Analysis of Transcriptome datasets of Jatropha Curcas Leaves Infected with Colletotrichum Gloeosporioides Using RNA-Seq

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Abstract:- Even though jatropha holds enormous potential as a biodiesel feedstock, there are still significant obstacles that can be overcome before it can be economically viable in many parts of the world. Jatropha curcas develops anthracnose due to Colletotrichumgloeosporioides and it affects each genotype in a different way. A gene-expression profiling study was performed with total RNA from two cultivars of Jatropha curcas that had and did not have a Colletotrichumgloeosporioides inoculation line in order to identify potential candidate genes involved in pathogenplant interactions. There were 213 significant differential gene expressions between the wild type 9-1 and RJ127 inoculated cultivar lines with Colletotrichumgloeosporioides. A comprehensive analysis of these genes provided information regarding the signaling system, hormone biosynthesis and regulation, transcription regulation, ubiquitin-mediated and proteolysis. Gene Ontology (GO:0006952) annotations were found for nine genes associated with defense response, six orthologous genes were associated with pathogen-plant interactions, and 27 transcription factors were associated with fungus responses. The expression levels of genes varied among cultivars when infected with Colletotrichumgloeosporioides, according to gene expression analysis. Transcriptome analyses were used to diagnose candidate genes for transcription factors and structural genes that may be involved in plant-pathogen interactions in Jatropha curcas leaves of the cultivated genotypes. These analyses provided new insights into the defense mechanisms controlling infection associated with cultivar Jatropha curcas 9-1. The structural genes and regulators identified here can also be used as efficient genetic markers for the selection of cultivars with excessive defense mechanisms.

Keywords:-*Jatropha*; *C. gloeosporioides*; *NGS*; *RNA-Seq*; *DGEs*; *biodiesel*; NGS;

I. INTRODUCTION

Biofuels existed long before vehicles were invented. The relatively modest gas and diesel prices have kept them at the periphery for quite some time. The increase in oil prices, as well as worldwide initiatives to combat the negative effects of environmental change, have lent fresh urgency to the search for spotless, sustainable [1]. Jatropha curcas (J. curcas) is a species endemic to Mexico and Central America and a member of the Euphorbiaceae family [2]. The tropical plant has extraordinary potential to produce biofuel for energy crisis mitigation, environmental management, and sustainable production management in the future [3]. A moderately small level genome ~320.5 MB for the RefSeq version of J.Curcas, available in the NCBI genome J.curcas Annotation Release 101, which syncs with the J. Curcas database (JCDB) and contains gene functional annotation, interaction networks, and expression matrices for use in J. Curcas functional genomics research [4][5][6][7][8][9][10]. A case of anthracnose was reported on the J.Curcas plant in 2010 and 2011, which was the first known case of the disease caused Colletotrichumgloeosporioides by (C. gloeosporioides) on J.Curcas in Korea [11]. C. gloeosporioides causes anthracnose, a serious Curcas disease, causing damage to leaves, stems, and fruits, which adversely affects the production and quality of seeds. Therefore, there is a limitation to the improvement of the biofuel industry due to parasitic diseases [12].

Improvements in innovation and strategies have allowed fast and generally reasonable RNA-Seq dataset generation over the decades [13]. As a result, RNA-Seq in various model frameworks has become an undeniably well-known strategy for transcriptome-wide changes. The development of information examination has propelled improvements in open access programs for quality control, mapping, and differential expression (DE) investigation [14]. NCBI SRA database contains raw data that can be downloaded from the NCBI distributed research archive, which gradually transfers distributed information to openly accessible files. Researchers can now examine RNA-Seq information with fewer obstacles, allowing them to make and test theories with RNA-Seq data that have progressively different logical foundations.

The present study compares the transcriptomic profiles of four Jatropha samples from the NCBI BioProjects PRJNA254929 and PRJNA254930 data using the Illumina HiSeq 2500 platform to identify differentially expressed genes related to resistance and susceptibility of Jatropha to C.gloeosporioides. Based on these results, we can identify some candidate genes for the resistance of two different Jatropha genotypes against C.gloeosporioides, and gain insight into the mechanism of Anthracnose disease on the two genotypes of J. Curcas.

II. MATERIALS AND METHODS

A. Raw Data Retrieval From Rna Sequencing

This study used data from NCBI BioProject PRJNA254929 and PRJNA254930. To investigate differentially expressed genes, fresh leaf samples of J. curcas strains RJ127 and Cultivar 9-1 had been artificially infected or not with C. gloeosporioides. Following the manufacturer's recommended protocol, RNA was extracted from control, induced tissue of susceptible and tolerant lines after 2hr, 24hr, 96hr, and 144hr of infection. To retrieve data from NCBI, we used SRAToolsKit 2.10.4.

B. Raw Data Processing And Mapping Of Reads To The Reference Genome

To analyze the raw reads, we preprocessed them in fastq format before analyzing them further. Trimming the raw fastq reads was automated using the Trim Galore, which is a wrapper tool around FastQC and Cutadapt to consistently apply quality and adapter trimming. Default settings include trimming all paired-end reads with a Phred score of less than 20, trimming only paired-end reads, and introducing an Illumina adaptor sequence for any sequence with a 5 base pair overlap. This filtered high-quality read data was used for all downstream analyses. STAR aligner with default settings was used to align trimmed reads to J. curcas genome (assembly JatCur_1.0).

C. Identification And Annotation of Differentially Expressed Genes (Degs)

Gene model GTF files were generated from combined transcriptomic sequences by Cufflinks. Each transcript's FPKM value was calculated across all samples. For each sample, we used Cuffmerge to obtain a final GTF file. Using Cuffdiff, we calculated the differential abundance of transcripts among different samples/sites. In order to identify the significance of an expression difference, we used p-values of ≤ 0.05 and log2Ratios greater than 1. Our analysis of significant genes with Gene Ontology (GO) uses Uniprot database after obtaining locusIDs from Cuffdiff result with uniprotIDs from the bioDBnet server.

III. RESULTS

A. Transcription Profiles Of Resistant And Susceptible Cultivars Of J. Curcus Infected With C. Gloeosporioides

In this study, two cultivars of J. curcus reacted contrastingly to the pathogen C. gloeosporioides. Four libraries were developed from C. gloeosporioides infected leaf tissues and non-inoculated leaf tissues to determine which cultivars are receptive to C. gloeosporioides and to design quality articulation designs (Table 1).

Abbreviation	Tissue	Condition	Description
Pool A	Leaf	9-1_Control (resistant)	A Control was formed by pooling uninfected genotypes of J. Curcas and
			tissue from tolerant lines 2, 24, 96, and 144 hours after infection.
Pool B	Leaf	9-1 induced (resistant)	The induced tissue of tolerant lines collected after 2, 24, 96, and 144 hours
			after infection was pooled as B induced with J. Curcas genotypes showing
			tolerance to C. gloeosporioides.
Pool C	Leaf	RJ127 control	A control group was made up of uninfected J Curcas genotypes, and tissue of
		(susceptible)	susceptible lines was collected after 2, 24, 96, and 144 hours of infection.
Pool D	Leaf	RJ127 induced	The induced tissue of tolerant lines collected after 2, 24, 96, and 144 hours
		(susceptible)	after infection were pooled as D induced with J. Curcas genotypes showing
			tolerance to C. gloeosporioides.

TABLE 1. CONTROLS AND SAMPLES INFORMATION OF J. CURCUS CULTIVARS

From total RNA isolates, cDNA libraries were synthesized to analyze the transcriptome of J. curcas. Illumina sequencing platforms were subsequently used to sequence cDNA libraries. Following the filtering and trimming of low-quality reads and adaptors, clean reads were mapped against the NCBI genome JatCur_1.0. Over 90% of the reads were uniquely mapped. The raw data and the QC processed data are available in table 2. Short reads assembling software Star aligner was used to assemble these clean reads. The FPKM method was used to calculate the expression levels of 25,515 genes based on the total mapped reads. An indication of significant-quality from cuffdiff yields based on p values and FDRs after Benjamini-Hochberg correction was used to identify response against C. gloeosporioide.

Samples	Raw reads	Raw bases	Clean Reads	Clean Bases	Valid ratio (base)	GC content (%)	Uniquely mapped
							(%)
Pool A	45280396	4573319996	44402527	4377837709	95.7%	43%	92.89%
Pool B	37461637	3783625337	36842449	3632603350	96.0%	43%	92.40%
Pool C	51169379	5168107279	49898983	5003568089	96.8%	43%	94.12%
Pool D	38496992	3888196192	37317006	3678977234	94.6%	43%	91.04%

B. Dges Profiles In Response To C. Gloeosporioides Infection

> Profile Of Gene Expression In Control Cultivars Versus C. Gloeosporioides Infected Cultivars

A total of 160 significant DGEs were identified between the resistance control the 9-1 cultivar and its inoculated libraries (PoolA vs PoolB). Out of 104 annotated genes, 101 were up-regulated while 3 were down-regulated. In the comparison between RJ127 and the inoculated libraries, 116 DGEs were identified out of 83 annotated DGEs, of which 77 were upregulated and 6 were downregulated (Table 3).

> Dge Profiles Of Both Cultivars In Control Samples

In the comparison between the resistant control and susceptible control, 94 significant DGEs were found out of 58 annotated DGEs, 46 of which were upregulated and 12 were downregulated in the RJ127 cultivar sample compared to the 9-1 cultivar sample(Table 3).

> Dges PROFILES OF BOTH CULTIVARS WITH C. GLOEOSPORIOIDES INFECTION

Based on the comparison of the inoculated 9-1 cultivar and the RJ127 inoculated cultivar, 88 significant DGEs were identified out of 63 annotated DGEs, in which there were 53 up-regulated genes and 10 down-regulated genes(Table 3). Overall, the number of DGEs was significantly higher in the 9-1 cultivar than RJ127 in Response to C. gloeosporioides Infection.

C. Gene Ontology (Go) Annotation

An outline of the utilitarian categories associated with infection-associated DGEs was developed using the Gene Ontology (GO) assignments within the UniProt database [*Fig. 1b*]. A total of 78 annotated DGEs associated with 9-1 and its inoculated libraries had been classified in the GO database, whereas 63 annotated DGEs associated with RJ127 and its inoculated libraries had been classified in the GO database. A venn diagram [*Fig. 1a*] that depicted the DGEs results from all 4 pools made it evident that unique as well as shared DGEs were distinguishable among, and between, sets.

Gene names	UniProt Accession	log2(fol d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)
JCGZ_1431 1	A0A067K8D2	-2.91851	Down				
	Q8VYU0	-1.71861	Down				
JCGZ_1540 1	A0A067LNK5	-1.62426	Down				
JCGZ_2134 8	A0A067JAT4	1.42747	UP				
JCGZ_0705 5	A0A067KNU6	1.44722	UP				
JCGZ_1750 4	A0A067JQW3	1.51161	UP				
WRKY57 JCGZ_0330 8	S5CFW3; S5CS97	1.54828	UP				
	C9E0E8	1.55778	UP				
JCGZ_1838 3	A0A067K4S0	1.56653	UP				
JCGZ_2575 2	A0A067JVT3	1.57314	UP				
JCGZ_1066 6	A0A067KG70	1.63609	UP				
JCGZ_2467 2	A0A067L0C9	1.6536	UP				
JCGZ_1891 1	A0A067JV40	1.65432	UP				
JCGZ_2332 4	A0A067JHR5	1.68004	UP				

TABLE 3. DGES IN DIFFERENT GROUPS

		log2(fol	Deculatio		Deculatio		Deculatio
Gene names	UniProt Accession	d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)
JCGZ_1674 4	A0A067L4V4	1.68064	UP				
JCGZ_1153 7	A0A067K4L2	1.70613	UP				
JCGZ_2628 3	A0A067JF33	1.74402	UP				
JCGZ_1604 1	A0A067LC07	1.74905	UP				
JCGZ_1190 1	A0A067KF78	1.84629	UP				
WRKY51 JCGZ_1435 3	S5CKC9	1.8522	UP				
JCGZ_0295 7	A0A067L1F8	1.85487	UP				
JCGZ_2520 8	A0A067L3L7	1.87466	UP				
JCGZ_2245 6	A0A067K1V9	1.88759	UP				
JCGZ_2685 0	A0A067L091	1.89331	UP				
JCGZ_0261 9	A0A067L678	1.93896	UP				
JCGZ_1327 2	A0A067K8C1	1.97321	UP				
JCGZ_0970 0	A0A067LAG1	1.97504	UP				
JCGZ_0205 0	A0A067KV70	1.99654	UP				
LOX2 JCGZ_1670 8	A0A067L4M8	2.00344	UP				
JCGZ_0874 5	A0A067KIP6	2.08499	UP				
JCGZ_0927 1	A0A067KIY5	2.10078	UP				
JCGZ_2577 3	A0A067JJH0	2.11164	UP				
WRKY47 JCGZ_1433 2	S5CKC4	2.11916	UP				
WRKY08 JCGZ_0040 0	M9TNQ5	2.20164	UP				
JCGZ_0743 9	A0A067KFV0	2.21122	UP				
JCGZ_0392 5	A0A067L6T7	2.23813	UP				
JCGZ_2488 2	A0A067L0X2	2.24577	UP				
JCGZ_0282 3	A0A067LCX8	2.27533	UP				
JCGZ_1940 0	A0A067JZK3	2.27653	UP				

		log2(fol							
Gene names	UniProt Accession	d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)		
JCGZ_0258 8	A0A067KX45	2.30864	UP						
JCGZ_0101 2	A0A067L449	2.31447	UP						
JCGZ_1965 6	A0A067JYJ8	2.3289	UP						
WRKY28 JCGZ_0709 1	S5CFS7	2.33565	UP						
JCGZ_0171 6	A0A067JTW8	2.34832	UP						
JCGZ_0870 3	A0A067KX47	2.3508	UP						
JCGZ_0932 8	A0A067KRX2	2.37089	UP						
JCGZ_1153 0	A0A067KH50	2.41079	UP						
JCGZ_0171 5	A0A067JJI5	2.42259	UP						
JCGZ_0796 8	A0A067LEP7	2.5398	UP						
JCGZ_0044 1	A0A067JTI6	2.62832	UP						
PAL4 JCGZ_0745 2	A0A067KCD4	2.71673	UP						
JCGZ_0281	A0A067LCW 8	2.92729	UP						
JCGZ_1247 4	A0A067KAG7	3.0316	UP						
JCGZ_0211 3	A0A067L6J1	3.16631	UP						
JCGZ_2495	A0A067KXS0	3.36643	UP						
JCGZ_1444 7	A0A067JXI1	3.38977	UP						
JCGZ_1501 6	A0A067L9M0	3.39243	UP						
JCGZ_1681 1	A0A067LH44	3.45701	UP						
JCGZ_2067 1	A0A067JZT2	3.57009	UP						
JCGZ_1516 3	A0A067LDL0	3.61511	UP						
JCGZ_1894 6	A0A067JV70	3.99563	UP						
JCGZ_0902 3	A0A067KTI1	4.2537	UP						
JCGZ_0757 6	A0A067KNY6	4.52988	UP						
JCGZ_2067 0	A0A067K136	4.77342	UP						
JCGZ_2468 9	A0A067L877	5.12327	UP						

Gene names	UniProt Accession	log2(fol d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)
JCGZ_1681 5	A0A067L8E6	5.39239	UP				
JCGZ_2063 3	A0A067JNL7	3.04133	UP			9.04184	Up
JCGZ_1894 2 JCGZ_1894 4	A0A067JVF0	3.73551	UP			4.00984	Up
JCGZ_2114 4	A0A067JD40	3.538	UP			3.70795	Up
JCGZ_0511 1	A0A067L2R2	2.27356	UP			3.53599	Up
JCGZ_2290 4	A0A067K243	2.95129	UP			3.12113	Up
JCGZ_2508	A0A067JKK3	2.70873	UP			2.85497	Up
CLA1 JCGZ_2301 2	A0A067LHX1	2.29141	UP			2.84838	Up
JCGZ_1247 6	A0A067KIB2	1.66776	UP			2.46842	Up
JCGZ_0061 5	A0A067JGC3	2.73234	UP			-2.77065	Down
JCGZ_0751 8	A0A067KCS8	1.83993	UP			-3.28346	Down
JCGZ_0408 7	A0A067L3V1	2.21339	UP			-3.86326	Down
JCGZ_1739 6	A0A067LMZ1	2.91394	UP	2.17882	Up	-2.69737	Down
JCGZ_1581 0	A0A067L2E4	2.11219	UP	2.45413	Up	-2.28623	Down
JCGZ_2141 8	A0A067JB04	1.49165	UP	1.55317	Up		
JCGZ_2384	A0A067LFL5	1.60033	UP	1.4277	Up		
WRKY07 JCGZ_1668 4	M9TGB8	1.74768	UP	1.61453	Up		
JCGZ_1689 4	A0A067L5A5	1.77871	UP	2.45938	Up		
JCGZ_1828 7	A0A067JZG6	1.78904	UP	1.76848	Up		
JCGZ_2605 0	A0A067JRT8	1.80388	UP	1.71348	Up		
JCGZ_2021 5	A0A067JTU3	1.81124	UP	1.64682	Up		
JCGZ_1631 5	A0A067L7T6	1.82742	UP	1.71468	Up		
JCGZ_1138 0	A0A067KGN0	1.84507	UP	1.61488	Up		
JCGZ_0131 3	A0A067LC73	1.84923	UP	1.82433	Up		
JCGZ_0877 2	A0A067KM96	1.87745	UP	1.68813	Up		

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Gene names	UniProt Accession	log2(fol d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)
JCGZ_2171 8	A0A067JBV0	1.91043	UP	1.71733	Up		
JCGZ_2057 2	A0A067JN47	2.01654	UP	1.51071	Up		
JCGZ_1587 2	A0A067LAD0	2.0847	UP	1.99865	Up		
JCGZ_2176 1	A0A067JMR1	2.15307	UP	2.17196	Up		
JCGZ_1580 9	A0A067KZ11	2.16436	UP	2.72198	Up		
JCGZ_0826 8	R4NHR7	2.30741	UP	1.92321	Up		
WRKY27 JCGZ_2699 0	S5CKF6	2.31891	UP	1.71705	Up		
JCGZ_0967 5	A0A067LAD6	2.32451	UP	2.38589	Up		
JCGZ_0345 1	A0A067L5Z5	2.35703	UP	1.87669	Up		
JCGZ_1580 8	A0A067KYX4	2.41251	UP	2.56576	Up		
GRAS17 JCGZ_0078 7	A0A067L3E6	2.53187	UP	1.85568	Up		
JCGZ_0875 0	A0A067KIQ1	3.10243	UP	2.36395	Up		
JCGZ_0480 2	A0A067L249	3.42307	UP	3.20016	Up		
WRKY54 JCGZ_2047 2	S5CKD4	4.05806	UP	3.42691	Up		
JCGZ_2125 0	A0A067JMQ3			-4.62991	Down		
JCGZ_0055 5	A0A067JG74			-4.54832	Down		
JCGZ_0696 7	A0A067L0K1			-3.55157	Down		
JCGZ_2398 0	A0A067JM79			-3.05669	Down		
JCGZ_0564 6	A0A067LIV1			-2.63614	Down		
DHAR JCGZ_0658 9	R9QAK3			-2.40991	Down		
JCGZ_2605 5	A0A067JRU1			-2.18077	Down		
JCGZ_2678 8	A0A067LBB3			-2.16355	Down		
JCGZ_2491 8	A0A067L9X5			-1.97304	Down		
JCGZ_0026 7	A0A067LD96			-1.86521	Down		
JCGZ_1670 9	A0A067L4S5			-1.6963	Down		

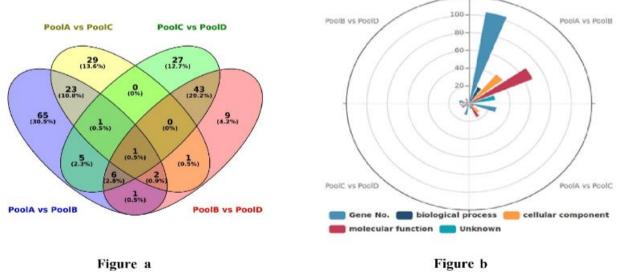
Gene names	UniProt Accession	log2(fol d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)
JCGZ_1655 7	A0A067K293			-1.5609	Down		
JCGZ_0332 0	A0A067JQ58			1.54494	Up		
JCGZ_0881 5	A0A067KIW7			1.69431	Up		
JCGZ_2029 8	A0A067K4Y8			1.73078	Up		
JCGZ_2560 1	A0A067JWD4			1.73781	Up		
JCGZ_1873 9	A0A067K1G2			1.75928	Up		
JCGZ_0622 5	A0A067KZA6			1.92552	Up		
JCGZ_1324 3	A0A067KBE6			1.95642	Up		
JCGZ_2679 5	A0A067L047			2.00783	Up		
JCGZ_1664	A0A067JZ78			2.2488	Up		
JCGZ_2245 7	A0A067JQK6			2.28603	Up		
JCGZ_1832 5	A0A067KC42			2.30001	Up		
JCGZ_0336 3	A0A067JD10			2.57497	Up		
JCGZ_2306	A0A067JH55			2.72935	Up		
JCGZ_1756 3	A0A067K2B0			2.76153	Up		
JCGZ_0254 0	A0A067L4T7			2.7958	Up		
JCGZ_2482 3	A0A067L9Q0			3.19224	Up		
JCGZ_0889 4	A0A067KK72			3.78448	Up		
JCGZ_0219 8	A0A067L6V1			4.03899	Up		
RHF	G9DD71					-2.88633	Down
JCGZ_2598 9	A0A067JH80					2.23663	Up
JCGZ_2134 4	A0A067JDN5					2.23891	Up
JCGZ_1209	A0A067KLX7					2.25554	Up
JCGZ_2610 4	A0A067JHM0					2.30378	Up
JCGZ_1896 7	A0A067JVH2					2.43708	Up
JCGZ_2602 0	A0A067JRP4					2.47485	Up
JCGZ_1147 8	A0A067K4N4					2.49398	Up

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Gene names	UniProt Accession	log2(fol d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)		
JCGZ_1017 3	A0A067LG90					2.50453	Up		
JCGZ_0066 3	R4N5K2					2.55145	Up		
JCGZ_2109 3	A0A067K1C8					2.58763	Up		
JCGZ_2110 3	A0A067K1D9					2.63856	Up		
JCGZ_0686 7	R4NFR5					2.67368	Up		
JCGZ_0777 9	A0A067KGW 9					2.69994	Up		
JHL06P13.1 6 JCGZ_0616 4	E6NUC5					2.70591	Up		
JCGZ_1547 7	A0A067LBY7					2.73333	Up		
JCGZ_0477 6	A0A067KT36					2.81096	Up		
JCGZ_1609 2	A0A067LB16					2.93859	Up		
JCGZ_0660	A0A067LCF8					2.97213	Up		
JCGZ_0781 5	A0A067KR22					2.98905	Up		
JCGZ_2632 4	A0A067JI47					3.09123	Up		
JCGZ_2176 6	A0A067JMR4					3.1424	Up		
JCGZ_0900 9	A0A067KH55					3.19027	Up		
JCGZ_0721	A0A067KMZ 2					3.20441	Up		
JCGZ_2687 9	A0A067L086					3.23976	Up		
JCGZ_0563 2	A0A067LHU5					3.34604	Up		
JCGZ_0119 5	A0A067LJ82					3.39571	Up		
JCGZ_2189 9	A0A067JF68					3.42101	Up		
SAM JCGZ_2247 9	K9UUN6					3.45222	Up		
JCGZ_0075	A0A067KS90					3.58836	Up		
JCGZ_1232 9	A0A067KA58					3.61842	Up		
JCGZ_0086 8	A0A067KSB6					3.62234	Up		
JCGZ_0463 2	A0A067L1K4					3.65871	Up		
-	I	1	1	1	1	1	1		

	log2(fol						
Gene names	UniProt Accession	d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)
JCGZ_2641 2	A0A067JF88					3.68785	Up
JCGZ_1867 6	E2CXJ0					3.77831	Up
JCGZ_0755 7	A0A067KCN2					3.86772	Up
JCGZ_1851 7	A0A067K4H1					3.88523	Up
JCGZ_2069 4	A0A067JRS2					4.00472	Up
JCGZ_0142 7	A0A067L934					4.03707	Up
JCGZ_0484 1	A0A067KTE0					4.06179	Up
JCGZ_2553 6	A0A067JX93					4.08249	Up
JCGZ_1364 4	A0A067KDF3					4.11712	Up
JCGZ_1563 4	R4N7L5					4.11958	Up
JCGZ_1583 6	A0A067LBD6					4.15701	Up
JCGZ_1011 0	A0A067LN77					4.2811	Up
WRKY35 JCGZ_2168 8	S5CH50					4.30072	Up
JCGZ_2444 8	A0A067JZ76					4.40172	Up
JCGZ_1138 9	A0A067K7J2					4.45745	Up
JCGZ_1028 4	A0A067LPI9					4.5021	Up
JCGZ_0215 9	A0A067KVC5					4.62141	Up
JCGZ_2050 7	A0A067JMZ0					4.62345	Up
JCGZ_2164 4	A0A067JEH9					4.84044	Up
JCGZ_2227 9	A0A067K3C6					4.86939	Up
JCGZ_1248 2	A0A067K6Z9					4.90456	Up
JCGZ_0348 2	A0A067KUR0					4.96825	Up
JCGZ_0714 2	A0A067KBH4					5.24689	Up
JCGZ_2236	A0A067L5S7					5.31491	Up
JCGZ_0077	A0A067L4J4					5.35103	Up
JCGZ_2125	A0A067JAJ5					5.40441	Up
	1		1	1	1		1

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Gene names	UniProt Accession	log2(fol d change) PoolA vs PoolB	Regulatio n (PoolA vs PoolB)	log2(fold_change) PoolA vs PoolC	Regulatio n (PoolA vs PoolC)	log2(fold_change) PoolC vs PoolD	Regulatio n (PoolC vs PoolD)
JCGZ_1176 2	A0A067K581					5.41423	Up
JCGZ_2508 1	A0A067JKU4					5.45967	Up
JCGZ_0104 4	A0A067KT39					5.91992	Up
JCGZ_0548 5	A0A067L9U2					6.05053	Up
JCGZ_2595 3	A0A067JE72					6.4195	Up
JCGZ_0740 2	A0A067KC92					6.64945	Up
JCGZ_2140 2	A0A067JAU2					6.7041	Up
JCGZ_0796 1 JCGZ_2101 9	A0A067LQR9					6.82357	Up
JCGZ_1828 9	D2D954					6.95559	Up
JCGZ_2482 7	A0A067L0T4					8.96372	Up
JCGZ_1455 1	A0A067K938					9.93983	Up



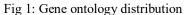


Fig 1: Gene ontology distribution: Gene ontology distributions for DGEs of J curcas related to plant physiology of the molecular function, cellular component, and biological process. Figure 1a: Venn Diagram of DGEs genes in all 4 pools. Figure 1b: Gene ontology distributions for DGEs of J curcas. A gene may be associated with several GO terms.

D. Go Terms For Defense Response Gene Ontology (Go:0006952)

Gene Ontology term GO:0006952 was assigned to nine genes for defense response. Five DGEs were annotated when 9-1 cultivar and its inoculated libraries were compared. When comparing RJ127 cultivar and its inoculated libraries, 1 DGE was annotated with GO:0006952. A total of two DGEs were annotated for both controls and 4 genes for both infected cultivars (*Fig. 2*).

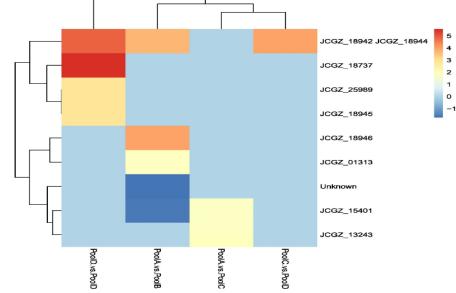


Fig. 2: DGEs of Gene Ontology [GO:0006952]:DGE distribution for GO:0006952 to understand the defense response against fungi.

E. Pathway Enrichment Analysis

Kyoto Encyclopedia of Genes and Genomes (KEGG) pathways were identified according to DGEs with a corrected p < 0.05. A total of 72 KEGG pathways were significantly enriched, of which 17 were enriched when 9-1 cultivar and its inoculated libraries were compared, 46 were enriched when RJ127 cultivar and its inoculated libraries were compared, while 33 were enriched in control samples for both cultivars were compared e.g. 9-1 cultivar compared with RJ127 cultivar without inoculation, while 31 were enriched both cultivars with C. gloeosporioides Infection i.e. 9-1 cultivar compared with RJ127 cultivar with inoculation were compared, while 5 are enriched all the time (*fig. 3*). DGEs

related to 'Plant metabolism' (KO01100) had the highest representation, followed by DGEs for 'Secondary metabolite biosynthesis' (KO01110), then those for 'Plant-pathogen interactions' (KO04626), 'Plant hormone signal transduction' (KO04075), 'Phenylpropanoid biosynthesis' (KO00940), and 'MAPK signaling pathway – plant' (KO04016).

These findings suggest that J curcus has adapted a range of molecular defenses that prevail in disease, and C gloeosporioides contamination causes predominant changes in numerous metabolic pathways. The process is clearly complex and is likely to be regulated by a number of pathways, such as the plant-pathogen interaction.

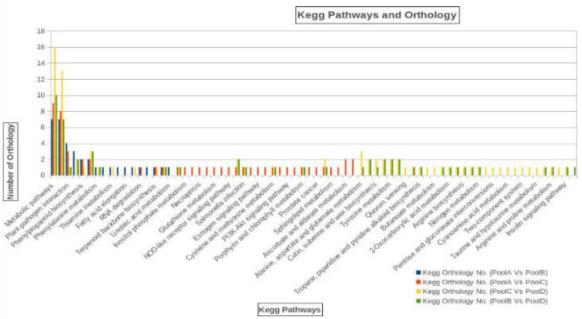


Fig. 3:KEGG pathways and Orthology

F. Dges Involved In Plant-Pathogen Interaction

Based on *fig. 4*, there were 4 gene orthology WRKY33, WRKY22, CML and KCS were significantly upregulated when 9-1 cultivar and its inoculated libraries were compared however CML is downregulated when RJ127 cultivar and its inoculated libraries were compared. The related KEGG map

id is KO 04626. When we were looking resistant control with susceptible control comparison, there WRKY33 and RBOH were significantly upregulated and HSP90A was downregulated. While we compared cultivars with C. gloeosporioides Infection CML is upregulated in 9-1 cultivar.

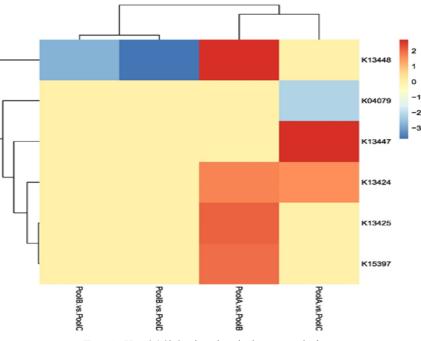


Fig. 4: KO:04626 related orthology regulation

G. Tfsrelated To C. Gloeosporioides Responses

TF families, such as WRKY (a conserved N-terminal sequence of WRKYGQK in conjunction with TF a Zn fingerlike motif.), NAC (a (N-acetyl cysteine), AP2/ERF (APETALA2/ethylene with a Zn finger-like motif.), bHLH (basic (homeodomain-leucine helix-loop-helix), HSF (heat shock transcription factor) play vital roles in regulating plant resistance mechanisms under abiotic stress.

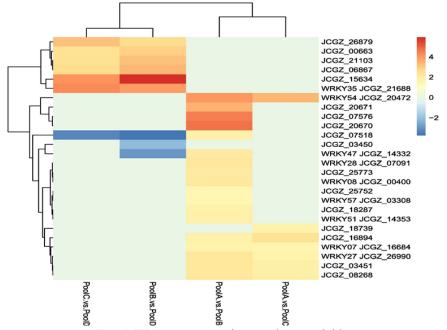


Fig. 5: TFsResponse Against C. gloeosporioides

In the present study, when 9-1 cultivar and its inoculated libraries were compared, 17 TFsfrom 4 families and one TF from an unknown family were detected. The RJ27 cultivar and its inoculated libraries showed seven TFsfrom four families. A comparison of resistant control with susceptible control revealed five TFsfrom three families and three from an unknown family. When inoculated libraries of the 9-1 cultivar were compared with inoculated libraries of the RJ127 cultivar, nine TFsfrom three families were identified. Figure 5 illustrates the expression of the related TFsin J.curcus induced by C.gloeosporioides.

IV. DISCUSSION

The pathogen Colletotrichum is one of the most important causes of anthracnose on trees, grasses, and a range of other types of plants[15]. There have been few studies of J.Curcus for disease resistance mechanisms against C **VI**. gloeosporioides that could provide insight into anthracnose [16]. In this study, gene expression levels were compared between two cultivars of J.Curcas leaves subjected to C.gloeosporioides infection and those grown without the pathogen. A total of 308 DGEs were identified.

In plants, inborn resistance is activated by means of the reaction of PRRs (pattern-recognition receptors) to PAMPs (pathogen-associated molecular patterns), along these lines giving the principal line of inducible pathogen-related molecular PTI (PAMP-triggered immunity), further it triggers MAPK (Mitogen-activated protein kinase) signaling pathway that activates defense genes for antimicrobial compounds. In our study MAPK signaling pathway genes, those responsible for early/ late response of pathogen infection and cell death due to pathogen attack were upregulated in 9-1 J. curcas cultivar. In RJ-127 cultivar genes related to stress adaption signaling and defense response activated by ethylene signaling were upregulated when both cultivars were infected by C.gloeosporioides. Expansion in the cytosolic Ca2+ fixation also acts as a controller for the creation of receptive species oxygen and limited customized cell death/hypersensitive reaction. Ca2+ fixation genes from plant-pathogen interaction induced stomatal closure were upregulated in inoculated 9-1 cultivar.

Plants can incite transcription factors to DNA groupings as a technique for changing in accordance with natural and abiotic stresses [17]. Pathogen-reaction qualities, for example, WRKY, assume urgent jobs in plant reactions. The WRKY family has been recognized in the plant realm, and diverse WRKY families are found in higher plants [18]. The transcript investigation uncovered that the WRKY transcript gets upregulated because of saltiness, parchedness, salicylic acid (SA), methyl jasmonate (MeJa), and the collar rot fungus Macrophomina[19]. Similarly, AP2/ERF transcription factors assume critical jobs in plant development, improvement and reactions to biotic and abiotic stresses [20]. In this study, certain DGEs encoding WRKY and AP2/ERF transcription factors exhibited different expression patterns, and we found that WRKY transcripts were accumulated in response to C. gloeosporioides(Fig. 5).

V. CONCLUSIONS

Using Illumina sequencing technology, a total of 25,515 genes were obtained and annotated. Analyzing the DGE library provided comprehensive, valuable information regarding the defense mechanism against anthracnose in J. curcas. The expression pattern analysis using RNA-Seq suggested the following: WRKY transcription factors, calcium-binding protein CML and KCS; 3-ketoacyl-CoA synthase upregulation activated defense-related genes induction in 9-1 cultivar which contributes to defense mechanism. Overall, our data provide valuable new clues and information about the C. gloeosporioides induced anthracnose in 2 different cultivars of J. curcas, and further study on the anthracnose-related genes will be helpful in elucidating the defense mechanism of J. curcas.

ABBREVIATIONS

TF: Transcription factor GO: Gene ontology SRA: Sequence read archive DGEs: Differential Gene Expression FPKM: Fragments Per Kilobase Million GTF: Gene transfer format NCBI: National Center for Biotechnology Information KEGG: Kyoto Encyclopedia of Genes and Genomes

VII. DECLARATIONS

a. AVAILABILITY OF DATA AND MATERIALS The datasets supporting the conclusions of this article are included within the article and RNA-Seq data is available as accession number SRP044808 in the NCBI SRA database

(https://www.ncbi.nlm.nih.gov//sra/?term=SRP044808).

 AUTHOR'S CONTRIBUTIONS
Dr. Neeta and Dr. Siddhivinayakconceived the study. Mr. Krunal carried out the analysis and wrote the manuscript. All authors read and approved the final manuscript.

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CONFLICT OF INTEREST

The authors declare that they have no Conflict of Interest.

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