

1 **Climate resilient technologies, adaptation, mitigation and their influence on**  
2 **farm income in Telangana State of India**

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5 **ABSTRACT**

6 Climate change is a challenge for developing countries like India due to climate  
7 variability. The study area of Telangana is also prone to frequent droughts, resulting in distress  
8 among farming communities, and witnessed crop failures leading to farmer's suicides. The State  
9 Action Plan on Climate Change (APCC) reported these issues of drought and heat waves in  
10 agriculture. Therefore, the present study was undertaken in 2017-18 by using an ex-post facto  
11 research design to assess the climate resilient technologies, mitigation strategies adopted here,  
12 and their influence on farm income among the already ascertained most vulnerable  
13 Mahabubnagar district and least vulnerable Khammam district farmers of Telanagana State. A  
14 total of 480 respondents, 240 from each district, were selected randomly. The technology  
15 adoption index was constructed to ascertain the extent of adoption of climate-resilient  
16 technology, and further study analyzed the factors influencing the adaptation of climate-  
17 resilient, technologies such as the technology adoption index on household farm income, along  
18 with other factors. The results of the study indicated that different strategies adopted by farmers  
19 in mitigating climate change, such as use of drought-tolerant varieties and life saving irrigations  
20 had a positive influence on adaptation to climate change. Further, the study reported that the  
21 technology adoption index (TAI) showed a positive, non-significant impact on farm income in  
22 both scenarios considered i.e. Mahabubnagar and Khammam which implies that there needs,  
23 adopting more technologies intensively and timely, have to significant influence on farm income.

24 **Keywords:** Climate resilience, Agriculture, Technology Adoption Index, Climate-resilient  
25 technologies, Telangana.  
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## 31 INTRODUCTION

32 Climate change can be understood as a long-term phenomenon indicating the statistical  
33 distribution of weather patterns over a period of time ranging from decades to millions of years  
34 (IPCC, 2011). Climate change is an important issue for the 21st century, requiring renewed  
35 attention by all stakeholders to address its adverse impact on agricultural productivity, the  
36 environment, and its effect on food security (Fischer, 2001). Livelihoods can be defined as an  
37 adequate stock and flow of food and cash for an individual or a family to meet its basic needs  
38 (Acharya, 2006). Agriculture provides such a livelihood to more than half of the Telangana  
39 state's workforce, which is crucial for restoring sustainable growth in the rural economy.  
40 According to (Rehana, 2018) the frequency of severe and widespread multi-year droughts is due  
41 to erratic summer monsoon, and an increase in air temperature, creating huge damage to crops  
42 and society. Crop failure due to drought has resulted in 342 farmers committing suicide, in 2015  
43 as reported by the Ministry of Agriculture, Government of India. In 2016, about 14 lakh people  
44 from Mahabubnagar, Ranga Reddy, Medak, and Nizamabad migrated to Hyderabad and other  
45 cities like Pune, Mumbai, and Ahmedabad, as mentioned in the report of farmers organizations  
46 (Hindustan Times, 2016). Hence, climate extremes pose a serious threat to crop yields, food  
47 security and farm income [Kalli and Jena, (2022, 2023)]. Effectively addressing climate change  
48 in agriculture requires a comprehensive strategy, with the focal point being a transition toward  
49 agricultural practices resilient to climate variations. Theoretically, farmers can adopt climate-  
50 resilient practices through a combination of external and internal factors. External factors play a  
51 role in rebuilding farmers' capacity to enhance their adaptive skills. Access to extension services,  
52 such as training [Tanti et al. (2023), Zakaria et al., (2020)], farm field schools (Osumba *et al.*,  
53 2021), and demonstrations, makes farmers aware and knowledgeable about the Climate-Smart  
54 Agricultural (CSA) practices [Mgendi et al., (2021), Makate et al., (2019)]. Farmers who receive  
55 credit from public banks, private banks, and cooperative societies are better equipped to adopt  
56 climate-resilient practices (Kongogo et al.,2021).

57 Based on the above background, it can be concluded that there is a low level of  
58 awareness about climate change and knowledge of the factors influencing the decision with the  
59 suitable adaptation strategies and the extent of adoption of those strategies by the farmers.  
60 Coping and mitigating adverse effects of climate change depends on the behavior of climatic  
61 variables, accordingly, the present study has been taken up at the district level in Telangana.

62

## 63 **METHODOLOGY**

64 The present study was undertaken in 10 erstwhile districts of Telangana state. Paddy,  
65 maize and cotton are the three major crops widely grown in Telangana, which constitute a major  
66 part of the total cropped area. The climate-resilient technologies for all the above mentioned  
67 crops were identified, and further primary data pertaining to the information on qualitative  
68 factors vis-a-vis climate change, adaptation strategies, and coping mechanisms adopted by the  
69 farmers was collected purposively from the already, selected villages of the most vulnerable  
70 (Mahabubnagar) and least vulnerable (Khammam) districts of Telangana state after ranking all  
71 the districts based on the estimated vulnerability indices.

72 1. Technology Adoption Index (TAI): An adoption index was constructed to quantify the  
73 adoption of such technologies:

$$74 \quad \mathbf{TAI = A/P*100}$$

75 where,

76 A = Number of practices adopted by respondents, and

77 P = Total number of practices recommended.

78 2. Multiple linear regression

$$79 \quad \mathbf{Y_i = B_0 + B_1X_1 + B_2X_2 + \dots + B_n X_n + e}$$

80 Where, for i=1 to n observations

81  $Y_i$ =dependent variable

82  $X_i$ =explanatory variables

83  $B_0$ = Intercept

84  $B_i$  = slope coefficient

85 e = error term

86 3. Tobit model

87 Adoption of agricultural technology generally involves a two-part decision-making process:  
88 whether to adopt and then how much to adopt. The Tobit model can be used in analyzing  
89 adaptation when the two decisions are made jointly.

## 90 **Results and discussion**

### 91 **1.1 Adaptation to climate change by sample farmers**

92 The data on various adaptation strategies being used by the farmers against the changing  
93 climate was collected and presented in Table 1. A dialogue with the respondents on adaptation of  
94 technologies against climate aberrations is followed by the respondents in both scenarios  
95 considered, i.e., the most and least vulnerable districts. The results revealed that Mahabubnagar  
96 and Khammam districts were practicing improved irrigation related to micro irrigation  
97 technologies as adaptation measures to combat the continuous occurrence of drought. In  
98 addition, to cope with climate variability, the study found that farmers adopted a wide range of  
99 management practices, such as adapting drought tolerant / resistant varieties, changing the  
100 variety in use and selecting a resistant variety (96.25%), change in cropping pattern, and delayed  
101 sowing (99.58%). The most vulnerable conditions prevailing in Mahabubnagar district due to  
102 frequent droughts might have been the reason for adapting drought tolerant varieties. As the  
103 farmers are comparatively well-to-do and are earning better in Khammam, they have the ability  
104 to change cropping patterns which might attract new investment or bear the risk of delay in  
105 sowing. Apart from these, use of trap crops, use of bio pesticides, staggered planting, delay in  
106 sowing, moisture retention by making deep furrows, mulching, adoption of contingent crops,  
107 leaving one season and going for the next season and providing one life saving irrigation *etc*  
108 were considered and ranked in the order of their usage by a large number of farmers. Most of the  
109 farmers in both the districts were leaving the existing crop and going for next season for  
110 cultivation in cases of climatic conditions. Due to recurring under rainfed conditions farmers in  
111 the study area preferred drought tolerant crop varieties. Information on adaptation to climate  
112 change utilizing these technologies was used to develop a technology adoption index, and the  
113 results are furnished in Table 1. Developing the adoption of technology indicators is difficult and  
114 complicated to calculate, as majority of the farmers cannot distinguish which measures they  
115 practice can be called technology and which they practice based from indigenous knowledge or  
116 experience. But the practices followed by respondents which clearly have a bearing on climate  
117 resilience, were considered carefully. A technology adoption index has been worked out for each  
118 sample farmer and found that technology adoption index ranged from 51 to 70 percent in both  
119 districts.

120 **Table 1 Adoptions of technology to climate variability by respondents in Khammam and**  
 121 **Mahabubnagar districts (n1= 240 and n2= 240)**

Adaptation	Khammam		Mahabubnagar	
	Number	Rank as per priority	Number	Rank as per priority
Change in variety such as Drought tolerance / resistant varieties	235 (97.91)	2	231 (96.25)	1
Change in cropping pattern and delayed sowing	239 (99.58)	1	227 (94.58)	2
Staggered planting	114 (47.50)	5	122 (50.83)	5
Leaving existing an crop and growing another crop	108 (45.00)	9	112 (46.67)	8
Leaving the existing crop and growing next season	125 (52.08)	3	130 (54.17)	3
Moisture retention by making deep furrows etc	109 (45.42)	8	125 (52.08)	4
Mulching	106 (44.17)	10	113 (47.08)	7
One life saving irrigation	115 (47.92)	4	111 (46.25)	9
Use of trap crops such as marigold, cowpea and bhendi	110 (45.83)	7	122 (50.83)	5
Use of pesticides	111 (46.25)	6	116 (48.33)	6

122  
 123 Based on the calculation of technology adoption index (TAI) for all the respondents were  
 124 classified as low adopters (0-25), medium adopters (26-50), semi medium adopters (51-75) and  
 125 high adopters (75 -100) and presented in Table 2. The technology adoption index ranged from 0-  
 126 100. Evidently, it was found that only 7.5 % and 2.92 % of the erstwhile Khammam and  
 127 Mahabubnagar districts were low adopters, so it can be understood that the lowest level of  
 128 technology adoption was observed in Mahabubnagar farmers compared to Khammam. This  
 129 might be due to the least loss incurred, so they might not have adopted many measures to combat  
 130 the drought in Mahabubnagar district. Mean Technology adoption index for Khammam sample  
 131 farmers is 48.2, while that of Mahabubnagar is 57.54.

132 Technology adoption index in case of medium adopters among Khammam sample  
 133 farmers about 57.5% were there when compared to 39.18% of farmers in Mahabubnagar. But

134 when the semi medium level of adaptation among these districts was compared to 42.08% of  
 135 Mahabubnagar farmers they were found to have adopted more technologies than 31.67% of  
 136 farmers in Khammam. It can be concluded that the vulnerability scenarios both do not show a  
 137 differentiated effect on the technology adoption index it is the income, resource endowment,  
 138 awareness, affordability and knowledge gained from the media and their fellow farmers that  
 139 might have helped them decide which set of adaptation measures to follow.

140 **Table 2 Technology adoption index among the sample farmers**

Distribution of TAI (%)	Number of Farmers	
	Khammam	Mahabubnagar
<b>Low Adopter TAI (0-25)</b>	18 (7.50)	7 (2.92)
<b>Medium Adopters TAI (26-50)</b>	138 (57.50)	94 (39.18)
<b>Semi Medium Adopters TAI (51-75)</b>	76 (31.67)	101 (42.08)
<b>High adopters TAI (&gt;75)</b>	8 (3.33)	38 (15.82)
<b>Mean Technology Adaptation Index</b>	48.2	57.54

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142 **1.2 Determinants of the level of adaptation to climate change in agriculture**

143 From the results, it is inferred that the coefficient of farm income has a positive influence  
 144 on adaptation to climate change in Khammam but negative in Mahabubnagar district, implying  
 145 that the higher the farm income, higher the willingness and level of climate adaptation  
 146 technologies and vice versa in Mahabubnagar as it is the most vulnerable district. Unlike other  
 147 limited dependent models, the coefficients in the Tobit model can be interpreted directly with the  
 148 magnitudes of influencing the level of adaptation to climate change. It could be interpreted as  
 149 one unit increases in the farm income, will be produce a chance of a 3.05 percent increase in  
 150 adaptation to climate change. Farm size was negatively associated with the level of adaptation to  
 151 climate change in both the districts, and the coefficient -1.806\* was statistically significant in  
 152 Khammam, implying that smaller the farm size, higher the intensity of adaptation to climate

153 change, which may be the cost to be incurred for following adaptation technologies might  
 154 increase as farm size increases. If one unit increases in the farm size, there will be a chance of a  
 155 decrease of 1.8 percent in adaptation to climate change among the Khammam district  
 156 respondents. Similarly, the coefficient of access to credit was negative but highly significant,  
 157 implying that the higher the access to credit, the lower will be the level of adaptation. If one unit  
 158 increases in the credit access will give chance of decreasing 4.95 percent in adaptation to climate  
 159 change in agriculture because farmers might not come forward for utilize their credit money  
 160 towards climate adaptation technologies due their risk averse attitude.

161 **Table 3 Determinants of climate change adaptation**

Variables	Khammam			Mahabubnagar		
	Coefficient	Standard error	Pr(> z )	Coefficient	Standard error	Pr(> z )
Intercept	2.719	4.568	2.00E-16	2.714	4.568	2.00E-16
Age	5.927	1.669	0.7225	1.779	1.338	0.1838
Farming experience	1.41	1.679	0.401	<b>5.361***</b>	2.102	0.0108
Education	2.884	1.962	0.1415	2.007	2.116	0.343
Farm income	<b>3.053*</b>	1.761	0.83	-1.457	1.696	0.948
Farm size	<b>-1.806*</b>	9.804	0.0655	-3.425	1.019	0.8863
Extension contact	2.21	2.077	0.9153	<b>3.425*</b>	2.007	0.0879
Access to credit	<b>-4.95***</b>	2.016	0.0141	2.02	1.995	0.3113
Membership in farmer associations	2.7	2.027	0.1829	2.215	2.038	0.2771
Climate change awareness	2.641	2.687	0.9217	3.533	2.024	0.8614
Depth of ground water levels	-1.27	1.635	0.4374	-2.514	1.719	0.1437
Log-likelihood	-993.193			-986.17		

162 Note:\* and \*\*\* indicate .1 and .001 level of significance

163 **1.3. Performance of various factors influencing farm household income in different**  
 164 **scenario**

165 In order to study if there is any significant influence of level of technology adaptation on  
 166 farm income along with other factors such as age, education, farm size, seed, cropping intensity,  
 167 fertilizer use, plant protection chemicals, experience, number of members available for farm  
 168 work, human labor, credit availability, and cropping intensity were considered and multiple  
 169 linear regression analysis and Tobit analysis were carried out, and the results were presented in  
 170 Table 4

171 **Table 4 Factors influencing farm household income in most and least vulnerable scenarios**

Variables	Khammam			Mahabubnagar		
	Coefficients	Standard Error	P-value	Coefficients	Standard Error	P-value
Intercept	4857.623	10274.161	0.637	7822.869**	5197.420	0.001
Age	393.100**	141.514	0.006	39.884	67.332	0.554
Education	26.102	188.828	0.890	171.269	115.932	0.141
Size of land holding	88.644	787.708	0.910	3703.311***	989.606	0.000
Experience	117.451	131.977	0.374	100.284	135.368	0.460
Technology adoption index (TAI)	8.318	61.382	0.892	15.451	36.540	0.673
No. of members available for farm work	458.833	430.202	0.309	2.000	669.998	0.555
Credit availability	0.372**	0.125	0.003	-0.023	0.024	0.334
Human labour	-0.081	0.290	0.781	-0.107	0.198	0.589
Machine labour	-0.999	0.787	0.206	-0.256	0.566	0.652
Cropping intensity	0.127	0.079	0.110	0.010	0.018	0.998
Irrigation	2.900	2.454	0.239	-2.345	1.262	0.064
Seeds	- 0.985**	0.289	0.001	- 0.724***	0.163	0.000
Fertilizer and manures	1.888***	0.424	0.000	0.384	0.285	0.179
Plant protection	-3.213**	1.003	0.002	-0.274	0.439	0.533
R square	31.330			36.510		

172 **Note: \*\* and \*\*\* indicate .01 and .001 level of significance**

173 The results presented in Table 4, indicated that the technology adoption index (TAI)  
 174 showed a positive, non significant impact on farm household income in both scenarios  
 175 considered i.e. Mahabubnagar and Khammam, which implied that they could adopt more  
 176 technologies intensively and timely; to have a significant influence on farm income. This calls

177 for intensification of research in the direction of developing more crop specific strategies,  
178 technologies, including drought tolerant and disease-resistant ones, and popularizing the same  
179 information using all the digital tools to ensure faster implementation of those technologies at the  
180 field level. Among the other variables considered, seeds and plant protection chemicals showed a  
181 negative correlation with farm income, indicating excessive use in both districts. However,  
182 fertilizers (1.888) and credit availability (0.372) have significantly influenced farm income in  
183 Khammam district. In Mahbubnagar, the size of the holding has significantly influenced the farm  
184 income implying that there is a need to consolidate the holdings to realize economies of scale.

185 The results are in line with the study conducted by Jena et al. (2023), where findings show  
186 that both paddy crop yield and net agricultural income have been higher for the adopters of crop  
187 rotation and integrated soil management, while, adoption of technologies resulted in an increase  
188 of 42–45% in farm income. Further, adoption of these practices led to a yield improvement of  
189 about 2.5 quintals per acre of paddy crop for the adopters. The major drivers that jointly  
190 influenced farmers to adopt these practices were found to be access to extension services, access  
191 to credit, subsidies for seed, and electricity use in agriculture.

## 192 **Conclusion**

193 It can be concluded that among all districts of Telangana, there was a general trend of  
194 narrowing crop diversification which has to be widened. The decline in crop diversification  
195 resulted in crop failures followed by market failures, and subsequently reduced farm income.  
196 The climate resilient technologies that are proved successful must be promoted. Providing  
197 subsidies for the cultivation of millets and creating market linkages might help in bringing crop  
198 diversity to the districts with high vulnerability. There is a need to reorient the extension  
199 approach to promote climate adaptation technologies, and necessary measure, have to be taken to  
200 increase the purchasing power of the farmer as the climate adaptation technology index produced  
201 a non significant positive impact on farm income.

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