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# Gene and Protein Network Analysis of  $AmpC$   $\beta$  Lactamase

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Abstract  $AmpC$   $\beta$ -lactamase is a cephalosporinase, which exhibits resistance against all existing  $\beta$ -lactam antibiotics except carbapenems. Their occurrence in many bacterial pathogens poses a threat to public health and is a growing concern in the medical world. The ampC gene is highly inducible in the presence of  $\beta$ -lactam antibiotics and can be expressed in high levels due to mutation. This inducible expression is regulated by several functional genes. Several studies on functional relationship of these genes and its resistance mechanisms are carried out but it still lacks comprehensible evidences. Thus, in our current study, we used computational gene networks to analyze ampC gene. Based on its interaction type, co-expression, Gene Ontology, and text mining, a functional interaction network is constructed. Around 247 functional genes in 15 different bacterial genus have a functional association with ampC gene. It is predicted that 19.8 % ampD, 13.3 % frdD, 8.5 % gcvA, 2.4 % ampR, and 55.7 % of other functional partners are associated with ampC gene. Our present study provides a glimpse about the functional gene network of ampC gene and also provides the integrated evidence for  $ampC$  gene in regulating the  $\beta$ -lactamase production and its role in antibiotic resistance.

Keywords Gene regulatory networks - AmpC b-lactamases - Molecular sequence annotation

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## Abbreviations



# Introduction

Over decades, the  $\beta$ -lactam antibiotics are being widely used to treat the bacterial infections. However, the resistance shown by the bacterial pathogens toward the  $\beta$ -lactam antibiotics is one of the serious problems worldwide [\[1](#page-12-0)]. Bacteria persist several modes of resistance and one of the most important modes is the synthesis of  $\beta$ -lactamase [\[2](#page-12-0)].  $\beta$ -lactamase hydrolyzes the  $\beta$ -lactam ring of  $\beta$ -lactam antibiotics and makes it ineffective [\[3](#page-12-0), [4](#page-13-0)].

The *amp* genes are first discovered in *Enterobacter cloacae* [[5\]](#page-13-0). Among various *amp* genes, the  $ampC$  gene is chromosomally encoded to produce cephalosporinase which exhibits resistance to a wide variety of  $\beta$ -lactam antibiotics including penicillin, narrow and broad spectrum cephalosporins, and also  $\beta$ -lactamase inhibitors [[6\]](#page-13-0). AmpC b-lactamase is placed in ''class C'' of Ambler molecular classification and ''Group 1'' of Bush functional classification. This is the first reported bacterial enzyme that destroyed penicillin [[7\]](#page-13-0). Due to their inducibility and expression in response to certain  $\beta$ -lactam antibiotics, AmpC  $\beta$ -lactamase is a clinically important enzyme [\[8](#page-13-0)]. AmpC β-lactamases are present in Gram-negative bacteria including in human pathogens such as Acinetobacter spp.,

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Aeromonas sp., Citrobacter freundii, Enterobacter sp., Escherichia coli, Pseudomonas aeruginosa, and Yersinia enterocolitica [\[9](#page-13-0), [10\]](#page-13-0). Chromosomal AmpC  $\beta$ -lactamases are inducible. Plasmid-mediated AmpC  $\beta$ -lactamases are formed through the transfer of chromosomal genes and having similar substrate profile [[11,](#page-13-0) [12](#page-13-0)]. Thus, it is a major threat for the successful treatment of bacterial infections.

The presence of  $\beta$ -lactam antibiotics or gene mutation can induce the expression level from 100 to 1,000 fold [[13,](#page-13-0) [14](#page-13-0)] resulting in the  $\beta$ -lactam antibiotic resistance [\[15](#page-13-0)]. This inducible expression of  $AmpC$   $\beta$ -lactamase is regulated by several regulatory genes through cell wall recycling pathway along with *ampC* gene. Several studies reveal that  $ampC$  induction pathway requires three major  $amp$  gene products namely, the AmpG permease, the AmpD cytoplasmic amidases, and the transcriptional regulator AmpR [\[5](#page-13-0), [14](#page-13-0), [16–18\]](#page-13-0).

In the present study, in silico functional interaction network analysis of ampC gene is done using various integrated evidence-based approaches such as network, pathway, functional enrichment analysis, and multiple sequence analysis.

Network visualization is done to identify the ampC gene, associated mutations, and its functional partners' role in antibiotic resistance through regulation of  $\beta$ -lactamase production.

Gene networks generally depict a large number of interactions. They provide information on the physiological state of an organism. Interaction types can be studied by the construction of biochemical networks at various levels. Significant biological information can be extracted from the literature mining  $[19, 20]$  $[19, 20]$  $[19, 20]$  $[19, 20]$ . This *ampC* gene network study provides the knowledge on associated genes/ expressed proteins which are involved in regulation of  $ampC$  gene expression and in the synthesis of  $\beta$ -lactamases. The constructed  $ampC$  gene network also provides a valuable insight about the associated functional partners and their interactions in the regulation of  $\beta$ -lactamases production.

# Materials and Methods

## STRING Network Analysis

We used "search tool for the retrieval of interacting genes proteins'' (STRING 9.0) for the study, a pre-computed database resource, involved in the analysis of gene/protein interactions. Gene was represented as ''node'', while the interactions between any two genes/proteins were represented as an ''edge''. There were direct (physical) and indirect (functional) interactions/associations. The associations were derived from various sources such as high-

throughput experimental data, mining of databases, literature, and analyses of co-expressed genes. Interactions between target gene and their closely related functional partners in the network were determined as combined confidence score. STRING provides a probabilistic confidence score for all associations. The scores were given by comparing the group of associations with manually created classification scheme of KEGG database. Each score represents a given association that provides information about the functional linkage between two proteins, i.e., least specific between a pair of proteins annotated in the same pathway. Majority of different scores of interaction or associated data from STRING were highlighted separately; in addition, a combined score is also calculated when the support for a given association is more than one. Combined score indicates the higher confidence. The confidence score values ranged from the lowest to highest. The highest confidence score was in the range of 0.9–1.0, high confidence score was of 0.7–0.9, medium confidence score was of 0.4–0.7, and low confidence score was up to 0.4  $[21-27]$ .

#### Pathway Databases and Sources

Penicillin and cephalosporin biosynthesis pathway for associated genes and other related information was retrieved from Kyoto Encyclopedia of Genes and Genomes (KEGG) database. Protein functions were described in the metabolic pathway database of KEGG [\[28](#page-13-0)].

Functional Enrichment Analysis by Gene Ontology

Gene Ontology (GO) and annotations were collected from the UniProt [[29–31\]](#page-13-0). STRING-based GO was grouped by using the  $p$  value. The  $p$  values and functional annotations such as biological process, molecular function, and cellular component for functional partners were extracted [\[21–27](#page-13-0)].

#### Multiple Sequence Analysis

The molecular evolutionary genetics analysis software [MEGA] (version 4) was used for multiple sequence alignment [[32,](#page-13-0) [33](#page-13-0)]. The MEGA was an integrated tool designed for comparative analysis of gene sequences and inferring phylogenetic trees by estimating the rates of molecular evolution. All AmpC protein sequences are subjected to ScanPROSITE web server [[34\]](#page-13-0) and Motif search tool [PROSITE pattern and ProDom] [\[35–38](#page-13-0)] to identify the pattern present in the sequences.

Construction of Gene Interaction Network

Graphical network model was generated by using Cytoscape software. Cytoscape was a free software package

<span id="page-2-0"></span>



scores

0.745

0.913

0.856 0.788

0.763 0.865

 $0.92$  $0.853$ 0.827 0.792 0.792

0.853 0.747 0.865

0.846 0.765  $0.7\,$ 0.947 0.931 0.759

0.853 0.787

0.857 0.791

 $0.855$ 0.784

0.852 0.747



# Table 1 continued

Functional partners Confidence

 $ampD$  0.953 frdD 0.775

ampD 0.953  $gcvA$  0.85 frdD 0.794  $ampD$  0.953 gcvA 0.853 frdD 0.75

ampD 0.953 gcvA 0.852 frdD 0.795 ampD 0.953 gcvA 0.851 frdD 0.798 ampD 0.8 EFER\_1643 0.8 frdD 0.757 Laribacter hongkongensis ampR 0.903 ampD 0.799

LPC\_1045 0.813

Legionella pneumophila Lens lpl1405 0.814 lpp1588 0.804

lpg1618 0.81

MSMEG\_2097 0.899

ampR 0.978

ampD 0.947

ampR 0.978

ampD 0.946 PA0305 0.899 PA14\_72760 0.853 PA3047 0.851 ampR 0.982 ampD 0.982 dacB 0.865 PSPA7\_6316 0.831 PSPA7\_0984 0.982 PSPA7\_6052 0.823 Bcenmc03\_0563 0.763 PFLU3466 0.801

ampR 0.989

ampD 0.951

scores

Table 1 continued



## Table 1 continued







used for visualizing, modeling, and analyzing molecular & genetic interaction networks. It supports several algorithms for the layout of networks such as spring-embedded layout, hierarchical layout, and circular layout. The large network can be visualized using Cytoscape version 2.8.3 [\[39](#page-13-0)].

## Results

## Network Analysis of ampC Gene Using STRING

This network analysis on *ampC* gene provides a clear view on the mechanism of the functional genes in  $\beta$ -lactamase induction. The association of  $ampC$  gene with other functional genes/proteins is analyzed using STRING tool. Taking only the highest  $(0.9-1.0)$  and high  $(0.7-0.9)$  confidence score values into consideration, 15 different bacterial genus and 247 functional genes/proteins are filtered. Hence, we preferred these organisms for further analysis.

The results reveal that, among the functional genes, 21.9 % (54) share the highest confidence score and 78.1 %





(193) share high confidence score. Among them, 65.1 % (161) genes are directly interacting, while 34.8 % (86) functional partners are indirectly interacting (sub network) with *ampC* gene by 34 interconnecting genes. Around 15 bacterial genus (human and nonhuman pathogens) included in network are Escherichia, Legionella, Pseudomonas, Rickettsia, Salmonella, Shewanella, Shigella, Yersinia, Acinetobacter, Aeromonas, Burkholderia, Mycobacterium, Vibrio, Enterobacter, and Laribacter. It is identified that out of 247 functional partners, 49 (19.8 %) are for ampD, 33 (13.3 %) are for frdD, 21(8.5 %) are for gcvA, 6 (2.4 %) for  $ampR$ , and 138 (55.7 %) for other genes, respectively.

The results are listed in Table [1](#page-2-0), 2 and Supplementary Table 1 which also contains descriptions of the functional partners collected from UniProt and National Center for Biotechnology Information (NCBI). Graphical model of ampC gene network and overall percentage of the functional partners are represented in Figs. [1](#page-6-0) and [2](#page-7-0).

Pathway Enrichment Analysis for Functional Partners

KEGG pathway enrichment analysis is carried out for all 247 functional partners. The results indicate that the most of genes and gene products shared a common

<span id="page-6-0"></span>

Fig. 1 a A graphical illustration of gene network represents the interaction between the target gene and functional partners (gene/ protein). b [Enlarge view] Hexagon shape in center indicates the AmpC gene (target gene) and interacting functional partners (color

nodes). Interconnecting genes (sub network) are represented in triangle shape. Genes and gene products involved in GO terms (blue color diamond shape nodes) and KEGG pathways (yellow color round rectangle shape nodes) are highlighted (Color figure online)

two-component system pathway [signal transduction systems (reference pathway ko02020)]. A majority of associated genes such as ampD, frdD, gcvA, and ampR are involved in cell wall recyclic pathway.

The next significant pathway is penicillin and cephalosporin biosynthesis pathway (reference pathway ko00311). From penicillin and cephalosporin biosynthesis pathway, it is found that the 30 functional partners are involved in the two reactions i.e., K01467 and K01434 (ko00311), respectively. 24 out of 30 genes are involved in  $\beta$ -lactamase synthesis (K01467) which is responsible for  $\beta$ -lactam resistance. The remaining 6 genes are involved in penicillin amidases synthesis (K01434). The list of functional partners is provided in Table [3](#page-7-0). The functional partners are highlighted (yellow color round rectangle shape) in Fig. 1. Cell wall recyclic pathway and penicillin & cephalosporin biosynthesis pathway are represented in Figs. [3](#page-8-0) and [4](#page-9-0).

## GO Enrichment Analysis for Functional Partners

GO enrichment analysis is performed using STRING tool and UniProt database. 607 GO terms which are involved in cellular component, biological, and molecular function of 247 functional partners are collected. Out of 607 GO terms, 264 are in molecular function, 262 are in biological process, and 81 are in cellular component, respectively. However, for 21 functional partners, GO terms are unavailable. These functional partners are not included for further analysis.

Among all GO terms, the functional genes (264) which are involved in molecular function are 61 are for N-acetylmuramoyl-L-alanine amidase activity [GO:0008745], 29 are for sequence-specific DNA-binding transcription factor activity [GO:0003700], 28 are for beta-lactamase activity [GO:0008800], 25 are for DNA-binding [GO:0003677], 17 are for penicillin-binding [GO:0008658], 7 are for penicillin amidase activity [GO:0008953], 3 are for hydrolase activity, acting on carbon–nitrogen (but not peptide) bonds, in linear amides [GO:0016811], and 94 are for other molecular processes, respectively.

The functional genes (262) which are involved in biological process are as follows: 51 are for peptidoglycan catabolic process [GO:0009253], 35 are for transcription, DNA-dependent [GO:0006351], 33 are for fumarate metabolic process [GO:0006106], 17 are for antibiotic catabolic process [GO:0017001], 11 are for response to antibiotic [GO:0046677], 9 are for beta-lactam antibiotic catabolic process [GO:0030655], and 106 are for other biological processes, respectively.

<span id="page-7-0"></span>

Fig. 2 Pie chart represents the overall percentage of the functional partners

Furthermore, the functional genes which are involved in cellular component are as follows: 30 are for integral to membrane [GO:0016021], 30 are for plasma membrane [GO:0005886], 13 are for cytoplasm, and 8 are for other cellular components, respectively.

The p value for the GO terms is obtained from STRING dataset. 94 GO terms have statistically significant values (p value  $\leq 0.05$ ). The significant GO terms (94) are then compared with complete GO term list (607). Interestingly, it is observed that 8 sets of GO terms have significant p value, remaining 7 sets of GO terms are found to be

insignificant (*p* value  $> 0.05$ ). In spite of statistical criteria of insignificance, by taking GO terms also as one of the criteria, they are considered as functionally important in  $\beta$ lactamase synthesis.

Based on p value  $( $0.05$ ), the percentage for the pre$ dominant functions in the network is calculated which is in the order of 29.8 % for peptidoglycan catabolic process [GO:0009253], 28.9 % for N-acetylmuramoyl-L-alanine amidase activity [GO:0008745], 18.1 % for beta-lactamase activity [GO:0008800], 11.7 % for antibiotic catabolic process [GO:0017001], 8.5 % for integral to membrane [GO:0016021], and 3.1 % for oxidoreductase activity [GO:0016491]. Apart from the above significant percentage values, other processes like DNA-binding [GO:0003677], transferase activity [GO:0016740], aromatic amino acid family metabolic process [GO:0009072], plasma membrane [GO:0005886], sequence-specific DNAbinding transcription factor activity [GO:0003700], carbohydrate metabolic process [GO:0005975], anaerobic respiration [GO:0009061], fermentation [GO: 0006113], electron carrier activity [GO:0009055], and hydrolase activity, acting on carbon–nitrogen (but not peptide) bonds, in linear amides [GO:0016811] are also involved in minimal percentages.

The results reveal that more than one function is shared by most of the associated genes. The associated genes in network are highly involved in regulating the expression of AmpC  $\beta$ -lactamase via, the three processes cellular component, biological process, and molecular function. List of annotated functional partners with significant  $(p \text{ value})$ 

Table 3 List of functional partners involves in penicillin and cephalosporin biosynthesis pathway

| Reaction<br>involved             | KEGG pathway ID   | Functional partners  | Organisms   |
|----------------------------------|---|--|---|
| K01467<br>Beta-<br>lactamase     | bac00311, spc00311, shw00311,<br>shn00311, shm00311, she00311,<br>sbp00311, psb00311, sbn00311,<br>sbm00311, sbl00311, sbl00311,<br>aha00311, pau00311, pap00311,<br>pap00311, bam00311, aci00311,<br>bmu00311, pmy00311, bcj00311,<br>ppf00311, bcm00311, bmu00311 | BamMC406_4877, Sputcn32_3157,<br>Sputw3181_0786, Shewana3_3440,<br>Shewmr7 3328, Shewmr4 0694,<br>Sbal223_0838, Psyr_3364,<br>Sbal195 0845, Shew185 0813,<br>Sbal_3522, Sbal_2086, AHA_3135,<br>PA14 72760, PSPA7 6316,<br>PSPA7_0984,<br>Bamb_6103, Bmul_3689, Bmul_6008,<br>Pmen 2038, BCAM0393,<br>Bcenmc03_4137, ACIAD3597,<br>Pput_2813 | Burkholderia ambifaria MC40,<br>Shewanella putrefaciens, Shewanella<br>sp. W3181, Shewanella sp. ANA3,<br>Shewanella sp. MR7, Shewanella sp.<br>MR4, Shewanella baltica OS223,<br>Shewanella baltica OS19, Shewanella<br>baltica OS185, Shewanella baltica<br>OS155, Pseudomonas syringae B728a,<br>Aeromonas hydrophila, Pseudomonas<br>aeruginosa PA7, Burkholderia<br>ambifaria AMMD, Burkholderia<br>multivorans 17616, Pseudomonas<br>mendocina, Burkholderia cenocepacia<br>MC03, Burkholderia cenocepacia<br>J2315, Burkholderia cenocepacia<br>MC03 and Pseudomonas putida F1 |
| K01434<br>Penicillin<br>amidases | acb00311, aby00311, sbo00311,<br>sbc00311, abn00311, abb00311   | A1S_1851, ABAYE1713, SBO_4393,<br>AB57_2186 SbBS512_E4877,<br><b>ABBFA 001601</b>  | Acinetobacter baumannii 17978,<br>Acinetobacter baumannii AYE, Shigella<br>boydii Sb227, Shigella boydii 3083,<br>Acinetobacter baumannii AB005 and<br>Acinetobacter baumannii  |

<span id="page-8-0"></span>

Fig. 3 AmpC beta-lactamase induction. a In cell wall recycling pathway, peptidoglycan releases the murein and get degraded into GlcNAc-anhMurNAc-tripeptide in periplasmic space. AmpG permease transports the muropeptide, GlcNAc-anhMurNAc-tripeptide into cytoplasm. Cytosolic enzyme N-b-acetylglucosaminidase cleaves GlcNAc-anhMurNAc-tripeptide into GlcNAc and anhMurNAc-tripeptide. AmpD produces amidases that hydrolyze the anhMurNAc-

 $(\leq 0.05)$ ) and insignificant GO terms is depicted in Table [4.](#page-10-0) The functional partners are highlighted (blue color diamond shape) in Fig. [1.](#page-6-0)

### Multiple Sequence Analysis

Around 61 AmpC protein sequences of Acinetobacter sp., Burkholderia cenocepacia, Burkholderia xenovorans, Enterobacter sp. 638, E. coli, E. fergusonii, P. aeruginosa, P. entomophila, P. putida, R. felis, Salmonella enterica Choleraesuis, Shewanella sp., Shigella sp., Legionella pneumophila Corby, Legionella pneumophila Philadelphia, and Laribacter hongkongensis are retrieved from UniProt for the analysis. MEGA software is used for multiple alignments to explore the associated features among AmpC proteins at sequence level. Motif search tool [PROSITE pattern and ProDom] is used to scan the pattern in AmpC  $\beta$ -lactamase. The characteristic motif for class C

tripeptide into anhMurNAc and tripeptide. Then, tripeptide re-enter into the murein biosynthesis. **b** Mutation in  $ampD$  leads to accumulate large amounts of anhMurNAc-tripeptide in cytoplasm and it acts as signaling molecules for transcriptional regulator AmpR which in turn triggers the beta-lactamase production. The muropeptides can induce the beta-lactamase expression by binding with the regulator AmpR

hydrolase beta-lactamase is found in region having residues 1–58, b-lactamase class-C active site is found in region having residues 91–98, and beta-lactamase hydrolase cephalosporinase precursor signal plasmid porin is found in region having residues 238–390, respectively. These functionally conserved residues as well as probable substitutions are analyzed from multiple alignments.

#### **Discussion**

Network Analysis of ampC Gene Using STRING

From the STRING results, based on combined confidence score, 247 associated genes from 15 different genus sharing interaction with ampC gene are selected. STRING utilizes an exclusive scoring framework. The scoring system is based on benchmarks of different types of associations

<span id="page-9-0"></span>

Fig. 4 Penicillin and cephalosporin biosynthesis pathway (KEGG:ko00311) is visualized in cytoscape. The functional partners (genes) involved in beta-lactamase (ko1467) and penicillin amidases (ko1434) synthesis reaction are represented in eclipse shape

against a common reference set to produce a single confidence score per prediction [\[40](#page-14-0)]. In this network, both target gene and their functional partners are represented as "nodes". They are connected by "colored edges" such as blue for co-occurrence, black for co-expression, deep pink for experiments, sky blue for databases, and green for text mining  $[21-27]$ . The probability of interaction between any two nodes indicates the strength of their functional relationship. It also reveals the possibility of genes to operate in the same or similar pathways. In addition to 30 genes which are identified from KEGG database (KEGG ko00311), GO for each functional partner (both human pathogen and nonhuman pathogenic bacteria) is extracted. Most of these associated genes or proteins in the network are directly or indirectly involved in the  $\beta$ -lactamase induction by regulating the expression of ampC gene. Overall results suggest that 19.8 % ampD, 13.3 % frdD, 8.5 % gcvA, 2.4 % of ampR, and 55.7 % functional partners are highly associated with ampC gene and it is represented in Fig. [2.](#page-7-0)

The results also reveal that the *ampD* (19.8  $%$ ) gene is in close interaction with ampC gene particularly in E. coli, P. aeruginosa, Acinetobacter ADP1, Y. enterocolitica, and Shigella sp. This ampD gene plays a vital role as a negative regulator of AmpC expression [\[41–44](#page-14-0)], ampD encodes forNacetyl-anhydromuranmyl-L-alanine amidase [GO:0008745] and peptidoglycan catabolic process [GO:0009253] which are involved in peptidoglycan recyclic pathway. It also regulates the expression of AmpC  $\beta$ -lactamase [\[44](#page-14-0), [45](#page-14-0)].

There are four important steps involved in recyclic pathway. They are Step I: ampG encodes an inner membrane permease for GlcNAc-1,6-anhydromuropeptides, which are peptidoglycan catabolites. Step II: GlcNAc-1,6-anhydromuropeptides are transformed to 1,6 anhydromuropeptides by NagZ. Step III: Interaction of 1,6-anhydromuropeptide with the LysR-type transcriptional regulator AmpR, which induces ampC gene to synthesis  $\beta$ -lactamase. Step IV: 1,6-anhydromuropeptides are processed by the N-acetyl-anhydromuramyl-L-alanine amidase ampD which blocks the AmpC induction [\[46–48](#page-14-0)]. Moreover, the mutations within the structural gene of ampD can lead to overproduction of AmpC  $\beta$ -lactamase [\[48](#page-14-0), [49](#page-14-0)] and accumulation of anhydromuramyl pentapeptide, which acts as signal for  $\beta$ -lactamase induction [\[48](#page-14-0), [50](#page-14-0)].

Subsequently, 2.4 % of ampR (Transcriptional regulator AmpR) and 8.5 % of gcvA regulate AmpC expression (transcription, DNA-dependent [GO:0006351]). This acts as a transcriptional regulator and a positive regulator for gene expression of  $\beta$ -lactamase (ampC). The ampR gene is located adjacent to ampC gene. It is divergently transcribed in C. freundii and E. cloacae, and P. aeruginosa [\[5](#page-13-0)]. This AmpR protein activates transcription by binding directly to the upstream promoter region of the ampC DNA. Certain studies also suggest that AmpR induces ampC by binding to anhydro N-acetylmuramyl peptides, which are cytosolic catabolites of peptidoglycan that gets accumulated when exposed to  $\beta$ -lactam antibiotics [[51\]](#page-14-0). In addition, gcvA gene also binds with the *ampR*-binding region of *ampC* and

<span id="page-10-0"></span>

IS insignificant, total count total number of gene appearance with the same annotations

activates transcription. Thus, gcvA mimics the activated state of ampR and provides a cross-talk between DNAbinding proteins of different inducible enzyme systems [\[52](#page-14-0)].

Another significant interaction is 13.3 % of frdD (fumarate reductase subunit D) which is involved in anchoring the catalytic components [GO:0005886] and fumarate metabolic process [GO:0006106]. Studies suggest

that fumarate reductase (frd) operon, in  $C$ , *freundii* is located next to ampC gene which is separated by 1100 base pairs  $[53]$  $[53]$ . In another study, promoter for the *ampC* gene is located within the last gene of the fumarate reductase (frd) operon in  $E$ . *coli*, and *amp* $C$  attenuator serves as the terminator for transcription of this preceding operon [\[54](#page-14-0)]. Although its locus is near to  $ampC$ , induction of  $\beta$ -lactamase by this gene still remains unclear.

These five genes ampC, ampR, ampD, frdD, and gcvA regulate the level of transcription both in the presence and absence of  $\beta$ -lactamase inducers. The inducers are the  $\beta$ -lactam antibiotics like cefoxitin and imipenem [\[15](#page-13-0)].

The remaining 55.7 % of functional partners are gene products, which exhibit various functions in regulating ampC gene such as N-acetylmuramyl-L-alanine amidase, negative regulator of AmpC, AmpD, beta-lactamase TEM precursor, beta-lactamase, transcriptional factor, LysR family transcriptional regulator, beta-lactamase expression regulator ampD, negative regulator of beta-lactamase expression, transcriptional regulator ampR, peptidase S45, and penicillin amidases that are listed in Table [1](#page-2-0) and Supplementary Table 1.

Collectively, the results from STRING conclude that these associated gene and gene products are involved in triggering the ampC gene expressions and in overproduction of  $\beta$ -lactamase resulting in increased resistance to  $\beta$ lactam antibiotics.

Pathway Enrichment Analysis for Functional Partners

Pathway enrichment analysis reveals that these functional partners are involved in major pathways such as two-component pathway, peptidoglycan recyclic pathway, and penicillin & cephalosporin biosynthesis pathway (KEGG ko00311) (Figs. [3](#page-8-0), [4\)](#page-9-0). In various bacterial species, beta-lactam resistance is regulated by two-component system. Twocomponent pathway comprises signal transduction that enables the adaptability in response to environmental changes [\[55](#page-14-0)]. It contains two signal transducers, sensor protein-histidine kinase and the response regulator. The response regulator regulates the gene expression and cellular physiology [[56](#page-14-0)]. Thus, two-component pathway permits the cells to sense and respond by inducing changes in transcription. Studies suggest that overexpression of the response regulators confers resistance to number of chemical compounds and many antibiotics [\[57](#page-14-0)]. From the results, it is observed that the functional partners (bla, EFER\_1643, PA14\_72760, blaOXA-50f, LPC\_1045, lpg1618, tem-1, Shew185\_0813, Sbal195\_0845, Sbal223\_0838, Sputcn32\_3157, Shewana3\_3440, Shewmr4\_0694, cphA, ampH, Bmul\_6008, Bmul\_3689, Bphyt\_4009, lpl1405, lpp1588, blaC, PSPA7\_6316, Sputw3181\_0786, Shewmr7\_3328, blaA, oxa-51, and bla-TEM) are involved in two-component pathway.

Since, peptidoglycan recyclic pathway in bacteria involves in recycling a significant proportion of the peptidoglycan components of their cell walls, which maintains cell integrity by sustaining internal osmotic pressure and keeps the regular bacterial shape. The recent researches clearly depict that there remains a direct linkage between beta-lactamase induction and cell wall metabolism among Gram-negative bacteria. In specific, muropeptides that are released during peptidoglycan recyclic pathway induce the expression of  $\beta$  -lactamase and hence this recycling pathway may serve as a signaling vehicle, which can be used as an novel target site for developing new antibacterial drugs or in supplementing the existing therapies [\[58](#page-14-0), [59\]](#page-14-0).The majority of functional partners namely ampD, ampR genes, and their gene products are associated with the cell wall recyclic pathway (Fig. [3\)](#page-8-0) [[48\]](#page-14-0).

KEGG database results suggest that 30 functional partners are involved in penicillin & cephalosporin biosynthesis pathway (Fig. [4](#page-9-0)). Among them, 24 functional partners are involved in  $\beta$ -lactamase (K01467) pathway. The analysis depicts that the associated genes stimulate beta-lactamase synthesis which in turn acts on penicillin. In the same pathway, six more genes that are involved in penicillin amidases synthesis (K01434) break the side chain of penicillin and 6-aminopenicillanate is liberated as an end product. This indicates that these reactions are involved in cleavage of  $\beta$ -lactam antibiotics. Since, this penicillin and cephalosporin biosynthesis pathway provides an overview of the diversity of ways that organisms can biosynthesize  $\beta$ -lactam antibiotics [\[28](#page-13-0), [60](#page-14-0)].

## GO Enrichment Analysis for Functional Partners

Enrichment analysis is carried out to gain insights to the functional roles for identified functional partners and to highlight their functional mechanisms at the network level. The enriched GO terms are evaluated specifically for the biological process. The GO terms are extracted from UniProt  $[29-31]$  with significant p value.

From the results, several functional genes or proteins having multiple functions are revealed. Among which, seven functional partners (Sputcn32\_3157, Shewana3\_3440, Shewmr4\_0694, Shewmr7\_3328, lpp1588, Sbal\_3522, and ampH) are involved in all the three functions namely antibiotic catabolic process [GO:0017001], penicillin-binding [GO:0008658], and beta-lactamase activity [GO:0008800].

Apart from these, ten more associated genes are involved in antibiotic catabolic process [GO:0017001], twenty one more functional partners are involved in beta-lactamase activity [GO:0008800], ten functional partners are involved in penicillin-binding [GO:0008658], eight functional partners are involved in beta-lactam antibiotic catabolic process [GO:0030655], and eleven functional partners are involved <span id="page-12-0"></span>in response to antibiotic [GO:0046677] (whereas seven functional partners are common in beta-lactam antibiotic catabolic process and response to antibiotic function).

Three genes (ampD, SBO\_4393, and A1S\_1851) are involved in hydrolase activity, acting on carbon–nitrogen (but not peptide) bonds, in linear amides. Seven genes (SbBS512\_E4877, AB57\_2186, ABBFA\_001601, ABAYE1713, Pmen\_1414, Pput\_1147, and Bcep18194\_ B0916) are involved in penicillin amidase activity [GO: 0008953] (Table [4](#page-10-0); Fig. [1](#page-6-0)).

Functional analysis reveals that GO terms of biological and molecular processes are highly represented in ampC gene interaction network. Particularly, the functional partner's occurrences within selected organisms are more variable. These predictions suggest that these proteins may be indirectly involved in  $\beta$ -lactamase stimulation. A summary of the biological, molecular, and cellular functions of these genes or proteins indicate the functional importance in the organisms. Therefore, the constructed gene or protein network would be helpful in discovering the functional association among the associated proteins [\[61](#page-14-0)]. The GO and pathway-based association analysis allow us to expand the knowledge about association of the gene and the mechanisms of network.

## Multiple Sequence Analysis

The PROSITE pattern represents ''[FY] - E - [LIVM] - G - S - [LIVMG] - [SA] - K" which indicates  $\beta$ -lactamase class-C active site (residues 91–98). ProDom suggested two motifs, first motif belongs to class C hydrolase betalactamase (residues 1–58) and second motif represents hydrolase cephalosporinase precursor signal plasmid porin (residues 238–390). Subsequently, we compared PROSITE pattern with the alignment results. The comparison results suggest that Phe (F) and Lys (K) residues are conversed in class C, and AmpC  $\beta$ -lactamase, Lys (K), and Ser (S) seem to play vital role in orientating the active site by electrostatic interaction. Electrostatic interaction between S and K creates a net positive potential in the catalytic site and predicated as the binding site of  $\beta$ -lactam antibiotics [\[62](#page-14-0)].

MSA reveals that all strains of E. coli, Shigella sp., Salmonella sp., and *Enterobacter* sp. strains share the highly conserved sequences and play an important role in the activity of AmpC  $\beta$ -lactamases. The residues that vary from one another in sequences are also identified in strain of Acinetobacter sp., Burkholderia sp., Shewanella sp., Rickettsia sp., Pseudomonas sp., Legionella spp., and Vibrio sp. The results suggest that the amino acid substitution occurred in the active site regions. Residues in the active sites regions are either structurally or functionally important for the  $\beta$ -lactamase activity.

In nutshell, AmpC b-lactamase sequences exhibit variability and only a few residues are conserved. These motifs represent characteristic significance of b-lactamases and can be further developed for better diagnostics and therapy.

# Conclusion

This study provides comprehensive evidence on  $ampC$ gene and their functionally associated genes in inducing  $\beta$ lactamase synthesis. The generated  $ampC$  gene networks emphasize that  $\beta$ -lactamase induction is accompanied by various regulatory genes. Thus, our study overviewed on the identified genes/proteins from various organisms and their role in regulatory mechanism of  $\beta$ -lactamase induction via biological process, molecular function, cellular process, pathway, and text mining. This constructed gene network provides critical information about functional relationships among biological pathways. It also provides information on the diverse biological process which includes gene functions and complex cellular mechanisms of  $\beta$ -lactamase induction. This would help in better understanding the functions of association partners and their impact on  $\beta$ -lactamase induction. The multiple sequence alignment of AmpC proteins will be useful to study the amino acid variation. To conclude, this study might provide new dimension to understand the regulatory mechanisms of  $\beta$ -lactamases production thereby discovering the inhibitors targeting beta-lactamase induction pathway to prevent the emerging of beta-lactam resistance and improve the efficacy of clinical beta-lactam antibiotics. This study provides useful information for researchers exploring in the field of  $\beta$ -lactamase-mediated antibiotic resistance and also for researchers in drug discovery and development.

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