

Original Research Article

Impact of on Custom Hiring Centers on Cost, Profitability, and Efficiency in Crop Production in Anantapur District of Andhra Pradesh

ABSTRACT

This study examines the impact of custom hiring centers (CHCs) on agricultural mechanization and productivity of groundnut of cultivation in Anantapur District of Andhra Pradesh in 2021. CHCs, established under the Submission on Agricultural Mechanization (SMAM) scheme, aim to address challenges faced by small-scale farmers in accessing improved machinery. Through farm business analysis, data envelopment analysis (DEA), and binary logistic regression, the study evaluates the adoption of improved farm machinery, changes in costs, profitability, and efficiency in crop production due to CHCs. Utilizing purposive sampling, data was collected from 160 beneficiary and non-beneficiary farmers in villages. Cost concepts were used to compare costs and returns in groundnut cultivation among CHC beneficiaries and non-beneficiaries. Findings indicate that CHC beneficiaries demonstrate lower input costs and higher net returns, signaling improved efficiency and profitability. DEA analysis reveals higher technical, allocative and cost efficiency among beneficiary farms compared to non-beneficiary ones. Additionally, binary logistic regression identifies significant factors influencing farmers' decisions to hire machinery from CHCs, including machine labor, net income, hired human labor, irrigation, and yield. Overall, the study underscores the pivotal role of CHCs in enhancing agricultural productivity and efficiency, particularly among small-scale farmers. It suggests that policymakers prioritize expanding CHCs and promoting mechanization to further improve agricultural outcomes and address challenges faced by marginalized farmers.

Keywords: Custom Hiring Centers, Data Envelopment Analysis, Binary logistic Regression, Submission on Agricultural Mechanization, Technical efficiency, Allocative efficiency and Cost efficiency

1. INTRODUCTION

Mechanization has been acknowledged as a pivotal factor in augmenting land productivity by facilitating timely execution of agricultural tasks, enhancing labor efficiency per unit of time through reduced exertion, and upgrading the overall quality of farm activities [14]. It leads to savings of approximately 15–20 percent on inputs such as seeds and fertilizers, 20–30 percent on labor and operational time, 5–20 percent on cropping intensity, and 10–15 percent on crop productivity [4]. Numerous challenges accompany the adoption of farm mechanization, including economic hurdles like high initial and maintenance costs, inadequate access to institutional credit, exorbitant custom hire charges, escalating production costs, limited availability of service centers, situational constraints such as small and fragmented land holdings, unsuitability of certain machines for various soil types, and technological limitations such as a lack of proficiency in operating machinery and implements [7].

Custom hiring centers (CHCs) serve as crucial facilitators in introducing advanced agricultural machinery to small-scale and marginalized farmers in India, thereby addressing the hurdles associated with mechanization and fostering increased productivity[9]. To promote agricultural mechanization, the Indian government has initiated various schemes and programs such as the RashtriyaKrishiVikasYojana (RKVY), the National Food Security Mission (NFSM), and the National Horticulture Mission (NHM). Notably, prior to 2014, the Government of India did not introduce specific schemes or programs for the promotion of agricultural mechanization. However, in the fiscal year 2014-15, the Submission on Agricultural Mechanization (SMAM) was launched to assist small-scale and marginalized farmers by establishing custom hiring centers (CHCs), farm machinery banks (FMBs), and high-tech hubs to facilitate access to farm machinery. The proliferation of CHCs, high-tech hubs, and FMBs at the grassroots level has ensured that small-scale and marginalized farmers have access to modern agricultural machinery for various field operations[8].

The state of Andhra Pradesh has emerged as a frontrunner in establishing CHCs under the SMAM scheme, with Anantapur district being one of its largest districts dominated by small-scale and marginalized farmers. In light of this backdrop, this study was undertaken to assess the changes in cost of operations, profitability and efficiency in crop production due to adoption of improved farm machinery owing to CHCs in the Anantapur District of Andhra Pradesh.

2. MATERIALS AND METHODS

2.1 Data

Andhra Pradesh state was chosen purposively to study the performance of custom hiring centers established with government support. It is because Andhra Pradesh is the leader in the establishment of CHCs with the highest number of CHCs established under the SMAM scheme. Rayalseema region was selected purposively because most of the farmers belong to small and marginal categories in this region and the objective of the government scheme is to cover small and marginal farmers. Anantpur district from the Rayalseema region was intentionally selected as it is one of the largest districts and most of the

Comment [A1]: The fact that it is any "efficiency" pre-empts the need to write "per unit time".

Comment [A2]: "percent" is a single word.

Comment [A3]: I think there is need for hard figures to back these claims like, for example the percentage of CHCs established in the district in relation to the regional/national total and the proportion of farmers in the district who are small-scale and marginalised in comparison to the national/regional average.

Comment [A4]: This somewhat unscientific, state the proportion/percentage of farmers belonging to the small and marginal categories in the region.

farmers are small and marginal farmers. Government subsidy support in ~~e~~Establishing CHCs plays an important role in such areas. In the district, 2 taluks were selected randomly namely Anantapur and Dharmavaram and from each of the selected taluks, 2 clusters of villages, one having formal CHCs established with government support and another without formal CHC and away from the first one so that custom hire services are not available in the villages from ~~the~~ formal CHC, ~~was~~~~were~~ selected. From each of the selected ~~clusters-villages~~~~of-villages~~, 40 farmers were selected randomly. Thus a total sample of 160 ~~farmers which constituting of~~~~includes~~ 80 CHC farmers and 80 ~~Non-B~~beneficiaries farmers were selected ~~randomly from the study area~~. The ~~p~~Primary data was collected using a personal interview ~~s~~method with the ~~help~~~~aid~~of a pre-tested ~~specially designed~~ schedule ~~designed~~ for the study. ~~The data was also collected from different stakeholders and officials involved in the implementation of the scheme.~~

Comment [A5]: There is need for consistence in the use of Upper or Lower case in naming this group and the author changed between upper and lower case throughout the document

2.2 Farm Business Analysis

Farm business analysis was performed to estimate different costs and returns in crop cultivation on ~~sample farms~~. For this purpose cost concepts used by the Commission on Agricultural Costs and Prices (CACAP), were used to estimate profitability and different costs to compare across farm size categories for both the beneficiaries and non-beneficiaries of CHCs to assess the changes due to mechanization through CHCs. ~~The sample farmers were classified into two groups i.e. beneficiaries and non-beneficiaries of CHC~~. The various cost concepts used to analyze the costs and profitability of Groundnut production are discussed below:

Comment [A6]: Was this data collected using a different technique or instrument from that indicated as "primary data"? If so, what technique and/or instruments were used?

Cost Concepts:

- Cost A_1 = Wages of hired labor, cost of input, hired machinery charges, Imputed value of owned machine power, depreciation on implements and farm buildings, land revenue, and interest on working capital.
- Cost A_2 = Cost A_1 + Rent paid for leased in land
- Cost B_1 = Cost A_1 + interest on the value of owned fixed capital (excluding land).
- Cost B_2 = Cost B_1 + rental value of owned land.
- Cost C_1 = Cost B_1 + imputed value of family labour.
- Cost C_2 = Cost B_2 + imputed value of family labour.
- Cost C_3 = Cost C_2 + 10 percent of Cost C_2 accounting for managerial input

Comment [A7]: Are these "sample farms" owned by the farmers selected as the sample for the study? In other words, what is the relationship between "selected farmers" and "sample farms"? If they are referring to the same subject of analysis, then it would be more efficient and clearer to use the same name to refer to them. As it is, it is not clear which is which.

Comment [A8]: I think a citation is needed to indicate which version of these standard/concepts are being referred to.

Farm Returns

- Farm business income = Gross income – Cost A_1
- Family labour income = Gross income – Cost B_2
- Net income over Cost C_1 = Gross income – Cost C_1
- Net income over Cost C_2 = Gross income – Cost C_2
- Net income over Cost C_3 = Gross income – Cost C_3

Comment [A9]: This has already been mentioned.

2.3 Data envelopment analysis approach (DEA):

DEA is a non-parametric linear programming method for evaluating the performance of a set of peer entities called decision-making units (DMUs). In this case, the individual farm was considered as a decision-making unit. It measures the technical efficiency based on estimated best-practice or efficient frontier or envelopment surface made up of a set of Pareto-efficient DMUs (efficiency score = 1). The efficiency of the farms was calculated about this and the efficiency score was a s between 0 and 1. Technical efficiency corresponding to constant return to scale (CRS) assumption is known as Overall Technical Efficiency (OTE) which captures efficiency due to both managerial and scale effects.

Considering N DMUs, $i=1, \dots, N$ and assuming that there are M outputs and N inputs. Let y_i and x_i denote, respectively, the output and input vectors for the i -th DMU. The $K \times N$ input matrix X and the $M \times N$ output matrix Y , represent the data of all N DMUs.

To estimate the technical efficiency, the envelopment form of the linear programming problem using the duality was used as

$$\begin{aligned} \min_{\theta, \lambda} \quad & \theta, \\ \text{st} \quad & -y_i + Y\lambda \geq 0, \\ & \theta x_i - X\lambda \geq 0, \\ & \lambda \geq 0, \end{aligned}$$

Where θ is a scalar and λ is an $N \times 1$ vector of constants. The value of θ is the efficiency score for the i^{th} firm [2]. Table 1 shows the description of variables along with their unit of measurement which were used to perform DEA analysis.

Table 1. Description of the input and output variables used in the DEA analysis

| S. No. | Variables | Units (per acre) |
|--------|--------------------|------------------|
| I | Output variable | |
| 1 | Profit | Rupees |
| II | Input variables | |
| 1 | Seed material cost | Rupees |
| 2 | Fertilizer cost | Rupees |
| 3 | Human labour | Total hours |
| 4 | Machine labour | Total hours |

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Comment [A10]: What is the meaning of this abbreviation? It is not defined anywhere in the document.

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Comment [A11]: I think by adjusting the alignment of the table contents it becomes easier to distinguish the categories of the variables from the variables themselves.

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2.4 Binary Logistic Regression Model

A logit model was used to identify the factors influencing farmers to hire machinery from CHCs. Using the maximum likelihood estimation method, the model predicts the probability of a binary outcome. It deals with situations in which the outcome for a target variable can have only two possible ways. Logistic regression model output helps identify important factors (Xi) impacting the target variable (Y) and also the nature of relationship between each of these factors and dependent variable

It helps to identify the factors (Xi) impacting the dependent variable (Y) and also shows the nature of relationship between dependent variable and explanatory variables.

The logistic function can be represented as,

$$Y = \ln \left(\frac{P_i}{1-P_i} \right) = \alpha + \beta_1 X_1 + \dots + \beta_n X_n + U_i$$

Where, Y is the dependent variable that takes value 1 if a farmer hired machinery from CHC and otherwise it takes 0.

- α= the constant of the equation;
- β= the coefficient of predictor variables
- Xi= the predictor variables
- U_i=error term

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In this particular case nine (9) independent variables were used as mentioned below shown in Table 2. The logistic function for this study was thus rendered as:

$$Y = \ln \left(\frac{P_i}{1-P_i} \right) = \alpha + \beta_1 \text{Age} + \beta_2 \text{Education} + \beta_3 \text{machine labour hours} + \beta_4 \text{Net income} + \beta_5 \text{Animal labour hours} + \beta_6 \text{Human labour hours} + \beta_7 \text{Irrigation} + \beta_8 \text{yield} + \beta_9 \text{total seed cost} + U_i$$

Marginal effects (dy/dx): Which reflect the change in probability of Y=1 given a unit change in the independent variable X.

Table 2 gives the outlook or description of dependent and independent variables taken in the binary logit model and also their unit of measurement

Table2. Description and units of variables used in the models

Comment [A12]: Made minor changes to the table columns.

| Variables | | Units of measurement |
|-----------------------------|--|----------------------|
| Dependent variable | | |
| Y _i | Whether Farmer is hiring machineries from CHC or not | 1=yes,0=no |
| Independent variable | | |
| X ₁ | Age | years |

| | | |
|----------------|----------------|---|
| X ₂ | Education | Illiterate=1, Primary=2, Secondary=3, PUC=4, Degree=5 |
| X ₃ | Machine labour | Hours |
| X ₄ | Net income | Rupees |
| X ₅ | Animal labour | Hours |
| X ₆ | Human labour | Hours |
| X ₇ | Irrigation | Yes=1, No=0 |
| X ₈ | Yield | q/ha |
| X ₉ | Seed cost | Total cost |

3. RESULTS AND DISCUSSION

3.1 Costs and return in groundnut cultivation on sample farms

Farm business analysis was performed to understand the economics of groundnut production in both the CHC adopted villages and non-adopted villages. The costs and returns over different costs have been estimated and presented in the Table 3 for both the CHC and ~~nNon-bBeneficiaries~~ farmers across farm size categories. Perusal of the table indicates that all the input cost excepting irrigation and plant protection cost was higher on non-beneficiary farmers in comparison to beneficiary farmers. However, yield was marginally higher by 0.51 percent on beneficiary farms in comparison to Non-Beneficiary farmers. Singh (2005) also found that state having highest mechanization index incurred lower cost of cultivation per quintal of the wheat crop due to increased yield.

The Cost C₂ was 4.8 percent lower on CHC farms as compared to ~~Nonnon-Beneficiaries-beneficiaries~~ farms on overall basis. Similar finding was reported by (Lokeshet al, 2018, Masayuki, 2009). The pattern was similar across farm size categories. The net returns over Cost C₂ of CHC beneficiaries was 20.8 percent more in comparison to non-beneficiaries and net return over Cost C₄ of CHC beneficiaries was 14.73 percent higher than ~~nNon-bBeneficiaries~~. This indicates that the increased use of farm machinery on beneficiary farms has led to ~~increase in~~ better use of inputs and hence reduction in cost of cultivation. ~~The better use of inputs led to increase in crop yield and hence the increase in gross returns, not returns and returns over cost C₂~~. Verma (2005) concluded that farm mechanization enhances the production and productivity of different crops due to timeliness of operations, better quality of operations and precision in the application of the inputs

Table 3. Costs and returns in groundnut cultivation at beneficiaries and non-beneficiaries of CHC farms (Rs/ha)

| Particulars | Beneficiary farmers | | | Non-bBeneficiaries farmers | | | Change over Non-bBeneficiaries farms (%) |
|-------------|---------------------|-------|---------|----------------------------|-------|---------|--|
| | Marginal | Small | Overall | Marginal | Small | Overall | |

Comment [A13]: This reference is not in the reference list. Also, its style of citation differs from others in the document.

Comment [A14]: In improve the sentence construction for clarity. I could not understand the point the author was trying to make.

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Comment [A15]: Exactly what was similar? A lower C₂ or that C₂ was lower than 4.8 %?

Comment [A16]: This citation is not in the reference list and also the style of citing is different of that used in most of the document.

Comment [A17]: Which pattern is being referred to as no pattern has been identified by the author?

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Comment [A18]: I think this is too simplistic an explanation in light of the multiple factors/variables at play. I think this statement can be deleted without reducing the value of the study findings but it does an injustice in explaining the interplay of variables responsible for the outcome.

Comment [A19]: These terms are not defined anywhere in the document. This makes it hard to relate the results to the previous discourse.

| | | | | | | | |
|---------------------------------|--------|--------|---------|--------|--------|---------|--------|
| Human labour | 16720 | 16480 | 16600 | 17507 | 18327 | 17917 | -6.68 |
| Animal labour | 6856 | 6644 | 6750 | 8015 | 7835 | 7925 | -13.49 |
| Machine labour | 15123 | 17172 | 16148 | 20146 | 21152 | 21152.5 | -28.50 |
| Seed | 14800 | 14450 | 14625 | 14910 | 14790 | 14850 | -0.34 |
| Manures & Fertilizers | 5682 | 5777 | 5715 | 5815 | 6010 | 5955 | -4.58 |
| Irrigation charges | 542.5 | 543.5 | 543 | 496 | 498.5 | 497.5 | 9.05 |
| Plant Protection Chemicals | 1455.5 | 1544.5 | 1500 | 1411.5 | 1489.5 | 1450.5 | 0.34 |
| Miscellaneous | 1435 | 1621 | 1528 | 1538 | 1553.5 | 1557.5 | -7.87 |
| Cost A ₁ | 60633 | 59900 | 60267 | 65900 | 65200 | 65550 | -7.50 |
| Cost A ₂ | 61591 | 60859 | 61225 | 66783 | 66083 | 66433 | -7.29 |
| Cost C ₁ | 67886 | 67155 | 67521 | 73078 | 72268 | 72673 | -6.59 |
| Cost C ₂ | 82883 | 80905 | 81894 | 88075 | 86048 | 87061.5 | -4.80 |
| Yield (q) | 19.6 | 19.8 | 19.7 | 19.2 | 19.7 | 19.5 | 0.51 |
| Price (Rs/q) | 5600 | 5620 | 5610 | 5560 | 5580 | 5580 | 0.36 |
| Gross return | 112958 | 114459 | 113709 | 110958 | 112957 | 111958 | 0.89 |
| Farm Business Income | 51367 | 53600 | 52484 | 44175 | 46874 | 45525 | 12.83 |
| Family labour Income | 34164 | 38054 | 36109 | 28630 | 31309 | 29969.5 | 14.00 |
| Net Returns over C ₁ | 45072 | 47304 | 46188 | 37880 | 40689 | 39284.5 | 14.73 |
| Net Returns over C ₂ | 30075 | 33554 | 31814.5 | 22883 | 26909 | 24896 | 20.80 |

Source: Compiled from field survey, 2021

3.2 Efficiency in groundnut production across CHC and Non-Beneficiaries farmers

With purpose to examine the impact of adoption of improved farm machineries owing to establishment of CHCs on efficiency in groundnut production, using DEA approach was used to determine technical efficiency (TE), allocative efficiency (TAE), scale efficiency (SE) and economic efficiency (EC) in groundnut production for both the CHC farmers and non-CHC farmers was estimated and to know the status of efficiency of groundnut farms, farms of both the CHC and Non-Beneficiaries villages were

classified into four categories on the basis of their technical efficiency score into (i) Efficient having TE equal to one; (ii) Less efficient having TE more than 0.8 but less than 1; (iii) Moderately efficient having TE score more than 0.6 but less than 0.8 and (iv) Inefficient having TE score less than 0.6, and are The results are summarized in the Table 4. Perusal of the table revealed that out of a total 80 sample farms, 18.8 percent of beneficiaries and 13.8 percent of non-beneficiaries of CHC were technically efficient; 50 percent of beneficiaries and 40 percent of non-beneficiaries were less efficient; 31.2 percent of beneficiary and 46.2 percent of non-beneficiary were moderately efficient and none of the farms either in CHC or Non-Beneficiaries villages were technically inefficient. The overall technical efficiency of beneficiaries was high compared to non-beneficiary's farms. Chinnappa *et al.*, (2018) also observed that farmers hiring machineries from private individual were less efficient compared to those hiring machineries from formal custom hire service providers.

Only 2.5 percent of beneficiaries and none of the non-beneficiaries of CHC had an allocative efficiency score 1 and efficient. However, 57.5 percent of beneficiaries and 75.0 percent of non-beneficiaries were inefficient in allocating their resources had allocative efficiency score of less than 0.6. Overall results convey that allocative inefficiency was in more in case of greater on non-beneficiaries of CHC compared to beneficiary's farms. necessary to expand their business to cover more number farmers as the demand is high during peak season. Ranade *et al.* (2006) and Kulkarni, (2009) observed that Custom hiring services have multiple opportunities to provide better implements to their clients at reasonable rates. Economic The cost efficiency was farms were also less lower in the case of for non-beneficiaries of CHC as compared to beneficiaries, and more than 72.5 percent of farms had cost efficiency scores less than 0.6 and hence inefficient in both cases. Hiremath *et al.* (2015) also found that the success of CHCs in the district of Raichur has substantially assisted small and marginal farmers to carry out farm scores operations on time at low cost.

Table 4. Distribution of farms (%) according to Technical (TE), Allocative (AE) Scale (SE) and Economic efficiency (CE) in groundnut production

| Efficiency level | TE | | AE | | EC | |
|--------------------------------------|---------------|-------------------|---------------|-------------------|---------------|-------------------|
| | Beneficiaries | Non-beneficiaries | Beneficiaries | Non-beneficiaries | Beneficiaries | Non-beneficiaries |
| Efficient (E=1) | 18.8 | 13.8 | 2.5 | 0.0 | 2.5 | 0.0 |
| Less efficient (E > 0.8 < 1) | 50.0 | 40.0 | 13.8 | 7.5 | 11.3 | 0.0 |
| Moderately efficient (E > 0.6 < 0.8) | 31.2 | 46.2 | 26.2 | 17.5 | 13.8 | 27.5 |

Comment [A20]: I don't understand why TE was used to categorize the findings. I would think it more meaningful to have a metric for total efficiency that is constituted of TE, AE, SE and EC (maybe just the average of all these) and use that to formulate the categories of efficient, less efficient, moderately efficient and inefficient. If TE is enough to capture all the dimensions of efficiency, then why have the rest?

Comment [A21]: This, again, highlights the need to clarify the relationship between the 160 sample farmers and the so-called "sample farms". One would expect that, if these were one and the same thing, there would be 160 "sample farms" but they are now 80, where did the other 80 vanish to? The author need to clarify this issue. Are these different entities? Is it a matter of response rate or of incomplete information with the other respondents?

Comment [A22]: Where is this result and how were the overall efficiencies calculated? Indeed, it would be very helpful if the overall TE, AE, SE and EC for all relevant categories were shown. This is of paramount importance because the author routinely refers to "overall" efficiency yet there are no results related to such metrics.

Comment [A23]: This citation style is different from most in the document and should be changed for the sake of consistency.

Comment [A24]: I failed to follow the argument advance by the author. How is this finding by Chinnappa *et al* relevant to the present study in light of the results presented? The author should give an explicit explanation of this relationship if it does indeed exist.

Comment [A25]: Again, there should be results of the so-called "overall" score for the different beneficiaries.

Comment [A26]: This citation style is different from most in the document and should be changed for the sake of consistency.

Comment [A27]: This citation style is different from most in the document and should be changed for the sake of consistency.

Comment [A28]: Again, accepting this explanation is hampered by the fact that the results do not give the "overall" / "Total" / "average" EC for each beneficiary category.

Comment [A29]: The table does not have results for SE as claimed by the caption.

| | | | | | | |
|-------------------------|-------|-------|-------|-------|-------|-------|
| Inefficient (E <0.6) | 0.0 | 0.0 | 57.5 | 75.0 | 72.5 | 72.5 |
| Total No. of Farms | 10080 | 80100 | 80100 | 80100 | 80100 | 80100 |

Source: Compiled from field survey (N = 80), 2021

Similarly, the share of scale efficient farm was more among beneficiary farms in comparison to non-beneficiary farms. However, none of the farms in both the categories were inefficient. The share of less efficient farms was 63.7 percent in case of beneficiary farms as compared to 70.0 percent in case of non-beneficiary farms.

Comment [A30]: Where are the results for scale efficiency?

The above discussion indicates that all the technical, allocative, scale and cost efficiencies were more higher among farms covered by formal CHC in comparison to farms in villages not covered by formal CHCs. However, this these results also indicates the existence of ample opportunity room for improvement in all the technical, allocative, cost and scale efficiency in the entire study area irrespective of whether the villages are covered by CHC or not. This may be due to low mechanization in the area as well as poor awareness about the improved package of practices of practices for growing groundnuts. In addition to all other efforts, mechanization of farms can play an important role in improving technical efficiency and hence greater emphasis is required on mechanizing the less mechanized farms.

Comment [A31]: Awareness does not necessarily result in adoption of good practices and so is not directly related to production levels. The actual practices being carried out are the ones directly related to production levels.

3.3 Factors influencing farmers to hire machinery from CHC

A binary logit model was used to identify the factors With a purpose to understand the factors affecting the decision of by a farmers to choose services from CHC, binary logit model was used to identify the factors. The results of the logit model presented in the Table 5 showed that machine labor, net income, hired human labour, yield and seeds were observed to be significant at 1 percent level of understanding of significance while age and irrigation have were found to be significant at 5 percent level of significance. The parameters like factors: age, machine labor, net income, irrigation and yield were positively related to the participation. The animal labour and seeds were found to negatively influence the farmer's decision to hire machinery from CHC. The marginal effect of age, machine labour, net income, irrigation and yield had negatively influenced the farmer's decision to hire machinery from CHC. With increase in age by one year the probability of hiring farm machinery from CHC by the farmer increases by 0.13 per cent keeping the other things constant. Similarly, if the farmer wants to increase machine labor hours by 1 hour the probability that farmers will go to hire machinery from CHC increases by 0.09 per cent. If farmers want to increase animal labour hours by 1 percent, then the probability to hire machineries by farmers from CHC decreases by 0.00002 per cent. To overcome the problem faced by marginal and small farmers in India during the peak season, use of farm machinery will solve the issue.

Table 5. Binary logit estimates for factors influencing farmers to hire machinery from CHC

| Parameter | Coefficient | Marginal effect (dy/dx) | Z value | P > z |
|---|--------------|-------------------------|---------|-------|
| Age(Years) | 0.06602** | 0.13** | 2.07 | 0.04 |
| Education (Illiterate=1, Primary=2, Secondary=3, PUC=4, Degree=5) | 0.2498 | 0.52 | 0.77 | 0.44 |
| Machine labour (Total hrs.) | 0.4636*** | 0.0973*** | 3.69 | 0 |
| Net income(Rs/ha) | 0.00053*** | 0.00011*** | 4.49 | 0 |
| Animal labour (Total hours) | 0.000102 | 0.0000215 | 0.44 | 0.87 |
| Hired human labour (Total hrs.) | -0.05053*** | -0.00218*** | -3.16 | 0.002 |
| Irrigation (1=yes,0=no) | 1.35005** | 0.307** | 1.99 | 0.04 |
| Yield (Qtl/Ha) | 1.4897*** | 0.312*** | 2.67 | 0.008 |
| Seeds(Total cost) | -0.000745*** | -0.0001*** | -3.49 | 0 |
| Pseudo-R-squared | 0.58 | | | |
| LR chi2(10) | 125.72 | | | |
| Prob> chi2 | 0 | | | |

Note: *** significant at 1% level, ** significant at 5% level

4. CONCLUSION

The study highlights the positive impact of CHCs on agricultural mechanization in Anantapur District, Andhra Pradesh. Beneficiary farmers experienced lower input costs, higher profitability, and greater efficiency compared to non-beneficiaries. Factors such as machine labor, net income, and yield influenced farmers' decisions to hire machinery from CHCs positively. Overall, the findings emphasize the importance of CHCs in promoting mechanization and suggest the need for further expansion to enhance agricultural productivity and sustainability in the region. To further promote agricultural mechanization and the effectiveness of Custom Hiring Centers (CHCs), policymakers should focus on expanding CHC networks, providing financial support and subsidies, offering training programs for farmers, investing in rural infrastructure, promoting information and awareness campaigns, fostering collaboration with the private sector, and implementing robust monitoring and evaluation mechanisms.

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