Architecture of genome-wide transcriptional regulatory network reveals

dynamic functions and evolutionary trajectories in *Pseudomonas syringae*

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Abstract

The model Gram-negative plant pathogen *Pseudomonas syringae* utilises hundreds of transcription factors (TFs) to regulate its functional processes, including virulence and metabolic pathways that control its ability to infect host plants. Although the molecular mechanisms of regulators have been studied for decades, a comprehensive understanding of genome-wide TFs in *Psph* 1448A remains limited. Here, we investigated the binding characteristics of 170 of 301 annotated TFs through ChIP-seq. Fifty-four TFs, 62 TFs and 147 TFs were identified in top-level, middle-level and bottom-level, reflecting multiple higher-order network structures and direction of information-flow. More than forty thousand TF-pairs were classified into 13 three-node submodules which revealed the regulatory diversity of TFs in *Psph* 1448A regulatory network. We found that bottom-level TFs performed high co-associated scores to their target genes. Functional categories of TFs at three levels encompassed various regulatory pathways. Three and 25 master TFs were identified to involve in virulence and metabolic regulation, respectively. Evolutionary analysis and topological modularity network revealed functional variability and various conservation of TFs in *P. syringae* (*Psph* 1448A, *Pst* DC3000, *Pss* B728a and *Psa* C48). Overall, our findings demonstrated the global transcriptional regulatory network of genome-wide TFs in *Psph* 1448A. This knowledge can advance the development of effective treatment and prevention strategies for related infectious diseases.

Keywords: transcription factor; *Pseudomonas syringae*; regulatory network

Introduction

Transcription is a pivotal process in cellular life events. Transcription factors (TFs)

play a crucial role in this process by acting as key regulators that coordinate various

biological activities(Lee & Young, 2013; Papavassiliou & Papavassiliou, 2016). TFs

control the recruitment of RNA polymerase by identifying and binding to the

promoters of downstream genes, thereby either activating or repressing the expression

of target genes(Lambert et al., 2018; Wade, 2015). Numerous studies have focused on

regulatory events in eukaryotic species, such as humans(Jolma et al., 2013),

mice(Badis et al., 2009), and *Saccharomyces cerevisiae*(C. Zhu et al., 2009), and

prokaryotic species, such as *Escherichia coli*(Shen-Orr, Milo, Mangan, & Alon, 2002).

However, limited comprehensive TF binding datasets for microbial pathogens are

available.

Pseudomonas syringae, an important Gram-negative phytopathogen and a model

pathogenic bacterium, infects many plants, including economically valuable crops,

resulting in substantial annual economic losses globally(Hirano & Upper, 2000).

Upon entering host cells, *P. syringae* employs several strategies, such as changing its

motility type and secreting phytotoxins, to overcome the plant's immune defences and

establish colonies(Bender, Alarcón-Chaidez, & Gross, 1999; Ichinose, Taguchi, &

Mukaihara, 2013; Taguchi & Ichinose, 2011). *P. syringae* causes severe disease by

secreting various effector proteins through the needle-like type III secretion system

(T3SS); this process is regulated by a cluster of TFs (Cunnac, Lindeberg, & Collmer,

2009; Hendrickson, Guevera, & Ausubel, 2000; Huang, Yao, Sun, Ji, & Deng, 2022;

Jingru Wang et al., 2018). The alternative sigma factor RpoN activates the

transcription of another alternative sigma factor, HrpL, which, in turn, binds to the

pathogenicity (*hrp*) box in the promoter region of T3SS genes, regulating most of

these T3SS genes (Alfano & Collmer, 1997; Lan, Deng, Zhou, & Tang, 2006;

Yingxian Xiao & Hutcheson, 1994). HrpS is one of the most important TFs that

regulate numerous biological processes (Jingru Wang et al., 2018). Its heterodimeric

complex, HrpRS, is modulated by at least six two-component systems (TCSs): RhpRS

(Deng et al., 2014), CvsRS (Fishman, Zhang, Bronstein, Stodghill, & Filiatrault,

2018), GacAS (Chatterjee et al., 2003), AauRS (Yan, Rogan, Pang, Davis, &

Anderson, 2020), CbrAB2 and EnvZ-OmpR (Shao et al., 2021). In particular, RhpRS

serves as a master regulator of T3SS in *Psph* 1448A. RhpRS senses plant-derived

signals, such as polyphenols, through the histidine kinase Pro40 within RhpS and

controls the expression of T3SS genes in response to environmental stress(Deng et al.,

2014; Yanmei Xiao et al., 2007; Xie et al., 2021). Within the sensor region, the

cognate response regulator RhpR undergoes modulation in its phosphorylation state

by RhpS, thereby regulating a group of T3SS genes(Deng et al., 2010).

Phosphorylated RhpR directly binds to the *hrpRS* promoter, suppressing the *hrpRS*

operon and the subsequent *hrpRS-hrpL-hrp* cascade (Deng et al., 2010; Deng et al.,

2014; Deng, Xiao, Lan, Zhou, & Tang, 2009; Shao, Xie, Zhang, & Deng, 2019;

Yanmei Xiao et al., 2007).

Recently, through a combined analysis of RNA sequencing (RNA-seq) and chromatin

immunoprecipitation sequencing (ChIP-seq), we identified seven additional TCSs

(ErcS, Dcsbis, PhoBR, CzcSR, AlgB/KinB, MerS and CopRS) that regulate the

virulence of *Psph* 1448A (Xie et al., 2022). In addition, we developed an intricate

PSTCSome (*Psph* 1448A TCS regulome) network containing numerous functional

genes that respond to changing environmental conditions(Xie et al., 2022).

Furthermore, we examined the overall crosstalk between 16 virulence-related

regulators under different growth conditions, such as King's B and minimal media. By

analysing differentially expressed genes and binding peaks, we constructed a *Psph*

1448A regulatory network (PSRnet), revealing the involvement of hundreds of

functional genes in virulence pathways(Shao et al., 2021). We also elucidated the

molecular mechanisms and functions of TFs binding within coding sequences (CDS)

and found that CDS-binding TFs interact with cryptic promoters in coding regions,

thereby regulating the expression of subgenus and antisense RNAs(Hua et al., 2022).

We propose a luminescence reporter system designed to quantitatively measure the

translational elongation rates (ERs) of T3SS-related proteins. Our findings

demonstrate the key roles of transfer RNAs (tRNAs) and elongation factors in

modulating translational Ers and facilitating T3SS protein synthesis(Sun et al., 2022).

Although many key virulence regulators in *Psph* 1448A have been studied, the global

regulatory mechanism and interactions of all 301 annotated TFs across various

biological processes remain unclear. To comprehensively explore the DNA-binding

features and map the transcriptional regulatory network of all TFs in *Psph* 1448A, we

constructed 170 TF-overexpressing strains and used ChIP-seq, a highly effective and

important technology for analysing protein–DNA interactions(Mathelier, Shi, &

Wasserman, 2015). This analysis not only provided insights into the interactions

between TFs and their target genes but also revealed the hierarchy (top, middle and

bottom) and co-association scores of all these TFs. We found that more than half of

270 TFs (100 TFs from HT-SELEX and 170 TFs this study) in downstream position

tended to be regulated by top TFs and bound to the target genes with high

co-associated scores. Different TF-pairs were classified into 13 basic three-node

submodules, including ringent loops and locked loops. In addition, we mapped the

hierarchical binding network of TFs and identified three virulence-related master TFs

and 23 metabolic master TFs. Furthermore, we employed ChIP-seq to determine the

binding sites of 5 TFs in 4 *P. syringae* lineages (*Psph* 1448A, *Pst* DC3000, *Pss* B728a

and *Psa* C48), revealing the diversity of TF binding events and the varying functions

of TFs among different *P. syringae* strains. Topological modularity classification of

the network, including TFs and target genes, revealed the diverse biological functions

of TFs in *Psph* 1448A. This study provides a global and convenient platform for

understanding the transcriptional regulatory characteristics and biological functions of

TFs in *P. syringae*. In addition, this study provides valuable insights that can inform

the development of effective therapies for not only *P. syringae* but also other

associated infectious diseases.

Results

ChIP-seq analysis revealed the binding specificities of 170 previously uncharacterised TFs in *Psph* 1448A

Based on the current annotations available on '*Pseudomonas* Genome DB' (https://www.pseudomonas.com/)(Winsor et al., 2016), we initially determined the locations of all 301 annotated TFs in the *Psph* 1448A genome **(Figure S1a)**. To elucidate the binding preferences and functional characteristics of TFs of *Psph* 1448A, we performed ChIP-seq for the 170 TFs, including three (1.8%) predicted transcriptional regulators, 132 (77.6%) annotated transcriptional regulators and 35 (20.6%) functional proteins with DNA-binding annotations. Based on the DNA-binding domains as annotated in the transcription factor prediction database(Wilson, Charoensawan, Kummerfeld, & Teichmann, 2008), we categorised the 170 analysed TFs into 25 families(Fan et al., 2020). The majority of TFs belonged to the LysR, TetR, AsnC, GntR and AraC families. Among these TFs, PSPPH4700, PSPPH3798, CysB, PSPPH1951, PSPPH4638, PSPPH3504, PSPPH3268 and Irp exhibited over 1,000 binding peaks **(Figure S1b)**. The enriched loci of these binding peaks indicated that these TFs displayed a significant preference for binding to promoters, directly regulating the transcription of downstream targets **(Figure S1c)**. The peak loci of 10 TFs (PSPPH0286, PSPPH0411, PSPPH0711, PSPPH1734, PSPPH2357, PSPPH2407, PSPPH2862, PSPPH3155, PSPPH3431, PSPPH3468, PSPPH4127, PSPPH4622 and PSPPH4768) were completely enriched in the promoter region, with the majority of them belonging to the LysR family. Taken

144 together, the 170 tested TFs in *Psph* 1448A had over 26,000 DNA-binding peaks

145 distributed across different regions of target genes, suggesting their direct regulatory

146 functions.

Hierarchical TFs reflected multiple higher-order network structures

Transcriptional changes in bacteria are often manipulated by a complex network of

TFs. However, bacterial TFs usually have been studied individually or in small

clusters with related functions. To comprehensively investigate the associations of all

TFs in *Psph* 1448A at a system level, we constructed a hierarchical network of 270

analysed TFs (100 TFs from HT-SELEX and 170 TFs this study). The findings

revealed 1,757 TF interactions among these 26,000 binding events (**Supplementary**

Table 1a). Subsequently, we computed information flow parameters for each

TF(Gerstein et al., 2012). In brief, we defined out-degree (O) and in-degree (I) as the

number of interactions of a TF in the hierarchical network, representing the regulation

of other factors by this TF and the regulation of this TF by other factors, respectively.

The difference between O and I indicated the direction of information flow in the

network. Hierarchy height (H) was defined as the normalised metric of information

circulation, calculated as $H = (O - I)/(O + I)$. When H was close to 1 ($H \approx 1$), it

indicated that these TFs tended to regulate other factors and occupy upstream

positions in the network. Conversely, when H was close to −1 ($H \approx -1$), it indicated

that these TFs were more likely to be regulated than to regulate other TFs, occupying

downstream positions in the network. Based on these criteria, we categorised the 270

analysed TFs (100 TFs from HT-SELEX and 170 TFs this study) into three levels: 54

(20%) executive TFs (such as AlgQ, LexA2 and PSPPH0222) at the top level, 62

(23%) communicative TFs (such as MexT, PsrA and PSPPH1100) at the middle level

and 147 (54%) foreman TFs (such as PobR, DksA2 and PSPPH0755) at the bottom

level **(Figure 1a, Supplementary Table 1b)**. The presence of a larger number of TFs

(147) at the bottom level indicated a high degree of information flow, suggesting the

maximisation of the number of downward-pointing edges in the network.

Different numbers of TFs in three levels in the hierarchical network revealed that

more than half of TFs (bottom-levels TFs) tend to be regulated by other TFs and then

directly bound to target genes. This tendency indicated a downward information flow

of transcription regulation of *P. syringae*. Therefore, we defined the direct binding

between two TFs as a direct interaction and investigated collaborations within and

between hierarchy levels, specifically intra-level (top to top 'TT', middle to middle

'MM' and bottom to bottom 'BB') and inter-level (top to middle 'TM', top to bottom

'TB' and middle to bottom 'MB') interactions, indirect interaction if two TFs

co-associated with one target DNA and cooperative TF pair was defined if their

common target was from a TF **(Figure 1b, Supplementary Table 1a)**. In terms of the

top-level TFs, direct interactions became more enriched as the hierarchy level of their

collaborators decreased. Direct interactions between TB pairs constituted the most

substantial portion, accounting for nearly half of all interactions. A similar pattern was

observed among the bottom-level TFs, where interactions diminished as the hierarchy

level of their collaborators decreased. Compared with interactions among the top- and

bottom-level TFs, middle-level TFs, serving as information transmission centres,

exhibited lower levels of intra-level collaborations. In summary, transcriptional

regulation within the TF hierarchy was predominantly manipulated by top-level TFs,

which directed the flow of information to downstream TFs.

Multiple three-node submodules revealed the regulatory diversity of TFs in the

***Psph* 1448A regulatory network**

Natural networks, including transcriptional regulation networks, usually show

complex characteristics(Newman, 2001; Strogatz, 2001). Among complex networks,

some small-scale networks demonstrate numerous connections between individual

information nodes and information clusters(Amaral, Scala, Barthelemy, & Stanley,

2000; Jeong, Tombor, Albert, Oltvai, & Barabási, 2000). To investigate the basic

structural features of our transcription network, we defined directed edges as direct

interactions between two TF nodes and identified global submodules comprising

different TF nodes. In this study, we specifically focused on three-node modules,

which were considered as 'network motifs'(Milo et al., 2002). Using algorithms

designed to detect recurring modules(Shen-Orr et al., 2002), we scanned our hierarchy

network and identified 40,307 different pairs across 13 basic three-node submodules

(Figure 1c).

In the first six submodules, we observed that two TF nodes established a relationship

only through another node. We denoted these submodules (M1 to M6) as 'ringent

loops'. These seemingly simpler regulation modules appeared more in the *Psph*

1448A transcriptional regulatory network, especially the first module (M1, n =

24,479), indicating that *Psph* 1448A favours the use of simple but efficient modes for

modulating transcriptional regulation. For example, PhnF and FruR were directly

regulated by CysB (M1), and CapA was coregulated by Anr and PSPPH4700

simultaneously (M3, n = 6,177). In addition, M6 (n = 20) contained pairs of mutually

regulating TFs (toggle switches), such as PSPPH3268, PSPPH2315 and PSPPH2733.

The remaining seven submodules, denoted as 'locked loops' (M7 to M13), comprised

subordinate three-node regulatory modules within our network. Notably, no instances

of a 'fully connected triad' (M12) were observed in our network (n = 0). We found 50

'self-loop' submodules (M7) in our network, including CspD, PSPPH4694 and

PSPPH3798, which engaged in mutual regulatory interactions. Among these

submodules, M9 (n = 10) was the least common and contained six pairs of toggle

switches involving mutually regulating TFs (such as PSPPH1951, PSPPH4612 and

CysB), which were similar to those found in the human transcriptional regulatory

network(Gerstein et al., 2012). Notably, the most enriched locked loop in our network

was M13 (n = 696), denoted as a 'feed-forward loop', which has been extensively

studied in other species such as humans and *E. coli*. In this submodule, upstream TFs

regulated targeted TFs either by binding directly or manipulating other TFs (**Figure**

S2). For example, TF Irp directly controlled PSPPH4899 and also indirectly regulated

it by binding to Lon.

Taken together, the simplest and most effective submodule M1 and the coregulatory

submodule M13 played crucial roles in the transcriptional regulation of TFs in *Psph*

1448A. In addition, we found 92 auto-regulators in our hierarchy network. These

auto-regulators are important and always act as repressors in scenarios of

multi-stability, such as plant intercellular spaces **(Figure 1d)** (Alon, 2007). For

example, LexA and CysB as negative autoregulators were indicated to reduce

cell-to-cell fluctuations in the steady-state level of the transcription factor (Becskei &

Serrano, 2000; Rosenfeld, Elowitz, & Alon, 2002). These regulators are regarded as

bistable switches that further influence the expression of downstream genes(Burda,

Krzywicki, Martin, & Zagorski, 2011). For example, our previous study demonstrated

that Lon is a dual-function regulator involved in the regulation of virulence and

metabolism in *Psph* 1448A (Hua et al., 2020). Lon was identified as an auto-regulator

in this study. Furthermore, DksA2, which is widely regarded as a protective protein

against oxidative stress(Fortuna et al., 2022), was identified as a new bistable switch

in this study. In summary, the classification of the TF hierarchy and the identification

of enriched network modules not only offer functional predictions for transcriptional

regulators but also provide insights into the communication network that governs TF

regulation in *Psph* 1448A.

High co-association pairs occurred more in bottom-level TFs in *Psph* 1448A

In addition to direct interactions between TFs, we found a notable preference for

co-binding peaks among different TF pairs. Briefly, we counted overlapping regions

within the binding peaks of all TF pairs. The ratio of intersection regions to the union

set of all peaks between two TFs was identified as the genome-wide co-association of

specific TFs. We first analysed co-association scores between TF pairs and grouped

the scores into 3 TF pairs (TT, MM and BB). The majority of TFs tended to cooperate

with other TFs and co-bind to specific genomic regions **(Figures S3a, S4a and S5a)**.

The dark distribution in BB pair indicated that high co-association scores preferred to

occur in bottom-level TFs. To identify the potential functions of TFs in each level, we

used the target genes that were co-bound by TF pairs to perform functional

annotations using hypergeometric tests (BH-adjusted $p < 0.05$) based on gene sets

derived from the Gene Ontology (GO) and Kyoto Encyclopaedia of Genes and

Genomes (KEGG) databases **(Figures S3b, S4b and S5b Supplementary Table 2a,**

c–d). The functional categories of TFs at these three levels encompassed various

regulatory pathways. For example, the top-level TF PSPPH4700 was involved in

siderophore transportation and phosphorelay signal transmembrane transportation.

The middle-level TF Lon regulated GTP binding and ribosome transcription. The

bottom-level TF PSPPH3486 was involved in amino acid transportation, PSPPH0101

in ATP binding and PSPPH1049 in catalytic activity. Notably, we found that TFs at

the top level, without cooperating TFs, exhibited a large number of binding peaks

(Figure S3a). This finding suggested that these TFs preferred to regulate target genes

by recruiting to specific sites of other TFs, facilitating the direct binding between

other TFs and their specific targets, a phenomenon defined as tethered binding (Jie

Wang et al., 2012). For example, the top-level TF PSPPH4700 yielded over 1,700

peaks but cooperated with only 24 top-level TFs with low co-association scores about

0.05 (**Supplementary Table 2b**).

To further elucidate this pattern, we examined the 125 TFs that were analysed through

ChIP-seq and exhibited high co-association scores. We determined the co-association

patterns among these regulators **(Figure 2a)** and classified them into four clusters,

denoted as C1 to C4. Notably, C1 and C4 contained higher proportions of

bottom-level TFs. Based on the analysis in Figure 1b, we found that the proportions of

bottom-level TF interaction in all the TF pair interactions and direct interaction were

43% and 49%. These results indicated that the bottom-level TFs tended to regulate

downstream genes through cooperating with other level TFs. When comparing the

co-association scores of TF pairs, we observed a stronger tendency toward

cooperation among lower-level TFs, especially bottom-level TFs. In particular, 35

bottom-level TFs in C4 (83%) exhibited coregulation at the same peak locations with

high co-association scores. For instance, PhnF and PSPPH4692 co-bound to three

target genes (PSPPH4117, PSPPH4216 and PSPPHB0021), with co-association

scores as high as 0.92. This co-binding relationship between all TF pairs was defined

as an indirect interaction. In contrast to the trend observed for direct interactions and

cooperativity, we observed stronger correlations between top-level TFs and TFs

situated at higher levels (top and middle levels; **Figure 1b)**. For example, the

top-level TFs PSPPH3798 and CysB co-bound to the promoter region of *flhB*, which

encodes a flagellar biosynthesis protein. We found a similar co-association pattern

among bottom-level TFs. However, middle-level TFs displayed the weakest

correlations internally, indicating that regulation by higher-level TFs was distributed

across different regions in targeted promoters. For instance, we observed that

PSPPH3798 bound upstream of *flhB*, whereas the peak of CysB was located in the

overlapStart region.

We found that the correlation between bottom-level TFs was weaker in C3 than in the

other clusters. To further explore the DNA-binding characteristics of all TFs within

the same cluster, we investigated the peak features of the LysR-family TF MexT in C3

as an example. The peak locations in target genes of seven top-level TFs (PSPPH1435,

PSPPH1758, PSPPH2193, PSPPH2454, PSPPH4638, PSPPH4998 and PSPPH3411),

three middle-level TFs (PSPPH1100, PSPPH5132 and PSPPH5144) and four

bottom-level TFs (PSPPH0700, PSPPH2300, PSPPH2444 and PSPPH2580) were

compared with MexT. MexT showed higher co-association scores (more than 60

scores) with more top-level-TFs. The analysis of the peak locations of MexT

demonstrated that MexT showed closer co-association relationships with top-level

TFs **(Figure 2b)**. Within the binding peaks of MexT, the target genes PSPPH3643

(LysR family TF), PSPPH2618 (sulphite reductase (NADPH) haemoprotein, CysI)

and PSPPH2411 (hypothetical protein) displayed high co-association scores with

other TFs. UpSet plots demonstrated that 12 TFs in C3 bound to a highly overlapping

region of CysI **(Figure 2c)**. These results suggest that these genes were prioritised for

co-regulation by MexT. To further explore the binding features in C3, we determined

the motifs of 11 TFs based on their binding sequences obtained via ChIP-seq using

MEME(Bailey et al., 2009), even though the binding motif of MexT was previously

investigated(Tian et al., 2009). We identified a 15-bp consensus motif (ATN11AT)

throughout the 11 analysed TFs **(Figure 2d)**, demonstrating a high degree of

consensus in co-association patterns among TFs in *Psph* 1448A.

Virulence-associated pathways were primarily regulated by top-level TFs in *Psph*

1448A

Because the pathogenicity of *P. syringae* mainly depends on T3SS and other virulence-associated pathways(He, 1998), we particularly focused on TFs that bind to numerous virulence-related genes. Seven pathways manipulate the virulence of *Psph* 1448A (Huang et al., 2022; Xie, Shao, & Deng, 2019), namely T3SS, biofilm production, motility, nucleotide-based secondary messenger function, quorum sensing (QS), phytotoxin production and siderophore production. To comprehensively investigate the transcription regulatory mechanism underlying the virulence of *Psph* 1448A, we calculated the hierarchical heights of TFs involved in virulence regulation and the virulence genes modulated by them. This analysis provided insights into the organisation of the virulence regulatory network in *Psph* 1448A, where virulence-involved TFs were categorised into three tiers **(Figure 3a)**.

We found three transcriptional regulatory channels governing virulence regulation in *Psph* 1448A. The first channel was the direct trigger, which has been extensively studied in previous studies and is referred to as the 'one-step trigger' here(Fan et al., 2020). These TFs are recognised as master regulators that directly respond to biological events without additional intermediaries(Chan & Kyba, 2013). In our previous study, we identified TrpI, RhpR, GacA and PSPPH3618 as master regulators in T3SS and 16 master regulators in other virulence pathways(Fan et al., 2020). In line with this definition, we recognised 35 TFs (PSPPH4644, AlgQ, PSPPH1100, CysB, PSPPH2555, PSPPH0239, PSPPH1762, Irp, PSPPH1951, PSPPH2193, PSPH3504, PSPPH1435, PSPPH2983, PSPPH0700, PSPPH5132, PSPPH1776,

PSPPH4700, SfsA, PSPPH3522, PSPPH3268, PSPPH4356, PSPPH3798,

PSPPH4638, PSPPH4127, Lon, PSPPH4324, PSPPH4673, MexT, PSPPH4844,

PSPPH2214, PSPPH0755, PSPPH4012, PSPPH4920, Anr and PSPPH2476) that

participate in various virulence pathways. More than 68% of these TFs (24 of 35 TFs)

were at the top levels, indicating that the virulence of *Psph* 1448A is primarily

regulated by top-level TFs. Among these TFs in the network, the top-level TFs

PSPPH1951, PSPPH2193 and PSPPH3268 were found to have abundant

virulence-associated target genes based on ChIP-seq results. The de novo motif

analysis of their peak sequences revealed the presence of three head-to-head motifs: a

17-bp motif (AT-N13-AT) for PSPPH1951, a 15-bp motif (ATC-N9-GAT) for

PSPPH2193 and a 10-bp motif (AC-N6-GT) for PSPPH3268 **(Figure 3b–d, Figure**

S6a–c).

Three master TFs were identified to participate in virulence

To further verify the biological functions of these three uncharacterised TFs, we first

purified the TF proteins and performed an electrophoretic mobility shift assay (EMSA)

to confirm their direct interactions with key virulence genes *in vitro*. Next, we

generated TF deletion strains to detect the transcription levels of target genes. We

found that PSPPH1951 directly regulated multiple T3SS genes, including

hrpRhopAE1 and *hopAH2* **(Figure 3b)**. Among them, the expression of *hopAE1* in

ΔPSPPH1951 was significantly increased by more than four-fold compared with the

WT, suggesting that PSPPH1951 acts as a repressor of *Psph* 1448A T3SS **(Figure 3b)**.

In addition, PSPPH1951 was found to bind to the promoters of type IV pili genes

(*pilG*, *pilF* and *pilZ*; **Figure S6d)**.

Pst DC3000 enhances its ability to infect the host by increasing bacterial

motility(Buell et al., 2003). ChIP-seq data (*in vivo*) and EMSA (*in vitro*) results

showed that PSPPH2193 interacted with the promoters of motility-related genes, such

as *fleQ* (encoding a flagellar regulator) and *flhF* (encoding a flagellar biosynthesis

regulator; **Figure 3c)**. RT-Qpcr results indicated that *fleQ* and *flhF* were

downregulated two-fold in ΔPSPPH2193 compared with the WT. As expected,

ΔPSPPH2193 exhibited weaker motility than the WT and complementary strain in

King's B medium **(Figure 3c)**, indicating that PSPPH2193 serves as an activator of

motility in *Psph* 1448A.

PSPPH3268 was found to influence the pathogenicity of *Psph* 1448A by regulating

multiple virulence-related pathways. During the initial stage of *P. syringae* infection,

the bacteria produce biofilm components, including extracellular polysaccharides

(EPSs), type IV pili and other highly viscous compounds. These components help the

bacteria to establish colonies, providing protection against the host's immune

defences and antimicrobial agents(Whitchurch, Tolker-Nielsen, Ragas, & Mattick,

2002). We found that PSPPH3268 strongly interacted with the promoters of key genes

involved in biofilm production such as *hrpR*, the alginate biosynthesis gene *alg44* and

the type IV pilus assembly gene *pilM* **(Figure 3d)**. Furthermore, the transcription

levels of *alg44*, *algX* (encoding the alginate biosynthesis protein) and *pilM* were

markedly enhanced in ΔPSPPH3268. This resulted in enhanced biofilm formation and

EPS production when the PSPPH3268 gene was deleted and then restored when

PSPPH3268 was expressed **(Figure 3d)**. These results demonstrated that PSPPH3268

acts as a master regulator in various virulence-related pathways. In addition, we

identified that the TF PSPPH3798 binds to the promoters of flagellar-related genes

(*fliK*, *fliE*, *fliD* and *fleQ*; **Figure S6e**).

In addition to the 'one-step trigger' mechanism, we found that TFs also regulate

downstream genes through one or two other TFs at different levels, which were

regarded as 'one jump-point trigger' and 'two jump-point trigger'. For example,

PSPPH2555 indirectly influenced biofilm formation (*algD*), motility (*fleQ*), T3SS

(*hrpR*), QS (*ahlR* and *secE*) and phytotoxin production (*aprD*) by directly regulating

the bottom-level TF PSPPH4920 **(Supplementary Table 3)**. PSPPH3504 was found

to   be   involved   in   a

PSPPH3504-Lon/PSPPH4324/PSPPH4844-PSPPH0755/PSPPH4920/PSPPH4012-tar

get gene pathway **(Figure S6f, Supplementary Table 3, '/' represents sibling nodes**

and '-' represents downward regulation). Among these TFs, PSPPH0755,

PSPPH4920 and PSPPH4012 are considered key performer TFs because they mediate

most of the transcription regulatory signals from multiple TFs. We also found reverse

regulatory pathways in our network. For example, the middle-level TF PSPPH4673

was found to directly regulate the top-level TF PSPPH4700 and then indirectly

control the transcription of many virulence-related genes through regulating the

bottom-level TFs PSPPH0755, PSPPH4920, Anr, and PSPPH2476 **(Supplementary**

Table 3). To further investigate the influence of these TFs on the pathogenicity of *P.*

syringae, we performed the plant infection assay. As expected, we found that the

bacterial numbers of Δ2193 strain were significantly reduced compared with WT

strain (**Figure S6g**). In summary, TFs regulate the pathogenicity of *Psph* 1448A

through diverse pathways, either by directly binding to target genes or indirectly

controlling other TFs.

Systematic mapping of TF targets revealed key metabolic regulators in *Psph*

1448A

In addition to enhancing pathogenicity and resisting host defences, *P. syringae* adjusts

its metabolic activities to survive in unpredictable environments(Rico, McCraw, &

Preston, 2011). To comprehensively understand metabolic regulation in *Psph* 1448A,

we constructed a hierarchical network that includes key regulators and the genes they

trigger, similar to the virulence hierarchical network. We focused on eight metabolic

pathways, namely amino acid biosynthesis, DNA replication, ATP-binding cassette

(ABC) transportation, oxidative phosphorylation, tricarboxylic acid (TCA) cycle,

RNA polymerase, phosphonate metabolism and methyl-accepting chemotaxis (**Figure**

4a). Compared with the virulence network shown in Figure 3A, we identified more

TFs involved in metabolic regulation, many of which exhibited numerous interactions

with genes related to oxidative phosphorylation (178 binding peaks) and the TCA

cycle (154 binding peaks; **Figure S7a–c**). In a previous study, 12 master regulators

were reported to control various metabolic pathways, including LexA1, PSPPH3004

and PSPPH1960, involved in reactive oxygen species (ROS) resistance(Fan et al.,

2020). RhpR participates in several metabolic pathways, such as ABC transporters

and oxidoreductase activity(Shao et al., 2019). Lon is involved in glucokinase and

oxidoreductase activity(Hua et al., 2020). MgrA, GacA, PilR, PsrA, RpoN, CvsR,

OmpR and CbrB2 participate in oxidation resistance, amino acid transportation and

other metabolic pathways(Shao et al., 2021). Here, we identified 111 TFs regulating

these eight metabolic pathways. Similar to the aforementioned virulence network,

three transcriptional regulatory channels were observed for the metabolic pathways.

To provide a detailed view of metabolic regulation in *Psph* 1448A, we counted the

number of functionally annotated genes related to each pathway and calculated the

proportion of targets for each TF, highlighting key regulators in these eight metabolic

pathways using radar plots **(Figures 4b and S8a)**.

We found that the TFs CysB and PSPPH3268 regulate all eight metabolic pathways,

whereas the TFs PSPPH1951, PSPPH3798, PSPPH3504 and PSPPH4700 were

predicted to regulate seven metabolic pathways. Notably, the TF PSPPH0755 was

found to bind to the promoters of PSPPH5210 (encoding ATP synthase F0F1 subunit

delta) and PSPPH3109 (encoding the NADH dehydrogenase subunit A NuoA). The

monomer motif (CTGAA) of PSPPH0755 was identified through MEME analysis.

The interactions in these two metabolic pathways were confirmed through EMSA

(Figure S8b). The TF PSPPH3798 was predicted to bind to the promoters of genes in

two pathways, including PSPPH3881 (encoding the methyl-accepting chemotaxis

protein WspA) and PSPPH5119 (encoding the phosphate transport system regulatory

protein PhoU). The 15-bp binding motif of PSPPH3798 was determined to have a

head-to head orientation (ATCG-N7-CGAT). EMSA results confirmed these

interactions **(Figure S8c)**. In addition to the TF PSPPH3798, the TF PSPPH4638 had

a binding site in the promoter region of the PSPPH3881 gene. PSPPH4638 was also

predicted to interact with PSPPH0550 that encodes phosphoserine phosphatase SerB.

The 8-bp monomer motif (ATTTTCAA) of PSPPH4638 was identified, and the

binding interactions were confirmed through EMSA **(Figure S8d)**.

In yeast, TFs in a functional category appear to bind to genes in the same

category(Simon et al., 2001), and we observed a similar pattern in *Psph* 1448A. For

example, TF Irp, a key regulator in the methyl-accepting pathway, bound to the

promoters of PSPPH3798, PSPPH4638 and PSPPH4700, which were also identified

as key regulators in the same pathway **(Figure 4c)**. In addition, we found that TFs

from different categories often bound to the promoters of TFs responsible for other

cellular processes. For example, key regulators controlling oxidative phosphorylation

(highlighted in red and blue; PSPPH1951, PSPPH2193, PSPPH2832, PSPPH3268,

PSPPH3504, PSPPH3798, PSPPH4638, PSPPH4673 and PSPPH4700) bound to TFs

playing key roles in the methyl-accepting pathway (highlighted in brown and blue;

PSPPH1951, PSPPH3268, PSPPH3504, PSPPH3798, PSPPH4638, PSPPH4700 and

Irp; **Figure 4d**). These results demonstrate that many regulatory processes are often

achieved through coregulation by a series of multifunctional TFs throughout the

network, enabling *Psph* 1448A to coordinate transcriptional regulation processes

across multiple cellular processes.

TFs indicated large functional variability across different pathovars in *P.*

syringae

Although TF functions exhibit both inter- and intra-species variability, most previous

studies on TFs have focused on the molecular mechanism of a single strain(Galardini

et al., 2015). To investigate the regulatory mechanism of TFs across different strains

of *P. syringae*, we selected four model strains: *P. syringae* pv. *Syringae* 1448A, *P.*

syringae pv. *Tomato* DC3000, *P. syringae* pv. *Syringae* B728a and *P. syringae* pv.

Actinidiae C48. We used the genome of 1448A as a reference and conducted a

homology analysis of 1448A protein sequences with those of the other three strains

(**Supplementary Table 4**). We determined a high proportion of homologous proteins

in the three strains (4983 in *Pst* DC3000, 4982 in *Pss* B728a and 4984 in *Psa* C48;

Supplementary Table 4). Across the four strains, all 301 annotated TFs were present

(Supplementary Table 5). We selected five TFs (Irp, PSPPH2193, PSPPH3122,

PSPPH4127 and OmpR) to construct TF-overexpressing strains in *Pst* DC3000, *Pss*

B728a and *Psa* C48 before performing ChIP-seq analysis. We identified the binding

sites of all these TFs in the four strains and found divergent binding preferences for

the same TFs in different strains. Most target genes of each TF in one or two strains

were unique. In particular, Irp bound to 19 target genes that were conserved in all four

tested strains, including *purB* (encoding adenylosuccinate lyase), *cceA2* (encoding the

chemotaxis sensor histidine kinase) and *gidA* (encoding the tRNA uridine

5-carboxymethylaminomethyl modification protein). Four highly conserved target

genes were also found to directly interact with PSPPH4127. Evolutionary analysis of

the binding peaks of TFs suggested high binding specificity and varying levels of

conservation of these TFs in the tested strains (**Figure 5a**).

In addition to the intersection between *Psph* 1448A and *Psa* C48, we observed

differences between the target genes of all five TFs (**Figure 5b**) and the peak

locations (TF target interactions; **Figure 5c**) in these four pathovars. The

inconsistency between the number of targets and peaks suggested that some target

genes were regulated by at least one different TF in these four strains, which is similar

to regulation in *P. aeruginosa*(Trouillon et al., 2021). To confirm the presence of

target genes regulated by the same TF or different TFs in various strains, we

compared the peak locations of *gidA* and *rpoD* (RNA polymerase sigma factor) as two

examples. TF Irp was found to bind to the promoter of *gidA* in four strains (**Figure 5d**)

and had 19 conserved target genes in these strains (**Supplementary Table 4**). In

addition, PSPPH4127 had four conserved target genes (*rpoD*, PSPPH_1001,

PSPPH_1998 and PSPPH_5016) in all four strains (**Figure 5e**). These results showed

that Irp and PSPPH4127 exhibited higher functional conservation than the other three

TFs. We also found that PSPPH2193 in 1448A and Irp in *Pst* DC3000 bound to the

promoter of *rpoD* (**Figure 5e**). Differences in the regulation of the same targets by

different TFs were also observed in more than 1,500 target genes, suggesting the

potential diversity of the transcriptional regulation of TFs in our network.

Furthermore, we performed motif and GO analysis to investigate the functional

characteristics of TFs (PSPPH3122 and PSPPH4127) in *P. syringae* strains (Figures

S9-S10). As expected, different motifs and specific functions were found in various

TFs in four *P. syringae* strains. For PSPPH3122, motif (AGACN$_4$GATCAA) and

motif (CGGACGN$_3$GATCA) were found in *Psph* 1448A and *Pst* DC3000 strains,

respectively. Functional enrichment analysis of target genes of PSPPH3122 indicated

its relationship with recombinase activity and DNA recombination in 1448A strain,

while relating with other functions in other strains, such as structural constituent of

ribosome in *Pss* B728a and *Pst* DC3000 strains. We also found four different motifs

of PSPPH4127 in four *P. syringae* strains. Functional enrichment analysis showed its

relationship with recombinase activity in *Psph* 1448A strain, and RNA binding,

structural constituent of ribosome, translation and ribosome in *Pss* B728a strain.

These results showed a highly functional variability of TFs in *P. syringae*.

Topological modularity of the transcription regulatory network exhibited

various functions in biological processes in *Psph* 1448A

Complex networks in nature often exhibit topological and/or functional

modularity(Dittrich, Klau, Rosenwald, Dandekar, & Müller, 2008; Olesen, Bascompte,

Dupont, & Jordano, 2007). To explore the modularity in the TF-binding network, we

used a partitioning algorithm (with a resolution of 0.9) to classify network elements

into different subsets using Gephi. TFs were the primary nodes, and each edge

represented the regulatory flow of TFs to target genes, which were small nodes. Our

analysed TFs and their target genes were divided into 16 modules, each represented

by a different colour **(Figure 6a)**. Multiple applicated modulatory can be achieved

depending on different clustering resolution. In our case, a resolution of 0.9 provided

moderate modularity and ensured that each module contained 2.7% to 12.1% of the

total elements and exhibited correlations with each other. We found that almost all

nodes in the network had connections both within and between modules, indicating

that the 16 modules were not isolated and contributed to extensive information flow

throughout the network to regulate transcription in *Psph* 1448A **(Figure 6b)**. Module

12 appeared to play a central role in facilitating large information flows with other

modules. Module 15 also exhibited transcriptional information exchanges both

between modules and within the same module.

Among the 16 modules, Module 2 was involved in most nodes (443 elements,

including 41 TFs and 402 target genes). Module 3 contained the least TFs (181

elements, including two TFs and 179 target genes). To investigate the potential

correlation between topological modularity and biological functions, we performed

GO term and KEGG pathway enrichment analysis for each module by

hypergeometric test (BH-adjusted $p < 0.05$). As expected, 15 modules were

enriched in specific biological functions **(Figure 6c and d)**. In some modules,

hundreds of elements were assessed. For example, genes in Module 2 (443 nodes),

Module 12 (327 nodes, including 22 TFs and 318 target genes) and Module 14 (432

nodes, including 3 TFs and 324 target genes) were mainly enriched in the regulation

of transcription and DNA binding **(Figure 6c)**. DksA, a TF in Module 2, played a key

role in regulating transcription-coupled DNA repair(Meddows, Savory, Grove, Moore,

& Lloyd, 2005) and also participated in oxidative phosphorylation, amino sugar and

nucleotide sugar metabolism and RNA degradation in *E. coli*. Twenty TFs (such as

CapA, CapB, CysB, FruR and MarR) classified in Module 12 were identified to be

involved in transcriptional regulation. The TF PSPPH3798 located in Module 14 was

observed to be involved in flagellar assembly, and these interactions were confirmed

by EMSA (**Figure S4e**). The genes in Module 4 were enriched in oxidoreductase

activity, and our analysis revealed that the TF PSPPH0755 played a role in regulating

the oxidative phosphorylation pathway (**Figure 6d**). MexT in Module 5, which was

previously associated with motility in *P. syringae* pv. *tabaci* 6605 (Kawakita et al.,

2012), was found to participate in in biofilm formation and QS pathways in our study.

In addition, we not only identified the crucial roles of the TFs PSPPH1951 (Module

6), PSPPH2193 (Module 11) and PSPPH3268 (Module 15) in T3SS pathways,

bacterial motility and biofilm formation, respectively (**Figure 3b–d**), but also reported

their potential biological functions in aminoacyl-tRNA biosynthesis (PSPPH1951),

RNA degradation (PSPPH2193) and catalytic activity (PSPPH3268) (**Figure 6d**). Our

results allowed us to identify the potential regulators of specific pathways and

perform functional predictions for hypothetical proteins in *Psph* 1448A. For instance,

PSPPH1503 in Module 15, which encodes a hypothetical protein, was possibly

correlated with glycerophospholipid metabolism.

Discussion

Most microbial studies on genome-wide transcriptional regulatory network focus on *S.*

cerevisiae and *E. coli*, which reveal the principles of architecture and interactions of

their regulatory networks. The analysis of the transcription regulatory associations in

S. cerevisiae mainly rely on the databases such as YEASTRACT (YEAst Search for

Transcriptional Regulators And Consensus Tracking)(Teixeira et al., 2018). In *E.coli*,

relative complete transcriptional regulatory network has been generated through

integrating three different data sources (RegulonDB, Ecocyc and TRN-SO)(Ma et al.,

2004). However, few study has yet comprehensively evaluated TFs in other

prokaryotic species throughout a genome(Ishihama, Shimada, & Yamazaki, 2016). In

this study, we successfully generated the most complete transcriptional regulatory

network and data source, which profiled the transcriptional regulatory features of both

the aforementioned 100 TFs (Fan et al., 2020) and an additional 170 TFs in *Psph*

1448A through ChIP-seq. By mapping the TF-target hierarchical regulatory networks,

we identified several novel master regulators involved in significant biological

processes. Furthermore, our evolutionary analysis and assessment of the topological

functional modularity of TFs and their respective targets revealed the evolutional

conservation and functional diversity of TFs in *P. syringae*.

Although transcriptional regulatory networks are considered conserved(Perez &

Groisman, 2009), many studies reveal highly functional variability of TFs in inter-

and intra-species(Galardini et al., 2015). These observed diversities between different

strains of the same species mainly result from the expression levels of TFs, contents

of target genes, and differences of binding sequences(Trouillon et al., 2021). In our

study, we observed large differences of DNA-binding characteristic of TF Irp between

the C48 strain and the *Pst* DC3000 strain. The functional diversity of TFs may arise

from the large difference in the contents of target genes and TFs, which are regarded

as the main determinant of transcriptional regulatory(van Duin, Krautz, Rennie, &

Andersson, 2023), although Irp display high homology in these pathovars.

Collaborations between TFs at higher levels (top and middle) were enriched, a pattern

similar to the tendencies observed in human TFs(Gerstein et al., 2012). In particular,

TT TF pairs exhibited a greater degree of cooperative gene regulation, whereas TB TF

pairs accounted for nearly half of direct interactions within all communications.

Furthermore, we observed that both direct physical regulation and cooperative

interactions were the least common among MM TF pairs. By contrast, in humans,

direct regulation tends to occur between TT or TM TF pairs. Furthermore, interactions

between TFs, in any form, within human and yeast transcriptional regulatory

networks are more likely to appear between middle TFs, which act as information

transfer centres(Bhardwaj, Yan, & Gerstein, 2010). This may be attributed to the

higher abundance of bottom-level TFs than higher-level TFs observed in prokaryotic

microorganisms, a pattern also found in *E. coli* (Bhardwaj et al., 2010). This finding

indicates that the bottom-level TFs that are more likely to be regulated are

evolutionarily preferred in multicellular eukaryotic organisms. However, they found

that the deficient bottom-level TFs are more commonly co-associated with other

same-level TFs in *E. coli*. When comparing co-associated TF pairs and the

cooperativity of TFs, we observed a distinct and inverse relationship in *Psph* 1448A

compared with yeast or *E. coli* (**Figure 1B**). The enriched cooperativity of

bottom-level TFs with high co-associated scores indicated that these bottom-level TFs

preferred to coregulate target genes by binding to the same peak locations. Notably,

seven TFs without correlations with other TFs appeared to independently participate

in biological processes. These findings not only shed light on the inherent properties

of direct regulation and co-association across various species but also indicate the

unique characteristics of *Psph* 1448A to response to dynamic environmental

variations.

The fundamental units of a transcriptional regulatory network are positive and

negative loops. For a more comprehensive description, these regulatory units can be

classified into six submodules, namely autoregulation, multicomponent loops,

feedforward loops, single-input, multi-input and regulator chain(Lee et al., 2002). In

yeast, only 10 TFs were found to autoregulate themselves, whereas the majority of

regulatory units among 116 TFs in *E. coli* exhibit autoregulation(Thieffry, Huerta, Pé

rez-Rueda, & Collado-Vides, 1998). Similarly, we identified 92 autoregulators in our

transcriptional regulatory network in *P. syringae*. Feedforward loops are highly

prevalent in eukaryotic transcriptional regulatory networks, such as human and

yeast(Gerstein et al., 2012; Lee et al., 2002). The 696 feedforward loops (M13)

identified in our study also highlighted this cooperative regulation of TFs in response

to small signals in prokaryotic species, such as *P. syringae*. This enriched submodule

is regarded as a temporal switch that provides constant feedback to respond rapidly to

signal impacts(Shen-Orr et al., 2002). Multistep regulation assists master regulators in

enhancing the initial information flow(Goldbeter & Koshland Jr, 1984). Unlike M3,

the most abundant module (868) in the human TF regulatory network, we observed

that M1 was the most prevalent module (24,479) in the *Psph* 1448A transcriptional

regulatory network. This finding indicates that *Psph* 1448A prefers achieving

transcriptional regulation through few TFs to ensure rapid transmission and response

to environmental signals.

Despite more than 20 years of research, our understanding of the global regulatory

network in the plant pathogen *P. syringae* remains inadequate. Our previous studies

reported 16 key virulence-related regulators(Shao et al., 2021), 25 master

virulence-related regulators(Fan et al., 2020) and seven global regulators acting as

TCSs(Xie et al., 2022). However, the regulatory relationships and functional crosstalk

among all TFs in *P. syringae* remain unknown. In this study, we integrated all

available interaction information concerning almost all TFs in *Psph* 1448A and

mapped the first comprehensive transcriptional regulatory network in this plant

pathogen (**Figure 7**). This network offers a global view of the multiple functions of

TFs in *Psph* 1448A. We also identified 35 vital TFs that participate in virulence

pathways and 111 key TFs involved in metabolic pathways across the global

transcriptional regulatory network in *Psph* 1448A. This analysis uncovered new

functions of previously characterised TFs, such as MexT. In addition to the previously

reported *mexEF-oprN* operon,(Sawada et al., 2018) this study identified *fleQ* (a

flagellar regulator) and *shcF* (a type III chaperone protein) as the targets of MexT. In

addition, we investigated the functional evolution and potential intra-species

variability of TFs in *P. syringae,* demonstrating the functional diversity of TFs among

P. syringae species during their long course of evolution. Based on the

aforementioned results, we established the *Psph* 1448A transcriptional regulatory

network (PSTRnet) database, which contains detailed binding peak information for all

TFs in *Psph* 1448A (https://jiadhuang0417.shinyapps.io/PSTF_NET/). This database

serves as a valuable platform for presenting, searching and downloading regulatory

information of transcription in *Psph* 1448A.

We observed crosstalk between not only TFs but also various pathways. For example,

the TF PSPPH4700 directly regulated *fleQ* and *hrpR* and indirectly regulated *fleQ* and

hrpR through the PSPPH4700/PSPPH4324/PSPPH0755 cascade. Feedforward loop

modules are usually coherent, meaning that the direct effect of downstream TFs has

the same regulatory direction (positive or negative) as the indirect effect of upstream

TFs(Shen-Orr et al., 2002). This observation enhanced our understanding of the

influence of TFs in the network on their target genes based on the identified effects of

other TFs. Furthermore, the TF PSPPH4700 was identified to bind to the promoters of

genes involved in seven metabolic pathways, namely the TCA cycle, oxidative

phosphorylation, methyl-accepting, amino acid biosynthesis, RNA polymerase, ABC

transportation and DNA replication. Our results proved the possibility of crosstalk

between different pathways in *Psph* 1448A.

Taken together, our study provides comprehensive insights into the DNA-binding

characteristics and potential regulatory pathways of almost all annotated TFs in *Psph*

1448A. The global transcriptional regulatory network can not only contribute to the

development of novel drugs to combat *P. syringae* infections but also advance

research on the molecular mechanisms of TFs in other pathogens.

Materials and Methods

Bacterial strains, culture media, plasmids, and primers

The bacterial strains, plasmids and primers used in this study are listed in

Supplementary Table 6. *P. syringae Psph* 1448A, *Pst* DC3000, *Pss* B728a and *Psa*

C48 strains were grown at 28 °C in King' B (KB) medium shaking at 220 rpm or KB

agar plates(King, Ward, & Raney, 1954). *E. coli* BL21(DE3) or DH5α strains were

grown at 37 °C in Luria-Bertani medium (LB) shaking at 220 rpm or on LB agar

plates. Antibiotics used for *P. syringae Psph* 1448A, *Pst* DC3000 and *Pss* B728a wide

type (WT) strains and mutants were rifampicin at 50 μg/ml; *P. syringae Psph* 1448A,

Pst DC3000 and *Pss* B728a overexpression strains were rifampicin at 50 μg/ml and

spectinomycin at 50 μg/ml, C48 overexpression strains were spectinomycin at

50 μg/ml; *P. syringae* 1448A strains with pK18mobsacB plasmids for mutant

construction were rifampicin at 50 μg/ml and kanamycin at 50 μg/ml. Antibiotics

used for *E. coli* with pET28a plasmids for protein purification, and pK18*mobsacB*

plasmids were kanamycin at 50 μg/ml; *E. coli* with pHM1 plasmids for

overexpression strain construction were spectinomycin at 50 μg/ml.

Construction of overexpression strains and mutants

Overexpression strains and mutants of *P. syringae* were constructed as previously

described(Kvitko & Collmer, 2011; Shao et al., 2021). In brief, for overexpression

strain, the open reading frame (Caillet, Baron, Boni, Caillet-Saguy, & Hajnsdorf) of

each TF-coding gene was amplified by PCR from *P. syringae* genome and cloned into

pHM1 plasmid. The ligated fragments were inserted into the digested pHM1 plasmids

(*Hind*III) using ClonExpress MultiS One Step Cloning Kit (Vazyme). The

recombinant plasmids were transformed into the *P. syringae* wild-type strain. The

single colonies were confirmed by western blot. For mutants, the upstream ($\sim$1500-bp)

and downstream ($\sim$1000-bp) fragments of TF ORF were amplified by PCR from the

Psph 1448A genome and digested with *Xba*I respectively, and then ligated by T4

DNA ligase (Jumper et al.). The ligated fragments were inserted into the digested

pK18*mobsacB* plasmids (*Xba*I) using ClonExpress MultiS One Step Cloning Kit

(Vazyme). The recombinant plasmids were transformed into the *Psph* 1448A

wild-type strain. The single colonies were selected to the sucrose plates, and further

screened in two kinds of KB plates (with kanamycin/rifampin and with rifampin)

concurrently. The single colonies losing kanamycin resistance were further verified by

real-time quantitative PCR (RT-qPCR) to detect the mRNA level of corresponding TF

genes.

ChIP-seq analyses

Psph 1448A was used for our main ChIP-seq. As previous description(Blasco et al.,

2012), the overexpression strains of corresponding TFs and *P. syringae* WT with

empty pHM1 plasmid was cultured in 30 mL KB medium to OD_{600}=0.6. Bacterial

cultures were cross-linked with 1% formaldehyde for 10 min at 28°C and then the

reaction was stopped by the addition of 125 mM glycine for 5 min. The centrifugated

bacteria were washed twice with Tris Buffer (20 mM Tris-HCl [pH 7.5] and 150 mM

NaCl) and washed again with IP Buffer (50 mM HEPES–KOH [pH 7.5], 150 mM

NaCl, 1 mM EDTA, 1% Triton X-100, 0.1% sodium deoxycholate, 0.1% SDS, and

mini-protease inhibitor cocktail). The centrifugated bacteria were preserved at -80□

or continued for the next experiments. The bacteria were subjected with IP Buffer and

then sonicated to pull down the DNA fragments (150-300-bp). The supernatant was

the DNA-TF-HA tag complex and used as IP samples. IP experiments and control

sample were incubated with agarose conjugated anti-HA antibodies (Sigma) in IP

Buffer. The complex of DNA-TF-anti-HA agarose was applied to washing, crosslink

reversal, and purification(Blasco et al., 2012). The 150-250-bp DNA fragments were

selected for library construction. The libraries were sequenced using the HiSeq 2000

system (Illumina). Two biological replications have been performed for all ChIP-seq

experiments. ChIP-seq reads were mapped to the *P. syringae Psph* 1448A genome

(NC_005773.3), *Pst* DC3000 (NC_004578.1), *Pss* B728a (NC_007005.1) and *Psa*

C48 (NZ_CP032631.1) using Bowtie2 (version 2.3.4.3)(Zhang et al., 2008).

Subsequently, binding peaks (q < 0.01) were identified using MACS2 software

(version 2.1.0). The enriched loci for each TF were annotated using the R package

ChIPpeakAnno (version 3.18.2)(L. J. Zhu et al., 2010). We defined intergenic region

before each TF sequence as the promoter region. As pHM1 plasmid takes its own

promoter, we amplified the TF-coding sequence and cloned into the plasmid. The TF

protein expression was activated by the promoter of plasmid.

Electrophoretic mobility-shift assay

DNA probes were amplified from *P. syringae* 1448A genome by PCR using primers

listed in Supplementary Table 6. The probes (20 ng) were mixed with various

concentrations of proteins in 20 μL of gel shift buffer (10 mM Tris-HCl, pH 7.4, 50

mM KCl, 5 mM MgCl$_2$, 10% glycerol). After incubation at room temperature for 20

min, the samples were analyzed by 6% native polyacrylamide gel electrophoresis (90

V for 90 min for sample separation). The gels were subjected to Gel Red dye (Tiangen

Biotech) for 5 min and photographed by using a gel imaging system (Bio-Rad). The

assay was repeated at least three times with similar results.

RT-qPCR

The RT-qPCR primers used are shown Supplementary Table 6 in the Supplemental

Information. The cultured bacteria were grown to OD$_{600}$=0.6 and the total RNA were

purified with Bacteria Total RNA Isolation Kit (Sangon Biotech). The RNA

concentrations were measured using a NanoDrop 2000 spectrophotometer

(ThermoFisher). cDNA synthesis was performed using HiScript III RT SuperMix

(Vazyme, China). RT-qPCR was performed with a SuperReal Premix Plus (SYBR

Green) kit (Vazyme, China) according to the manufacturer's instructions. The

reactions used 100 nM primers and were run for 40 cycles at 95°C for 30 s and 95°C

for 10 s, and at 60°C for 30 s. The fold change represents relative expression level of

mRNA, which can be estimated by the values of $2^{-(\Delta\Delta Ct)}$. The relative expression of

target genes in WT was set to 1. All the reactions were conducted with three repeats.

Motility assay

The motility assay was performed based on our previous study(Shao et al., 2021).

Swimming plates were KB agar plates containing 0.3% agar (MP Biomedicals, UK)

and rifampicin at 25 μg/ml. Overnight bacterial cultures were inoculated on

swimming plates as 2 μL aliquots and incubated at room temperature for 3-5 days.

Finally, the diameter of motility trace represented the swimming motility of *P.*

syringae strains. Photographs were taken by using the Bio-Rad imaging system. The

assay was repeated at least three times with similar results.

Congo red assay

Congo red assay was performed as previous study(Shao et al., 2021). Congo red

plates were KB agar plates containing 1.0% agar (MP Biomedicals, UK) and

rifampicin at 25 μg/ml. Overnight bacterial cultures were inoculated on Congo red

plates as 2 μL aliquots and incubated at 28□. The colony staining was photographed

after 5-7 days. The assay was repeated at least three times with similar results.

Biofilm formation assay

Biofilm formation assay was performed as previously describe(Shao et al., 2019).

Overnight bacterial cultures were transferred to a sterile 10 mL borosilicate tube

containing 2 ml KB medium with rifampicin (25 μg/ml) with the original

concentration $OD_{600} = 0.1$. The cultures grew at room temperature for 36 hr without

shaking. 0.1% crystal violet was used to stain the biofilm adhered to the tube for 20

min without shaking. Tubes were washed for more than three times with distilled

deionized water gently and other components on the tube loosely was washed off. The

tubes were dry and photographed. The remaining crystal violet was fully dissolved in

1 ml 95% ethanol with constantly shaking and measured its optical density at 590 nm

(Biotek microplate reader). The assay was repeated at least three times with similar

results.

Plant infection assay

Bean leaves were used for pathogenicity. Plant materials were grown in a greenhouse

as described previously(Pan et al., 2020; Yanmei Xiao et al., 2007). Overnight

bacterial cultures were diluted to OD_{600}=0.2. The diluted cultures were washed three

times by sterile water resuspend with equal volume sterile water, and then continued

to diluted 100-fold. The diluted bacterial solution was individually injected into bean

leaves form stomata on the underside of a leaves. The plant was continued to grow at

28□. Disease symptoms on bean leaves were measured 5 days after inoculation.

Infected leaves (1 cm^2) were removed after 5 days growth and homogenized in sterile

water. Bacteria were diluted to proper concentration and plated on a KB plate

containing rifampicin at 25 µg/ml for bacterial count. The bacterial number

represented colony-forming unit (CFU) formed in one leaf disks (1 cm^2) with four

repeats.

Network and functional enrichment analysis

810 Network analyses were performed on Gephi 0.10. Functional annotations were

811 retrieved from the *Pseudomonas* database and GO functional enrichment analyses and

812 KEGG analysis were performed using DAVID v6.8.

Statistical analysis

814 Two-tailed Student's *t* tests were performed using Microsoft Office Excel 2010.

815   *$P < 0.05$, **$P < 0.01$, and ***$P < 0.001$ and results represent means $\pm$ SD.

816   All experiments were repeated at least twice.

Data Availability

Sequencing data have been deposited and publicly available in Gene Expression Omnibus (GEO) under accession number GSE247395. Source data contain the numerical data used to generate the figures of EMSA. Analysis codes have been deposited as accession https://github.com/dengxinb2315/PS-PATRnet-code. Hierarchical information and functional categories of TFs are available in **Supplementary Table 1-3**. Evolutionary details are provided in **Supplementary Table 4**. Primers and strains used in this paper are provided in **Supplementary Table 6**.

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X.D., Y.S., J.W. and J. H. conceived the project. Y.S., J.W., J.H., S.L. and Y.L. carried out experiments. J.H, S.L., Y.L. and B. L. performed data analysis. X.D., Y.S, J.W., and J.H. wrote the paper.

Ethics declarations

Competing interests

The authors declare no competing interests.

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 Figure legends

 Figure 1. Hierarchical height and collaboration of TFs reveal the multiple

 regulatory patterns in *Psph* 1448A.

 a, Close-up representation of 262 TFs hierarchy in *Psph* 1448A (eight TFs showed no

 hierarchical characteristic). Nodes depict TFs. Colors of edges represent source-bases.

 b, Enrichment of different collaborating (direct interaction, indirect interaction and

 cooperativity) TF-pairs at top (T), middle (M) and bottom (B) levels. We defined

 indirect interaction if two TFs co-associated with one target DNA. Cooperative TF

 pair was defined if their common target is from a TF. Gray nodes below the graph

 represent TFs. The dashed orange line indicates the averaged level of collaboration. **c,**

 Thirteen Three-node sub-modules with the number of occurrences and an example.

 Spire loop is the most enriched sub-module. Edges represent the regulatory direction.

 d, Autoregulations are accompanied by the number of occurrences and 13

 auto-regulators as examples.

 Figure 2. Bottom TFs share the same binding sequences to coregulate in *Psph*

 1448A.

 a, The co-association map for 170 TFs in *Psph* 1448A shows the co-associated scores

 of binding peaks of these TFs (rows) that overlap each TF peak (columns). The three

 colored rectangles represent three different TF levels. C1-C4 represent four clusters of

 TFs according to the co-associated scores. The TFs in corresponding cluster are

 shown in the circle diagram. Orange nodes represent top TFs. Green nodes represent

middle TFs. Purple nodes represent bottom TFs. The colors of edges are the mixture

of two source TF colors. **b,** The heat-map of MexT indicates the associated scores of

binding peaks of TFs in C3 (columns) that overlap the binding peaks (rows) of MexT.

PSPPH2411, PSPPH3643 and CysI are the top 3 TFs with high associated scores. **c,**

UpSet plot shows the number of genes uniquely targeted TFs or co-targeted by

multiple TFs in C3. The vertical black lines represent shared TF-binding sites. The y

axis represents the number of overlapped binding sites across the linked TFs. The x

axis represents the number of binding sites for each TF. Orange line represents the

most enriched gene *cysI* that is co-targeted by 12 TFs in C3. **d,** Motifs of MexT and

other 10 TFs in C3 which show similar binding sequences.

Figure 3. Virulence hierarchical regulatory network reveals 35 TFs involved in

virulence.

a, Virulence hierarchical regulatory network shows the TF hierarchy and the large

pool of target genes of multi-TF. Target genes are related with seven key virulence

pathways, including biofilm formation, secondary messengers, motility, T3SS, QS,

phytotoxin production and siderophore transporter. Orange nodes represent top TFs.

Green nodes represent middle TFs. Purple nodes represent bottom TFs. Blue nodes

represent target genes. Yellow edges represent downward point. Red edges represent

upward point. **b,** The head-to-head binding motif of PSPPH1951, the validation of the

binding sites of PSPPH1951 by EMSA, and the detection of expression of target gene

hopAE1 in WT, ΔPSPPH1951 and complementary strain by RT-qPCR. The validated

binding sites are from the promoters of the *hrpR*, *hopAE1* and *hopAH2*. **c,** The

binding motif of PSPPH2193 is head-to-head. EMSA confirms the direct binding of

PSPPH2193 to the promoters of *fleQ* and *flhF*. RT-qPCR confirms the positive

regulation of PSPPH2193 on the expression of *fleQ* and *flhF*. Motility assay validates

the weaker motility of ΔPSPPH2193 than WT and complementary strain. **d,** The

binding motif of PSPPH3268 is head-to-head. EMSA confirms the direct binding of

PSPPH3268 to the promoters of *hrpR*, *alg44* and *pilM*. RT-qPCR confirms the

negative regulation of PSPPH3268 on the expression of *alg44*, *algX* and *pilM*. Crystal

violate staining assay and the quantification of biofilm formation validate the negative

regulation of PSPPH3268 on the biofilm formation. Congo red assay confirms the

negative regulation of PSPPH3268 on colony morphologies and EPS production.

Student's *t* test. n.s., not significant, $*P\square\leq\square0.05$, $**P\square\leq\square0.01$, and $***P\square\leq\square0.001$.

Figure 4. Hundreds of TFs are identified to participate in metabolic pathways.

a, Metabolic hierarchical regulatory network shows the TF hierarchy and the large

pool of target genes of multi-TF. Target genes are related with eight key metabolic

pathways, including biosynthesis of amino acids, DNA replication, ABC transporter,

oxidative phosphorylation, TCA cycle, RNA polymerase, phosphonate metabolism

and methyl-accepting chemotaxis. Orange nodes represent top TFs. Green nodes

represent middle TFs. Purple nodes represent bottom TFs. Blue nodes represent target

genes. Yellow edges represent the direct interaction. **b,** Radar plots show the putative

key regulators identified in six different metabolic pathways, including TCA cycle,

oxidative phosphorylation, methyl-accepting, biosynthesis of amino acids, RNA

polymerase and ABC transporter. Each radiation line represents a key regulator, and

the radial length of the thick colored line is the rate of target genes to the associated

genes, representing the significance of the enrichment of the TF target genes within

each pathway. **c,** TFs involved in the methyl-accepting pathway bound to the

promoters of TFs in the same category. Brown nodes represent the TFs that are

responsible for methyl-accepting pathway. The brown arrows point to the targeted TFs.

d, TFs involved in the oxidative phosphorylation pathway bind to the promoters of

TFs in the methyl-accepting pathway. Red nodes represent the TFs that are

responsible for oxidative phosphorylation pathway. Brown nodes represent the TFs

that are responsible for methyl-accepting pathway. Blue nodes represent the TFs

involved in these two pathways. The arrows point to the targeted TFs and the arrow

colors are source-based.

Figure 5. Various conservations are observed in TFs between different *P. syringae*

pathovars.

a, Proportion of the TF target genes detected in one, two, three or four tested genomes.

c1-c4 represent the conservation of targets in one, two, three and four strains. **b-c,**

Repartition of the total pool of target genes (**b**) or TF-target interactions (**c**) in four

tested strains. **d,** Enrichment coverage tracks of ChIP-seq against negative controls for

TF Irp with binding sites on the promoter of *gidA* in all four genomes. **e,** Enrichment

coverage tracks of ChIP-seq against negative controls (gray tracks) for the three TFs

(Irp, PSPPH2193 and PSPPH4127) with binding sites on the promoter of *rpoD* in four

tested strains.

Figure 6. Functional modularized regulatory network in *P. syringae* exhibits the

specific functions of both TFs and their target genes.

a, The functional categorical regulation network in *Psph* 1448A analyzed by Gephi

(resolution 0.9). The 16 modules (both TFs and their target genes) are labeled in

different colors. TF nodes are shown as corresponding sized circles representing their

expression level. Their target genes are shown as corresponding-colored dots in the

background. TF-target edges are shown as corresponding-colored lines between nodes.

b, Graph diagram indicates the connections between TF and their targets in modules.

Module nodes are shown as corresponding-colored circles with size proportional to

the number of nodes within. Edge colors are source-based, and edge thicknesses

represent the connected quantity between modules. **c-d,** Functional category

enrichment analysis of genes in each module, $p < 0.05$.

Figure 7. Global transcriptional regulatory network in *Psph* 1448A.

The integrated transcriptional regulatory network reflects the interactions between all

TFs classified into 39 families from different DNA binding domain types and target

genes annotated from pathway annotations. The targets are shown in 12 pathways

with various colors. TF nodes are gray as corresponding sizes representing TF number

in the family. Edge colors are target-based.

Figure S1. Summary of ChIP-seq results in *Psph* 1448A.

a, Locations of all the 301 annotated TFs in *Psph* 1448A genome. Blue lines represent

the TF loci. **b,** ChIPed TFs are classified as respective family with different colors.

Square size represents the targeted enrichment of each TF. **c,** The preferential

enrichment at different genome loci of each ChIPed TF, including upstream,

overlapStart, inside, overlapEnd, downstream and includeFeature regions.

Figure S2. Graph diagram of feed-forward loop in *Psph* 1448A.

TF columns in feed-forward loop (M13, n=696) are classified as families. TF-TF

edges are distinguished with separate colors.

Figure S3. Co-association and virulence-related functional category of TFs at top

level.

a-b, The co-associated map and functional category of top TFs. The peak number of

top TFs are shown at upper corresponding to the TFs in the lower map. Functional

enrichment analysis was performed by hypergeometric test (BH-adjusted $p < 0.05$).

Figure S4. Co-association and virulence-related functional category of TFs at

middle level.

a-b, The co-associated map and functional category of middle TFs. The peak number

of middle TFs are shown at upper corresponding to the TFs in the lower map.

Functional enrichment analysis was performed by hypergeometric test (BH-adjusted

$p < 0.05$).

Figure S5. Co-association and virulence-related functional category of TFs at

bottom level.

a-b, The co-associated map and functional category of bottom TFs. The peak number

of bottom TFs are shown at upper corresponding to the TFs in the lower map.

Functional enrichment analysis was performed by hypergeometric test (BH-adjusted

$p < 0.05$).

Figure S6. The validation of the binding sites of virulence-related TFs in *Psph*

1448A.

a-c, The promoter of *rpoD* is the negative control of PSPPH1951 and PSPPH2193.

The promoter of PSPPH3658 is the negative control of PSPPH3268. **d,** The validation

of binding sites of PSPPH1951. The validated binding sites are from the promoters of

pilZ, *pilF* and *pilG*. **e,** The validation of binding sites of PSPPH3798. The validated

binding sites are from the promoters of *fliK*, *fliE*, *fliD* and *fleQ*. The promoter of

PSPPH2263 is the negative control. **f,** A regulatory cascade of PSPPH3504. / means

sibling nodes. – means downward regulation. **g,** Bacterial number (colony-forming

units, CFUs) of Δ2193 strain on the bean leaf surface after dip inoculation. $*P < 0.05$.

Results were indicated as mean $\pm$ SD. All experiments were repeated at least three

times.

Figure S7. Metabolic functional category of TFs at three different levels.

a-c, Functional category according to different metabolic pathways of top (**a**), middle

(**b**) and bottom (**c**) TFs.

Figure S8. Key TFs in different metabolic pathways.

a, Radar plots show the putative key regulators identified in two different metabolic

pathways, including DNA replication, and phosphonate and phosphonate metabolism.

Each radiation line represents a key regulator, and the radial length of the thick

colored line is the rate of target genes to the associated genes, representing the

significance of the enrichment of the TF target genes within each pathway. **b,** The

monomer motif of PSPPH0755 and the validation of the binding sites of PSPPH0755

by EMSA. The validated binding sites are from promoters of PSPPH5210 and

PSPPH3109. The promoter of PSPPH4598 is negative control. **c,** The head-to-head

motif of PSPPH3798 and the validation of the binding sites of PSPPH3798 by EMSA.

The validated binding sites are from promoters of PSPPH3881 and PSPPH5119. **d,**

The monomer motif of PSPPH4638 and the validation of the binding sites of

PSPPH4638 by EMSA. The validated binding sites are from promoters of

PSPPH3881 and PSPPH0550. The promoter of PSPPH4598 is the negative control.

Figure S9. Motif and functional enrichment analysis of PSPPH3122 in *Psph*

1448A, *Pss* B728a and *Pst* DC3000 strains.

Motifs of PSPPH3122 in *Psph* 1448A (**a**) and *Pst* DC3000 (**b**) strains. **c**, GO analysis

by hypergeometric test (BH-adjusted $p < 0.05$) of PSPPH3122 in *Psph* 1448A, *Pst*

DC3000 and *Pss* B728a strains.

Figure S10. Motif and functional enrichment analysis of PSPPH4127 in *Psph*

1448A, *Pss* B728a, *Pst* DC3000 and *Psa* C48 strains.

Motifs of PSPPH4127 in *Psph* 1448A (**a**), *Pss* B728a (**b**), *Pst* DC3000 (**c**) and *Psa*

C48 (**d**) strains. **e**, GO analysis by hypergeometric test (BH-adjusted $p < 0.05$) of

PSPPH4127 in *Psph* 1448A, *Pss* B728a, *Pst* DC3000 and *Psa* C48 strains.

Figure 1.

Figure 2.

Figure 3.

Figure 4.

Figure 5.

Figure 6.

Figure 7.

Direct interaction

Figure S1

a

c

b

Figure S2

Feed-forward loop

n=696

Figure S3

a

b

Figure S4

a

b

Figure S5

a

b

a PSPPH1951 μM
0   2
Promoter of *rpoD*
(Negative control)

b PSPPH2193 μM
0   4
Promoter of *rpoD*
(Negative control)

c PSPPH3268 μM
0   1
Promoter of
PSPPH3658
(Negative control)

d PSPPH1951 μM
0   2
Promoter of *pilZ*

PSPPH1951 μM
0   2
Promoter of *pilF*

PSPPH1951 μM
0   2
Promoter of *pilG*

e PSPPH3798 μM
0   1
Promoter of *fliK*

PSPPH3798 μM
0   1
Promoter of *fliE*

PSPPH3798 μM
0   1
Promoter of *fliD*

PSPPH3798 μM
0   1
Promoter of *fleQ*

PSPPH3798 μM
0   1
Promoter of
PSPPH4518
(Negative control)

f

g Bacterial number (CFU/cm^2)
10^9
10^8
10^7
10^6
10^5
WT   Δ*2193*

Figure S7.

Figure S8.

a

DNA replication

Phosphonate and phosphinate metabolism

b

Monomer

PS0755 (µM)

0   1

Promoter of PS4598
(Negative control)

PS0755 (µM)

0   1

Promoter of PS5210
(ATP synthase F0F1)

PS0755 (µM)

0   1

Promoter of PS3109
(NADH dehydrogenase *nuoA*)

c

Head-to-head

PS3798 (µM)

0   1

Promoter of PS3881
(methyl-accepting
chemotaxis gene *wspA*)

PS3798 (µM)

0   1

Promoter of PS5119
(phosphate transport system
regulatory gene *phoU*)

d

Monomer

PS4638 (µM)

0   4

Promoter of PS3881
(methyl-accepting
chemotaxis gene *wspA*)

PS4638 (µM)

0   4

Promoter of PS0550
(phosphoserine
phosphatase *serB*)

PS4638 (µM)

0   4

Promoter of PS4117
(Negative control)

Figure S9

Psph 1448A

a

Motif of PSPPH3122

Pst DC3000

b

Motif of PSPPH3122

c   GO_plot of PSPPH3122

GO:0006412

translation

translation

GO:0006310

DNA recombination

GO:0005840

ribosome

ribosome

GO:0003735

structural constituent of ribosome

structural constituent of ribosome

GO:0003723

RNA binding

GO:0003677

DNA binding   DNA binding

GO:0000150

recombinase activity

GO (y-axis label)

Count (x-axis label), 10, 20

strain
- *Pss* B728a
- *Pst* DC3000
- *Psph* 1448A

−log10(p.adjust)
- 2.5
- 5.0
- 7.5
- 10.0
- 12.5

Figure S10

Psph 1448A

a

Motif of PSPPH4127

Pss B728a

b

Motif of PSPPH4127

Pst DC3000

c

Motif of PSPPH4127

Psa C48

d

Motif of PSPPH4127

e

GO_plot of PSPPH4127