* pv. *oryzae*; RLX, rice leaf homogenate; KACC, Korean

Agricultural Collection Center; P-activated, Pathogenicity-activated; LC-MS/MS, liquid

chromatography-tandem mass spectrometry; COG, Clusters of Orthologous Groups; RPKM,

Reads per kilobase per million mapped reads

References

Garau, G., Garcia-Saez, I., Bebrone, C., Anne, C., Mercuri, P., Galleni, M., . . . Dideberg, O.
(2004). Update of the standard numbering scheme for class B beta-lactamases.
Antimicrob Agents Chemother, 48(7), 2347-2349. doi: 10.1128/AAC.48.7.2347-
2349.2004

Hershey, J. W. B., Sonenberg, N., & Mathews, M. B. (2019). Principles of Translational Control.
Cold Spring Harb Perspect Biol, 11(9). doi: 10.1101/cshperspect.a032607

Jackson, S. A. (2016). Rice: The First Crop Genome. *Rice (N Y), 9*(1), 14. doi: 10.1186/s12284-
016-0087-4

Jones, J. D., & Dangl, J. L. (2006). The plant immune system. *Nature, 444*(7117), 323-329. doi:
10.1038/nature05286

Kay, S., Hahn, S., Marois, E., Wieduwild, R., & Bonas, U. (2009). Detailed analysis of the
DNA recognition motifs of the Xanthomonas type III effectors AvrBs3 and
AvrBs3Deltarep16. *Plant J, 59*(6), 859-871. doi: 10.1111/j.1365-313X.2009.03922.x

Kim, S., Cho, Y. J., Song, E. S., Lee, S. H., Kim, J. G., & Kang, L. W. (2016). Time-resolved
pathogenic gene expression analysis of the plant pathogen Xanthomonas oryzae pv.
oryzae. *BMC Genomics, 17*, 345. doi: 10.1186/s12864-016-2657-7

Kim, S. H., Lee, S. E., Hong, M. K., Song, N. H., Yoon, B., Viet, P., . . . Kang, L. W. (2011).
Homologous expression and quantitative analysis of T3SS-dependent secretion of TAP-
tagged XoAvrBs2 in Xanthomonas oryzae pv. oryzae induced by rice leaf extract. *J
Microbiol Biotechnol, 21*(7), 679-685.

Kim, S., Nguyen, T. D., Lee, J., Hong, M. K., Pham, T. V., Ahn, Y. J., . . . Kang, L. W. (2013).
Homologous expression and T3SS-dependent secretion of TAP-tagged Xo2276 in
Xanthomonas oryzae pv. oryzae induced by rice leaf extract and its direct in vitro
recognition of putative target DNA sequence. *J Microbiol Biotechnol, 23*(1), 22-28.

Lee, B. M., Park, Y. J., Park, D. S., Kang, H. W., Kim, J. G., Song, E. S., . . . Go, S. J. (2005).
The genome sequence of Xanthomonas oryzae pathovar oryzae KACC10331, the
bacterial blight pathogen of rice. *Nucleic Acids Res, 33*(2), 577-586. doi:
10.1093/nar/gki206

Li, J. Y., Wang, J., & Zeigler, R. S. (2014). The 3,000 rice genomes project: new opportunities
and challenges for future rice research. *Gigascience, 3*, 8. doi: 10.1186/2047-217X-3-8

Liu, Y., Kong, D., Wu, H. L., & Ling, H. Q. (2020). Iron in Plant-pathogen Interactions. *J Exp*
Bot. doi: 10.1093/jxb/eraa516

Mew, T. W., Alvarez, A. M., Leach, J. E., & Swings, J. (1993). Focus on Bacterial-Blight of
Rice. *Plant Disease, 77*(1), 5-12. doi: Doi 10.1094/Pd-77-0005

Mukherjee, S., & Kearns, D. B. (2014). The structure and regulation of flagella in Bacillus
subtilis. *Annu Rev Genet, 48*, 319-340. doi: 10.1146/annurev-genet-120213-092406

Oliva, R., Ji, C., Atienza-Grande, G., Huguet-Tapia, J. C., Perez-Quintero, A., Li, T., . . . Yang,
B. (2019). Broad-spectrum resistance to bacterial blight in rice using genome editing.
Nat Biotechnol, 37(11), 1344-1350. doi: 10.1038/s41587-019-0267-z

Quibod, I. L., Atieza-Grande, G., Oreiro, E. G., Palmos, D., Nguyen, M. H., Coronejo, S. T., . . .
Oliva, R. (2020). The Green Revolution shaped the population structure of the rice
pathogen Xanthomonas oryzae pv. oryzae. *ISME J, 14*(2), 492-505. doi:
10.1038/s41396-019-0545-2

Schubert, O. T., Rost, H. L., Collins, B. C., Rosenberger, G., & Aebersold, R. (2017).
Quantitative proteomics: challenges and opportunities in basic and applied research.
Nat Protoc, 12(7), 1289-1294. doi: 10.1038/nprot.2017.040

Tsuge, S., Furutani, A., & Ikawa, Y. (2014). Regulatory network of hrp gene expression in
Xanthomonas oryzae pv. oryzae. *Journal of General Plant Pathology, 80*(4), 303-313.
doi: 10.1007/s10327-014-0525-3

Tyanova, S., Temu, T., & Cox, J. (2016). The MaxQuant computational platform for mass
spectrometry-based shotgun proteomics. *Nat Protoc, 11*(12), 2301-2319. doi:
10.1038/nprot.2016.136

Tyanova, S., Temu, T., Sinitcyn, P., Carlson, A., Hein, M. Y., Geiger, T., . . . Cox, J. (2016). The
Perseus computational platform for comprehensive analysis of (prote)omics data. *Nat*
Methods, 13(9), 731-740. doi: 10.1038/nmeth.3901

Ullah, J. H., Walsh, T. R., Taylor, I. A., Emery, D. C., Verma, C. S., Gamblin, S. J., & Spencer,
J. (1998). The crystal structure of the L1 metallo-beta-lactamase from
Stenotrophomonas maltophilia at 1.7 A resolution. *J Mol Biol, 284*(1), 125-136. doi:
10.1006/jmbi.1998.2148

Scheme legend

Scheme 1. Schematic representation of the in vitro assay system and combined analysis of

the time-resolved transcriptome and proteome using RNA-Seq and LC-MS/MS

  Figure legends

  Figure 1. Time-supervised hierarchical clustering of duplicate pathogenicity-activated

  proteome datasets. The heatmap (left) of genes clustered based on the similarities in the

  protein expression pattern over time is presented for duplicate samples of pathogenicity-

  activated and control Xoo cells. Low to high expression is indicated by a change in color from

  green to red. Selected cluster profile patterns (right) are presented for the downregulated (blue

  box) and upregulated (red box) genes in the P-activated Xoo cells.

  Figure 2. Time-resolved protein expression patterns associated with the COG categories.

  Two functional categories of genes (COG) associated with the greatest changes in the protein

  level are indicated by red boxes, nine functional COGs associated with moderate changes are

  indicated by yellow boxes, and other nine functional COGs associated with minor changes in

  the protein level are indicated by blue boxes. The inset for each functional COG indicates the

  control (untreated cells). The Y-axis represents the fold change in the protein expression level

  in comparison with that at 0 min, and the maximum value on the Y-axis was set as 3. The X-

  axis indicates the time from 0 to 120 min. Each line indicates the protein expression level of a

  gene. The protein expression level is represented as grey (from 0 to 30 min) and black (from

  30 to 120 min) lines. Note that in contrast to the transcript data, the proteome data was not

  available between 0 and 30 min (grey part).

  Figure 3. Time-resolved mRNA and protein levels of genes associated with cell motility

  and inorganic ion transport. (A) Time-resolved expression of cell motility-related genes. The

  mRNA expression levels from RNA-Seq are indicated by dashed red lines. The protein

expression levels from LC-MS/MS are indicated by black dashed (from 0 to 30 min) and black

solid (from 30 to 120 min) lines. (B) Time-resolved expression of inorganic ion transport-

related genes. The mRNA expression levels from RNA-Seq and protein expression levels from

LC-MS/MS are indicated as in (A). The Y-axis represents the fold change.

Figure 4. Gene clusters and time-resolved expression patterns of cell motility-related

genes. (A) Gene clusters of flagellar biosynthesis-related genes (groups I-A and I-B) and

chemotaxis genes (group II). (B) Time-resolved mRNA and protein expression levels of genes

in groups I-A, I-B, and II. The down- and upregulation peaks are shown in green and red,

respectively. The yellow and grey ovals indicate unaltered and undetected expression levels.

The thin-bordered oval (at 5, 10, 15, and 45 min) for the proteome data is to ensure consistency

with the pictorial format of RNA-Seq data. The dotted arrows and ovals indicate genes that are

not directly related to flagellar biosynthesis and chemotaxis. Time is expressed in min.

Figure 5. Time-resolved mRNA and protein levels of iron transport-related genes. (A)

Time-resolved mRNA (red) and protein (black) expression levels of *FecA* and *CirA* genes. (B)

Time-resolved mRNA (red) and protein (black) expression level of *IroN* genes. The Y-axis

represents $\log_2$(fold-change).

Figure 6. Time-resolved mRNA and protein expression and secretion of effector proteins.

The mRNA (red) and protein (black) expression levels of *XoAvrBs2* (A) and *XoAvrBs3* (B)

genes in the transcriptome and proteome data (above) and the cytosolic and secreted XoAvrBs2

(A) and XoAvrBs3 (B) proteins in the dot blot data (below). NC, negative control; PC, positive

control

Figure 7. Schematic representation of genome-wide pathogenic gene expression and effector secretion via in vitro assay system. Early response Xoo genes from the initial interaction with RLX include genes related to cell motility, iron uptake, and effector, of which expression are upregulated as early as 5 min in mRNAs and 30 min in proteins and effector secretion is found since 15 min.

Supplementary data

Supplementary figure legends

Figure S1. Detailed analysis of the proteome data of pathogenicity-activated Xoo cells (A) Number of proteins identified from the high-resolution mass spectrometry-based quantitative proteomic analysis and (B) coverage percentage of the protein sequence of the identified proteins (median sequence coverage was ~24%). (C) Number of identified peptides and peptide-spectrum matches. (D) Quantile normalization box plot. (E) Multi-scatter plot for analyzing the correlations between samples.

Figure S2. Comparative analysis of the proteome of the samples (A) Quantile normalization box plot. The quantitative proteomic values were normalized using quantile normalization. (B) Multi-scatter plot for analyzing the correlation between samples. Pairwise comparisons of protein expression levels in all samples are presented as a multi-scatter plot. Pearson's correlation coefficients of 0.98-0.99 were obtained. (C) Principal component analysis of Xoo proteins revealed close relationships among the proteomes of all controls and changes in proteome concentrations upon RLX treatment.

Figure S3. Analysis and comparison of gene expression patterns in the datasets. The STRING maps (Benjamin-Hochberg at FDR 0.05) of selected cluster profile patterns with (A) downregulated (blue box) and (B) upregulated (red box) genes are shown.

Figure S4. Comparison of time-resolved mRNA and protein levels of cell motility-related

genes. (A) Time-resolved mRNA levels of flagellar biosynthesis and chemotaxis-related genes

for pathogenicity-activated Xoo cells. All genes in groups I-A, I-B, and II exhibited the lowest

expression level at 5 min and the highest expression level at 30 min. (B) Time-resolved protein

levels of flagellar biosynthesis and chemotaxis-related genes in pathogenicity-activated Xoo

cells. Genes in groups I-A and I-B exhibited the highest expression level at 30 min, whereas

those in group II exhibited the highest expression at 90 min.

Figure S5. Time-resolved mRNA and protein expression of iron uptake-related genes.

Time-resolved mRNA (red) and protein (black) expression levels of iron uptake-related genes

(A) *FecA* and (B) *CirA*. The Y-axis represents $\log_2$(fold change).

Figure S6. Gene cluster and time-resolved mRNA and protein expression levels of

phosphate uptake-related genes. (A) Gene cluster of the phosphate uptake regulation genes

(*OprO-PhoX-PstSCAB-PhoU*). (B) Time-resolved mRNA and protein expression levels of

phosphate uptake-related genes *OprO*, *PhoX*, *PstSCAB*, and *PhoU.* The line colors correspond

to those of the arrows in (A).

Figure S7. Gene clusters of cell motility-related genes in *E. coli* and Xoo and time-resolved

mRNA and protein expressions of Xoo genes. Gene cluster of cell motility-related genes of

(A) *E. coli* and (B) Xoo. (C) Time-resolved mRNA and protein expression of cell motility-

related genes in Xoo. The cell motility-related genes in *E. coli* are classified as class I (red), II

(yellow), III (blue), and II+III (green). The ortholog genes in Xoo are indicated using the same

color. For the proteome data, * indicates the expression peak at 30 min; **, between 30 and 90

min; ***, at 90 min. The Y-axis represents $\log_2$(fold change).

Figure S8. Gene clusters and time-resolved mRNA and protein expressions of iron

uptake-related genes in Xoo. (A) Gene cluster of iron uptake-related genes, labelled in blue.

(B) Time-resolved mRNA and protein expression levels of iron uptake genes. Low to high

expression is indicated by a change in color from blue to red. Black cells indicate undetected

expression.

Figure S9. Gene clusters and time-resolved mRNA and protein expressions of genes

coding for effector molecules. (A) Gene cluster of effector genes, indicated in red. (B) Time-

resolved mRNA and protein expressions of effector genes. Genes with available time-resolved

proteomic data are labelled with *. Of the *avrXa7* and *avrXa3* genes, the mRNA levels of only

the former were measured, indicated by +. Transposase genes are indicated in blue.

Supplementary Table legends

Table S1. The raw data of time-resolved expression levels of mRNAs and proteins in the transcriptome and proteome of P-activated and control Xoo cells. The RPKM values of mRNAs and TMT intensities of proteins were obtained from independent duplicated transcriptome and proteome analyses, respectively, representing the expression levels of mRNAs and proteins of each gene. The gene products that were not identified or measured are indicated by blank cells.

Table S2. The list of selected genes with similar protein expression patterns in the time-supervised hierarchical clustering. The list of genes marked with blue (downregulated) and red (upregulated) rectangles in Figure 1, which presented distinct synchronized protein expression patterns. Of the 40 flagellum-related KEGG annotated genes, 21 were included in the list, of which 20 genes were upregulated and one gene was downregulated in the proteome.

Table S3. The fold change of time-resolved expression levels of mRNAs and proteins in the transcriptome and proteome of P-activated and control Xoo cells. The fold change (compared to 0 time) of expression levels is calculated by dividing the expression levels of mRNAs and proteins at a given time point by that at 0 min. Cells are colored based on the expression level; low to high expression is indicated by a change in color from blue to red, with white indicating a fold change of 1.

Table S4. The control-corrected fold change of time-resolved expression levels of mRNAs and proteins in the transcriptome and proteome. For the convenient comparison of the mRNA and protein expression levels, the expression levels of mRNAs and proteins in P-activated Xoo cells at each time point was divided by that of the control cells (RLX-untreated Xoo cells) at the same time point. The number and percentage of genes exhibiting more than 20%, 50%, and 200% (2-fold) up- and downregulation are indicated at the bottom of the table.

Table S5. The number of consistently up- and downregulated genes for 120 min. The numbers of more than 20%, 50%, and 2-fold up- and downregulated genes are measured in Xoo cells for the initial 20 min since the first interactions with rice.

Table S6. The list of genes having 2-fold upregulated proteins and mRNAs at 30 min. The lists contain genes having more than 2-fold upregulated proteins at 30 min in both duplicate proteome datasets (S6-1), genes having more than 2-fold upregulated proteins at 30 min in the proteome repetition dataset 1 (S6-2) and 2 (S6-3), and genes having more than 2-fold upregulated mRNAs at 30 min in the transcriptome dataset (S6-4).

Table S7. The list of genes having 50% downregulated proteins at 30 min in the duplicate proteome datasets.   The list contains genes having more than 50% downregulated proteins at 30 min in both proteome datasets.

Scheme 1

Figure 1

Figure 2

Figure 3

A

COG of cell motility (N)

Fold change

Time (min)

B

COG of inorganic ion transport (P)

Fold change

Time (min)

Figure 4

Figure 5

Figure 6

Figure 7