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Mathematical modelling of paddy drying in hybrid greenhouse dryer

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Abstract

During paddy growing seasons, paddy drying experiments were performed on three separate bed thicknesses (5, 10 and 15 cm) in hybrid greenhouse dryer. Paddy-drying continued until a moisture content of 12 percent (d.b) was achieved, moisture content converted to a moisture ratio (MR). Moisture ratio MR of paddy during drying was fitted to the eleven thin layer drying models. The most fitting models was the logarithmic model to explain the drying characteristics of paddy dried during Rabi season in thin layer. Logarithmic model gave $R^2 = 0.9891$, $\chi^2 = 0.00689$, RMSE = 0.024477 and MBE = 4.02×10 -8. Midilli model was the best one that can describe the drying behavior of the *Kharif* Paddy thin-layer drying. Midilli model gave $R^2 = 0.9953$, $\chi^2 = 0.0000408$, RMSE = 0.019042 and MBE = 1.3×10 -5.

Keywords: Solar drying, greenhouse drying, mathematical modeling, moisture ratio

Introduction

India, after China, is the second largest rice producer in the world (*Oryza sativa L.*). In India, both area and production are predominated by paddy. Rice production in India was 115.6 MT during the year 2018-19 which was higher than the previous year's production of 111.01 MT (Anonymous, 2019)^[1]. The drying of farm products is carried out either for preservation or as it as part of the production process. Grain drying using solar energy is the most commonly used process in the world, but the final dried product quality in sun drying is comparatively lower. Sun-drying of grains often contributes to the possibility of infestation and extinction of birds and insects. In fact, the drying process is unregulated. Solar drying is one of the most interesting and feasible applications in tropical and subtropical countries.

For the optimization of the operational parameters and improvements in efficiency of drying systems, mathematical drying modeling is essential. Solar drying must be examined and modeled in order to understand the different needs of each farm product and forecast the solar drying system's effectiveness (Steinfeld and Segal, 1986 ^[2]). The sample must be mathematically modeled based on the drying conditions obtained from the experimental investigation. The thin layer drying includes several critical parameters to obtain drying characteristics of the product. Many researchers have studied the drying characteristics of paddy (Basunia and Abe, 1998 ^[3]; Noomhorm and Verma, 1986 ^[4]; Wang and Singh, 1978 ^[5]; Agrawal and Singh, 1977 ^[6]; Verma *et al.*, 1985 ^[7]) under natural convection at variable temperature and humidity of air (Basunia and Abe, 2001 ^[8]) and various models have been applied with more or less effectiveness for the drying rate prediction. Mathematical simulation equations have been used in a variety of studies involving primarily thin-layer drying simulation. Five of the studies refer to the drying with variable air temperatures and humidity of paddy in forced convection. Hence the present study was taken up to study the mathematical modeling of paddy drying process under forced ventilated conditions.

Materials and Methods Drying Procedure

Paddy drying experiments were conducted during paddy growing seasons namely *kharif* and *Rabi at* three different bed thicknesses (5, 10 and 15 cm) between 9:00 AM and 5 PM using hybrid solar greenhouse dryer by Madhava *et al*, 2017 ^[9]. The paddy harvested in the fresh (Variety: BPT 5204) has been packed with 18 trays and place in the hybrid greenhouse dryer.

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Senior Scientist, Regional Agricultural Research Station, Anakapalle, Andhra Pradesh, India The drying process continued until the moisture content of 12 percent (d.b.) was reached. Four 40 watt DC fans were installed for forced ventilation. The DC exhaust fans were powered by two no-150-watt photovoltaic panels, with 18.5 V rated voltage and 8.10 a rated current.

Mathematical Modeling

Selected models of the thin layer drying are fitted with drying data obtained by thin layer hybrid greenhouse drying of paddy during Kharif and Rabi seasons to evaluate the drying behavior of the paddy. Empirical models developed from the fundamental diffusion models are suitable for food grains in general. Eleven mathematical models were tested for the best model to explain the drying behavior of the paddy in the present study. The drying data involving moisture content versus drying time were transformed into a dimensionless parameter called moisture ratio versus time in order to normalize drying curves.

In order to establish the best paddy drying model that performs adequately under forced-convection greenhouse drying, eleven empirical or semi-empirical models, as outlined in Table 1. The following assumptions are taken in establishing a mathematical model for polyhouse dryers (Janjai *et al.*, 2011 [10])

- (a) There is unidirectional airflow and no air stratification inside the dryer.
- (b) The paddy dried in thin layers and computation is based on the thin layer drying model.
- (c) The specific heat of the air, the cover and the products are constant.

Of the eleven drying models as shown in Table 2, the drying data as the moisture ratio (MR) versus drying time were used. The non-linear regression analysis was performed using the SPSS16.0 software package to establish the model constants and to get the best fit model. The best model to explain paddy drying behavior was selected based on high coefficient of determination (R2) and low chi-square (χ 2), mean bias error (MBE) and root mean square error (RMSE) values (Togrul and Pehlivan 2002 [11]; Diamante and Munro, 1991 [12]; Erenturk *et al.*, 2004 [13]).

Moisture Ratio

In order to compare the experimental results, the moisture content (M) of the sample converted into a moisture ratio (MR). The paddy moisture ratio (MR) has been determined using the following equation during drying.

$$MR = \frac{M - Me}{M_0 - Me}$$

Where

M =Instantaneous moisture content, %, d.b

Mo = Initial moisture content, %, d.b.

Me = Equilibrium moisture content (EMC) of the material, %, d.b.

The modified Henderson equation and the Chung-Pfost equation are two of the best scientific moisture equations. The modified Henderson equation was used to determine the moisture content in accordance with the following equation and paddy constants (ASAE standards: ASAE D245.4).

$$M_e = \frac{1}{100} \left[\frac{\ln(1 - RH)}{-K(T + C)} \right]^{\frac{1}{N}}$$

RH = Relative Humidity, decimal

T = Temperature, °C

For paddy K=1.9187×10⁻⁵, N=2.4451, C= 51.161; Standard error=0.0097

Statistical Analysis

The statistical computer software program was used to perform regression analysis, the key criteria to pick the best equation to interpret the drying curve equation was correlation coefficient (R²). The computation of the validity of fitness was augmented by the calculation coefficient (R²) (Thompson *et al.*, 1979 ^[14]), Reduced chi-square (χ 2), total square error (SSE) and a mean square root error (RMSE) Hagan *et al.*, 1996 ^[15]; Principe *et al.*, 2000 ^[16]. These parameters are defined as follows

Coefficient of determination (R²) draws a distinction between the predicted (MRpre) and the experimental (MR exp) values, and it is essential to obtain the best fit model.

$$R^{2} = \frac{\sum_{i=1}^{N} [(MRexp)(MRpre)] - \frac{1}{N} \sum_{i=1}^{N} (MRpre)}{\sqrt{\left[\left[(MR_{exp})^{2} - \frac{1}{N} (\sum_{i=1}^{N} (MRexp)^{2})\right] \left[(MRexp)^{2} - \frac{1}{N} (\sum_{i=1}^{N} (MRexp)^{2})\right]}}$$

Reduced chi-square $\chi 2$ is used to establish the goodness of the fit, lower value of the $\chi 2$ gives the best fit.

$$\chi 2 = \frac{\sum_{i=1}^{N} (MRexp - MRpre)^{2}}{N - n}$$

N = Total number of observations

n = Number of constants

Root mean square error provides guidance on the short-term performance of the correlations by allowing a term-by-term comparison of the actual deviation between the predicted (MR pre) and the experimental (MR exp) values. RMSE is always positive, but a null value is ideal. When the RMSE is lower, the calculation is more reliable.

$$RSME = \sqrt{\frac{\sum_{i=1}^{N} (MRexp - MRpre)}{N}}$$

Mean bias error (MBE) gives the mean deviation between predicted (MRpre) and experimental (MR exp) values, it must be equal to zero at ideal conditions. A positive value of MBE indicates an over-estimate while a negative value indicates an under-estimate. The MBE can be computed by an equation as follows.

$$MBE = \frac{\sum_{i=1}^{N} (MRexp - MRpre)}{N}$$

The drying equations generally used for thin layer greenhouse drying of food grains are given in Table 1. The primary criterion for selecting the best equation know variations of drying curves of dried samples was the reduced χ -square, root mean square error (RMSE) and mean biased error (MBE). Reduced χ -square is the mean square of the differences between the experimental and estimated values for the models and has been used to determine the fit's goodness. The lower the χ -square value, the better the performance. The RMSE varies between the predicted and the experimental values and is required to achieve zero

Results and Discussion

Results of the paddy drying experiments conducted in hybrid greenhouse dryer were fitted to the eleven empirical or semiempirical models in order to establish the best paddy drying model that performs adequately under forced-convection greenhouse drying, the results are reported below.

Mathematical Modeling

Geometric

Wang& Singh

Diffusion approach Midilli

Figure 1 and Figure 2 indicates the moisture ratio for paddy drying under various seasons and various bed thicknesses. It was clear that, the moisture ratio has decreased continuously with the drying time. The constant rate period was clearly absent, and paddy drying was completed through the entire time in the falling rate period. The falling rate period during drying showed that, the internal mass transfer has occurred by diffusion. The drying process produces moisture gradients within the grain and hence the drying rate retarded. The removal of moisture from the paddy slows down and gets closer to the equilibrium moisture content. The drying of paddy has been recorded with similar observations (Manikantan *et al.*, 2014 [17], Khanali, *et al.*, 2012[18], Akin *et al.*, 2014 [19] and Omid, *et al.*, 2006 [20])

Grain bed thickness also affected the drying rate, with a bed thickness of 5 cm; the drying rate was high due to higher grain temperature and air flow rates compared to 10 cm and 15 cm of bed thickness. Increased drying air temperature, air flow rate and decreased air humidity increases the drying capacity (Hacıhafızoglu, 2008^[21]). The air flow rate has the

effect of moving moisture from the surface and therefore the drying potential increased considerably. Higher air temperature, air flow and lower relative humidity were observed during *Rabi* season and, consequently, the drying rate was higher in Rabi season than in *Kharif* season.

Table 2 and Table 3 provide the results of the statistical analysis carried out on these greenhouse drying models of Kharif and Rabi paddy, the drying model coefficients, and the other comparison criteria used to evaluate goodness of fit namely R2, Chi-square, RMSE, MBE. The logarithmic model was found to be the most appropriate model for depicting the drying curve of the thin drying layer of Rabi paddy. The logarithmic model gave R2 = 0.9892, χ 2 = 0.000689, RMSE = 0.024477 and MBE = 4.02X10-8. Logistical model and Newton models are other appropriate models to characterize the experimental results in addition to logarithmic model.

The Midilli model was found to be the best model in the Kharif season to explain the thin layer drying performance paddy in greenhouses under forced ventilation. Midilli model gave R2 = 0.995523, $\chi 2$ = 0.0000408, RMSE = 0.019042 and MBE= -1.3X10-5. Other appropriate models for describing experimental data are the Wang & Singh, and logarithmic models. Experimental data and predicted data using these models are shown in Figure 5.76. The findings obtained are well in line with the findings, of Omid, 2006 [20]; Oktay *et al.*, 2008 [22]; Manikantan *et al.*, 2014 [17]; Konan *et al.*, 2014 [23]; Basunia and Abe, 2001 [8] and Khanali *et al.*, 2012 [24].

Chandra and Singh (1995)

Wang and Singh (1978) [5]

Kassem (1998)

Midilli et al, (2002)

Norms	Model	References
Newton	MR = exp(-kt)	O'Callaghan et al.(1971)
Page	$MR = exp(-kt^n)$	Agrawal and Singh (1977) [6]
Modified page	$MR = exp(-kt)^n$	Diamante and Munro (1993) [12]
Henderson & Paris	MR = aexp(-kt)	Chhinman (1984)
Logarithmic	$MR = a_0 + aexp(-kt)$	Chandra and Singh (1995)
Logistic	$MR = \frac{a_0}{1 + aexp(-kt)}$	Chandra and Singh (1995)
Two-term	$MR = a_0 + aexp(-k_0t) + bexp(-bt)$	Henderson (1974)

Table 1: Mathematical models given by various authors for drying curves

Table 2: Mathematical	modeling of solar	greenhouse dryer	during Rahi \$20eason
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 $MR = at^{-n}$

 $MR = 1 + a_1t + a_2t^2$

 $\overline{MR = aexp(-kt) + (1-a)exp(-kbt)}$

MR = aexp(-kt) + bt

Model Name	Model Coefficients	R2	RMSE	MBE	χ-square
Newton	K=0.060911	0.9888	0.036458	0.006644	0.006946
Dogo	k=0.038508	0.9875	0.026528	0.001542	0.001688
Page	n=1.176874	0.9873			
Modified page	k= -0.050670	0.9888	0.026459	0.006644	0.001456
Modified page	n= 1.202117	0.9888	0.036458		
Henderson	a=1.076026	0.9879	0.02585	0.000688	0.000732
Henderson	k=0.067306	0.9879	0.02363	0.000088	0.000732
	$a_0 = -0.12174$		0.024477	4.02E-08	0.000689
logarithmic	a = 1.175896	0.9892			
	k = 0.054827				
	$a_0 = 3.951308$		0.024957	0.000133	
Logistic	a = 2.763762	0.9887			0.000716
	k = -0.07922				
Two tame	a=-0.04804 b=1.116448 k=0.087997	0.9882	0.024322	0.003768	0.000716
Two-term	g=0.066728	0.9882	0.024322		
Geometric	a=1.146416	0.8468	0.072835	0.038343	0.00581
Geometric	n=0.331726	0.6406			0.00381
Wong & Singh	$a_1 = -0.0477057$	0.9824	0.02501	-0.01137	0.000686
Wang & Singh	$a_2 = 0.00047589$	0.9824	0.02301		0.000080
Diffusion approach	a=-0.07049 b=0.022546				

	k=2.928041	0.9882	0.02435	0.004064	0.000682
Midilli	a=1.042437	0.9881	0.02274	0.00487	0.000595
	b=-0.00582 k=0.05183	0.9881			

Table 3: Mathematical modeling of solar greenhouse dryer during Kharif Season

Model Name	Model Coefficients	R2	RMSE	MBE	χ-square	
Newton	K=0.06694	0.9762	0.069784	0.004756	0.004939359	
Dogo	k=0.020983	0.9839	0.036492	-0.00565	-0.0061	
Page	n=1.431555	0.9639	0.030492	-0.00303	-0.0001	
Modified Page	k = -0.05584	0.9762	0.069784	-0.00476	0.005259	
Modified Fage	n=1.198813	0.9702			0.003239	
Henderson	a=1.129239	0.9679	0.054402	0.010737	0.003196	
Henderson	k=0.075763	0.9679	0.054402	0.010737	0.003190	
Lagarithmia	$a_0 = -1.04998 \ a = 2.074701$	0.9891	0.019363	-4.7E-06	0.000422	
Logarithmic	k= 0.024358	0.9891			0.000422	
	a ₀ = 1.25638		0.033447	0.003003	0.001259	
Logistic	a = 0.278394	0.9866				
	k= -0.14332					
	a=-9.31593 b=10.29381		0.039041	0.004414		
Two-term	k=0.144336	0.9821			0.001789	
	g=0.131581					
Geometric	a=1.264466	0.7450	0.745002	0.011933	0.023138	
Geometric	n=0.453295	0.7430				
Wang & Singh	a ₁ =-0.0455896 a ₂ =0.00034478	0.9949	0.020712	-0.0017	0.000463	
Diffusion approach	a=-12.4893 b=0.935878	0.9818	818 0.039211	0.005172	0.00173	
Diffusion approach	k=0.13998	0.9818				
V4: 1:11:	a=1.027964 b=-0.01278	0.0055	0.010042	019042 -1.3E-05	0.000400	
Midilli	k=0.037978	0.9955	0.019042		0.000408	

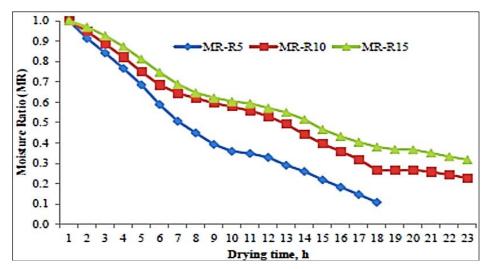


Fig 1: Variation of moisture ratio with drying time during Kharif season

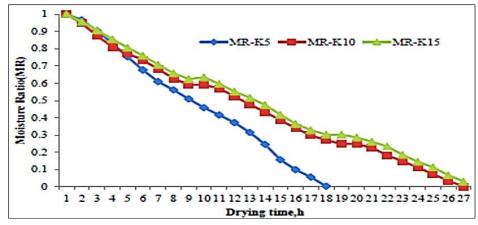
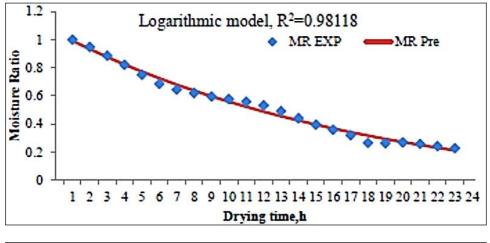
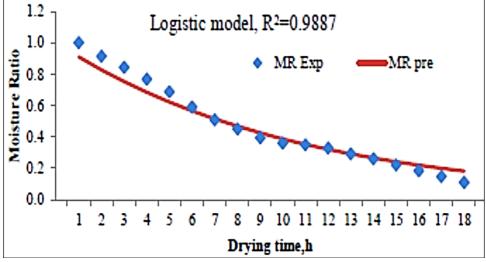


Fig 2: Variation of moisture ratio with drying time during Rabi season





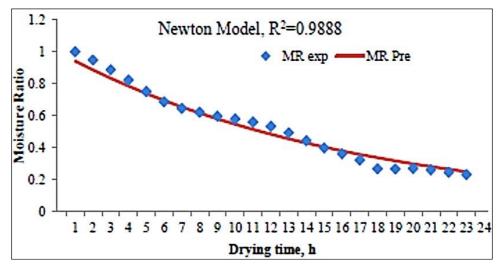
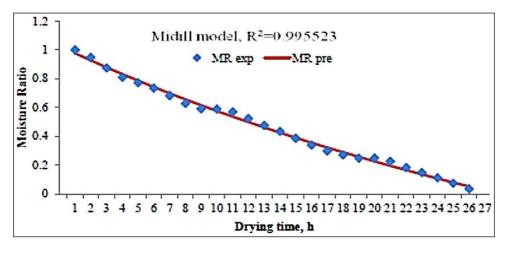
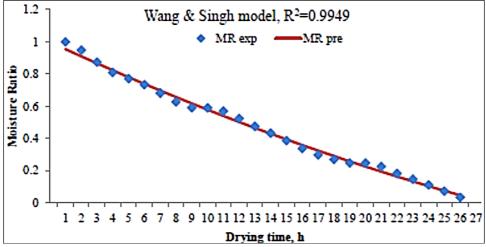


Fig 3: Comparison of experimental and predicted moisture ratios with drying time by different models during Rabi season





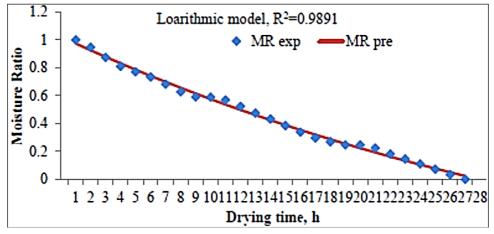


Fig 4: Comparison of experimental and predicted moisture ratios with drying time by different models during Kharif season

Conclusion

The logarithmic model has been found to be the best model to describe the drying characteristics of Rabi paddy among eleven thin-layer drying models. It was also found that the Midilli model was best able to explain the drying behavior of Kharif paddy with thin layer drying.

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