





# Kinetic aspects of a dried thin layer carrot in a heat pump dryer

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#### Abstract

This article presents a mathematical model for drying thin layer carrot slices (*Daucus carota*) using a heat pump dryer (HPD). To select the equation that best describes the drying curve, 10 semi-theoretical and/or empirical models were evaluated. The parameters were determined using the Sigma-Plot<sup>®</sup> program, and their goodness of fit was compared using the correlation coefficient,  $R^2$ ; Chi-squared,  $\chi^2$ ; standard error of the estimate (*SEE*) and root mean square error (*RMSE*). Additionally, the effect of the relative moisture, sample thickness and air velocity on the effective diffusivity of the process was evaluated using a response surface tool. Although all the models correctly fit the experimental data, based on the statistical tests, the Wang-Singh model was selected as the best.

Keywords: drying; thin layer; modeling; diffusivity.

# Aspectos cinéticos del secado de capa delgada de zanahoria en un secador de bomba de calor

#### Resumen

Este artículo presenta un modelo matemático de capa delgada para secado de rodajas de zanahoria (*Daucus carota*) usando un secador de bomba de calor (HPD). Para seleccionar la mejor ecuación que describe la curva de secado, se evaluaron 10 modelos semi-teóricos y/o empíricos. Los parámetros se determinaron usando el programa Sigma-Plot® y la bondad de su ajuste se comparó usando el coeficiente de correlación  $R^2$ ; Chi-cuadrado,  $\chi^2$ ; error estándar del estimado (*SEE*) y raíz del error cuadrado medio (*RMSE*). Adicionalmente, se evaluó el efecto de la humedad relativa, el espesor de la muestra y la velocidad del aire sobre la difusividad efectiva del proceso usando la herramienta de superfície de respuesta. Aunque todos los modelos ajustaron correctamente los datos experimentales, se seleccionó el modelo de Wang-Singh como el mejor, basado en las pruebas estadísticas.

Palabras clave: secado; capa delgada; modelación; difusividad.

## 1. Introduction

Regarding the drying process, the term "thin-layer" applies to either a particle suspended freely in the drying air, or one layer of particle or a polylayer of many particles' thicknesses; the temperature and relative moisture of the drying air can be considered to be in the same thermodynamic state during the drying period [1]. Using this definition, any mathematical model for a particle also models the particles drying in a thin layer using any drying method, and the thin layer thickness may change with the velocity, temperature and relative moisture of the drying air.

The thickness of a thin layer increases if the drying air

speed increases and when the thermodynamic state of the drying air reaches equilibrium with dry particles in the layer [1]. Due to the thin sample structure, a uniform temperature distribution can be assumed and may be modeled using lumped parameters [2].

The thin layer equations can be theoretical, semitheoretical or empirical. The first only accounts for internal resistance to the moisture transfer between the product and air, while the others only consider external resistances to this moisture transfer. Theoretical models explain the product's behavior during drying and can be used for all process conditions despite including many assumptions that cause considerable error. The most widely used theoretical models are derived from Fick's second law of diffusion. Similarly, semi-theoretical models are generally derived from Fick's second law and modified to a simplified form. However, using experimental data requires making assumptions, and these theories are only valid within the applied process conditions. Empirical models have similar features to semi-theoretical models that strongly depend on the experimental conditions and provide limited information on the product behavior during drying [2]. The carrot is frequently used in studies of different preservation techniques due to its physical characteristics and available modeling and simulation data [3-12].

This article evaluates the experimental data fit for 10 thin layer models and explores the effect that relative moisture, sample thickness and air speed have on the kinetic behavior of dried carrot slices. The kinetic expression was established from analyzing the response surfaces and kinetic parameters controlling the heat pump drying process. Additionally, we determined the effective moisture diffusivity for each experiment and the activation energy of the processes.

#### 2. Materials and methods

Nantes carrots (Daucus carota L.) were washed and cut into slices 3 cm in diameter and with a thickness of 2, 3 and 4 mm, according to the experimental design. The average initial moisture content of a fresh carrot was  $0.8935 \pm 0.024$ (bh). The heat pump dryer (HPD) (Dártico brand) consisted of a cooling circuit containing a condenser, evaporator, compressor and expansion valve. The drying chamber contained 5 trays ( $0.36 \times 0.36$  m). The relative humidity (*RH*) and air velocity in the HPD were fixed. The drying temperature was directly linked to the RH (Table 1). A threefactor face-centered central composite design (FCCCD) was used with three repetitions at the central point, as shown in Table 1. The total weight (trays plus samples) was recorded for 5 hours with 2 min intervals for the first 10 minutes, 5 min intervals for half an hour, 10 min intervals for an hour and a half and 30 minute intervals for the last two hours.

Table 1.

A three-factor face-centered central composite design (FCCCD)

		level			value			
Experiment	$x_1$	$x_2$	<i>x</i> <sub>3</sub>	RH (%)	<i>e</i> (mm)	<i>V</i> (m/s)		
1	-1	-1	-1	20 (50°C)	2	0.8		
2	1	-1	-1	50 (35°C)	2	0.8		
3	-1	1	-1	20 (50°C)	4	0.8		
4	1	1	-1	50 (35°C)	4	0.8		
5	-1	-1	1	20 (50°C)	2	1.2		
6	1	-1	1	50 (35°C)	2	1.2		
7	-1	1	1	20 (50°C)	4	1.2		
8	1	1	1	50 (35°C)	4	1.2		
9	-1	0	0	20 (50°C)	3	1.0		
10	1	0	0	50 (35°C)	3	1.0		
11	0	-1	0	35 (40°C)	2	1.0		
12	0	1	0	35 (40°C)	4	1.0		
13	0	0	-1	35 (40°C)	3	0.8		
14	0	0	1	35 (40°C)	3	1.2		
15	0	0	0	35 (40°C)	3	1.0		
16	0	0	0	35 (40°C)	3	1.0		
17	0	0	0	35 (40°C)	3	1.0		
Source: The outh	ore							

Source: The authors

Table 2 shows the semi-theoretical models evaluated during the kinetic study [13]. The parameters for each model were estimated using Sigma-Plot<sup>®</sup> software. The moisture ratio (*MR*) value was determined according to the external conditions. If the relative humidity of the drying air is constant during the process, the equilibrium moisture content is also constant. The *MR* value was calculated using eq. (1).

$$MR = \frac{M_t - M_e}{M_i - M_e} \quad , \tag{1}$$

where Mi is the initial moisture content (dry basis) obtained by drying in a vacuum oven at a constant weight and at 60°C,  $M_t$  is the moisture content (dry basis) at time *t* based on the recorded weight, and  $M_e$  is the equilibrium moisture content obtained from the carrot sorption isotherms (dry basis) [14, 15]: 0.055 [20% *RH*; 50 °C], 0.070 [35% *RH*; 40 °C] and 0.080 [50% *RH*; 35 °C].

The model validities were checked using statistical parameters: correlation coefficient ( $R^2$ ), reduced Chi-squared test ( $\chi^2$ ), standard error of the estimate (*SEE*) and root mean square error (*RMSE*). The highest  $R^2$  and lowest  $\chi^2$ , *SEE* and *RMSE* values determined the goodness of fit. The aforementioned criteria were calculated using eq. (2) – (5) [2,16].

$$R^{2} = 1 - \frac{\sum_{i=1}^{N} (MR_{pred,i} - MR_{exp,i})^{2}}{\sum_{i=1}^{N} (\overline{MR_{pred}} - MR_{pred,i})^{2}}$$
(2)

$$\chi^{2} = \frac{\sum_{i=1}^{N} (MR_{exp,i} - MR_{pred,i})^{2}}{N - n}$$
(3)

$$SEE = \left[\frac{\sum_{i=1}^{N} \left(MR_{pred,i} - MR_{exp,i}\right)^{2}}{N - n}\right]^{1/2}$$
(4)

$$RMSE = \left[\frac{1}{N} \sum_{i=1}^{N} \left(MR_{pred,i} - MR_{exp,i}\right)^{2}\right]^{1/2},$$
 (5)

where N is the number of observations, n is the number of constants,  $MR_{pred,I}$  is the *i*th predicted moisture ratio and  $MR_{exp,I}$  is the *i*th experimental moisture ratio.

The Root Mean Squared Error has the advantage that it retains the units of the forecast variable and is thus more easily interpretable as a typical error magnitude. The Chisquared Test is easy to implement (with multivariable data for example) [17] and also quite flexible..

The Henderson-Pabis model was rearranged to determine the diffusivity coefficient based on eq. (6) [2]:

$$ln(MR) = ln(a) - kt, \qquad (6)$$

where a is a shape index and, k is the drying constant defined by eq. (7):

$$k = -\pi^2 D_{eff} / A_2 \quad , \tag{7}$$

Table 2.				
Kinetic	parameters	for thir	laver	models

$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	Weibull	
EXP k n k a c k_o a k_I b k   1 0.00315 1.109 0.00550 1.010 0.00451 1.124 -0.1215 0.0503 -0.123 0.0061 1.092 180.4   2 0.00053 1.234 0.00194 1.023 0.00041 4.029 -3.0140 0.0193 -0.071 0.0022 1.080 451.4   3 0.00177 1.277 0.00719 1.050 0.00548 1.204 -0.1676 0.0414 -0.257 0.0087 1.247 142.9   4 0.00137 1.059 0.00187 1.002 0.00084 1.969 -0.9724 0.0030 2.796 0.0038 -1.800 505.9   5 0.00154 1.365 0.00906 1.055 0.00598 1.286 -0.2559 0.0163 7.524 0.0184 -6.535 115.2	$MR = exp\left[-\left(\frac{t}{k}\right)^n\right]$	
1 0.00315 1.109 0.00550 1.010 0.00451 1.124 -0.1215 0.0503 -0.123 0.0061 1.092 180.4   2 0.00053 1.234 0.00194 1.023 0.00041 4.029 -3.0140 0.0193 -0.071 0.0022 1.080 451.4   3 0.00177 1.277 0.00719 1.050 0.00548 1.204 -0.1676 0.0414 -0.257 0.0087 1.247 142.9   4 0.00137 1.059 0.00187 1.002 0.00084 1.969 -0.9724 0.0030 2.796 0.0038 -1.800 505.9   5 0.00154 1.365 0.00906 1.055 0.00598 1.286 -0.2559 0.0163 7.524 0.0184 -6.535 115.2	n	
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	.365	
6 0.00373 0.983 0.00331 0.986 0.00218 1.330 -0.3523 0.0026 1.070 -0.0017 -0.092 295.3 ···	.983	
7 0.00132 1.388 0.0088 1.062 0.00645 1.228 -0.1845 0.0129 1.822 0.0278 -0.836 118.5	.388	
8 0.00286 1.023 0.00315 0.992 0.00193 1.421 -0.4379 0.0078 -0.473 0.00423 1.453 306.4	.023	
9 0.00054 1.393 0.00410 1.029 0.00106 2.929 -1.9230 0.0263 -0.395 0.0058 1.346 223.5	.393	
10 0.00006 1.627 0.00210 1.089 0.00069 2.816 -1.7360 0.4419 -0.103 0.0022 1.105 401.0	.627	
11 0.00015 1.493 0.00266 1.109 0.00280 1.070 0.0394 1.0620 -0.123 0.0027 1.123 361.3	.493	
12 0.00222 1.109 0.00396 1.008 0.00219 1.509 -0.5141 0.0021 1.615 0.0002 -0.621 246.6	.109	
13 0.00130 1.165 0.00319 1.023 0.00194 1.477 -0.4623 0.0036 1.095 0.0219 -0.089 299.3	.165	
14 0.00081 1.295 0.00405 1.057 0.00222 1.600 -0.5574 0.2265 -0.096 0.0042 1.083 243.8	.295	
15 0.00305 1.049 0.00389 0.999 0.00277 1.255 -0.2649 0.0071 -1.572 0.0056 2.559 250.2	.049	
16 0.00061 1.300 0.00341 1.083 0.00355 1.055 0.0288 84.230 -0.097 0.0035 1.093 297.8	.300	
17  0.00457  0.967  0.00371  0.980  0.00302  1.124  -0.1483  0.0039  1.011  0.0369  -0.045  263.2  (1.104)  -0.045  -0.04	.967	

Source: The authors

Table 2.

Kinetic parameters for thin layer models (cont.)

Model	Newton		Modified Henderson-Pabis					Wang	ng-Singh Ve		Vega-Lemus		Proposed model	
	MR = ex	p(-Kt)	$MR = a \exp(-kt) + b \exp(-gt) + c \exp(-ht)$		(-ht)	$MR = 1 + at + bt^2$		$MR = (n + kt)^n$		$MR = ab^t$				
EXP	K	k	а	g	b	h	С	а	b	k	п	а	b	
1	0.00537	0.2234	0.385	0.1420	-0.457	0.0060	1.075	-0.0047	6.74E-06	-0.0021	0.992	1.0202	0.9944	
2	0.00179	0.0083	-0.179	0.4070	-0.017	0.0025	1.195	-0.0014	-3.5E-07	-0.0009	1.009	1.0267	0.9980	
3	0.00667	0.0085	1.212	0.0602	-0.291	0.1712	0.080	-0.0055	8.16E-06	0.0027	-1.009	1.0897	0.9921	
4	0.00186	0.0016	1.000	-0.0153	-1.248	-0.0153	1.244	-0.0017	7.9E-07	-0.0008	0.999	1.0039	0.9981	
5	0.00845	0.0105	6.951	0.0074	-2.276	0.0142	-3.676	-0.0065	1.04E-05	-0.0034	1.010	1.2592	0.9873	
6	0.00342	0.0066	-1.751	0.2636	0.040	0.0051	2.712	-0.0032	3.4E-06	-0.0014	0.987	0.9925	0.9966	
7	0.00809	0.0034	-0.641	0.0124	1.605	0.3702	0.037	-0.0064	0.000011	0.0033	-1.015	1.1645	0.9892	
8	0.00322	0.0038	1.155	884.10	0.026	0.0122	-0.181	-0.0029	2.9E-06	0.0013	-0.990	0.9982	0.9968	
9	0.00388	0.0526	-0.485	0.0055	1.249	0.1703	0.239	-0.003	1.3E-06	0.0017	-1.008	1.0560	0.9955	
10	0.00154	0.7902	-0.086	0.0023	1.119	0.0426	-0.032	-0.0007	-3E-06	0.0010	-1.041	1.0923	0.9979	
11	0.00195	0.8034	-0.132	-0.0004	0.158	0.0036	0.974	-0.0012	-1.9E-06	-0.0012	1.048	1.1071	0.9974	
12	0.00385	0.0060	2.179	0.0088	-1.202	0.1764	0.025	-0.0033	0.000003	-0.0016	0.996	1.0229	0.9959	
13	0.00303	0.0037	1.143	0.0125	-0.122	0.1195	-0.022	-0.0026	1.6E-06	-0.0014	1.006	1.0299	0.9967	
14	0.00361	0.0027	1.491	0.3389	-0.073	0.0003	-0.430	-0.0028	1.05E-06	-0.0017	1.020	1.0758	0.9957	
15	0.00390	0.0051	1.496	53.980	0.015	0.0089	-0.512	-0.0035	3.8E-06	0.0016	-0.992	1.0086	0.9960	
16	0.00282	0.0204	0.061	0.0032	1.046	1.0140	-0.111	-0.0023	4.8E-07	0.0015	-1.033	1.0815	0.9967	
17	0.00387	0.2358	1.229	0.2124	-1.230	0.0039	1.003	-0.0037	4.7E-06	0.0015	-0.983	0.9847	0.9962	
0 7	C1 (1													

Source: The authors

Table 3.

Statistics parameters

Statisties parameters									
Model		$R^2$	S	SEE		$\chi^2$		RMSE	
Newton	0.9771	±0.0233	0.0352	±0.0216	0.0017	±0.0017	0.0345	±0.0211	
Page	0.9913	±0.0119	0.0219	$\pm 0.0162$	0.0007	$\pm 0.0010$	0.0211	±0.0156	
Henderson-Pabis	0.9916	$\pm 0.0104$	0.0239	$\pm 0.0142$	0.0008	$\pm 0.0009$	0.0229	±0.0137	
Logarithm	0.9950	$\pm 0.0067$	0.0184	$\pm 0.0117$	0.0005	$\pm 0.0006$	0.0173	$\pm 0.0110$	
Two terms	0.9979	±0.0031	0.0116	$\pm 0.0075$	0.0002	$\pm 0.0003$	0.0106	$\pm 0.0069$	
Wang-Singh	0.9986	±0.0023	0.0087	$\pm 0.0065$	0.0001	$\pm 0.0002$	0.0083	$\pm 0.0062$	
M Henderson-Pabis	0.9986	±0.0023	0.0095	$\pm 0.0071$	0.0001	$\pm 0.0002$	0.0084	$\pm 0.0062$	
Vega-Lemus	0.9947	$\pm 0.0070$	0.0183	$\pm 0.0116$	0.0005	$\pm 0.0006$	0.0176	$\pm 0.0111$	
Weibull	0.9913	$\pm 0.0120$	0.0218	±0.0163	0.0007	$\pm 0.0010$	0.0209	±0.0157	
This paper	0.9866	$\pm 0.0196$	0.0311	$\pm 0.0287$	0.0017	±0.0035	0.0305	$\pm 0.0281$	

Source: The authors

where  $A_2$  is a geometric constant ( $4e^2$  for infinite slices), and e is half the slice thickness if drying occurs on both sides and the full thickness if drying occurs on only one side. Eq. (6) indicates ln(MR) varies linearly with t and the slope is equal to k.

Eq. (8) was used to calculate the activation energy  $(E_a)$ :

$$ln(D_{eff}) = ln(D_0) - 10^3 \left(\frac{E_a}{R}\right) \left(\frac{1}{T + 273.15}\right) , \qquad (8)$$

where  $D_0$  (m<sup>2</sup>/s) is the Arrhenius factor that is generally defined as the reference diffusion coefficient at an infinitely high temperature,  $E_a$  (kJ/mol) is the diffusion activation energy, and *R* (kJ/kmolK) is the universal gas constant. Eq. (8) is linear and allows  $E_a$  to be calculated from the slope.

### 3. Results and discussion

Table 2 shows the parameters for each model that fit the experimental data and Table 3 shows the fit for each model using the statistical parameters from each experiment. In general, all models had high correlation coefficients and low values for the other statistics. The Wang-Singh and modified Henderson-Pabis models exhibited the best values; however, the Wang-Singh model has fewer adjustable parameters. A statistical discrimination study based on nonlinear regression ensures the Wang-Singh equation best fits the data [18].

Thin layer equations may be theoretical, semi-theoretical, and empirical models. Semi-theoretical models are generally derived from Fick's second law of diffusion and are modifications of its simplified forms [2].

Empirical models used are: The Lewis (Newton) model, which is analogous with Newton's law of cooling. The Page model modifies the Newton model to get a more accurate model by adding a dimensionless empirical constant (n) [2].

The models derived from Fick's second law of diffusion are: the Henderson-Pabis model, and the Logarithmic model, which is formed by adding an empirical term. The Two-Term model uses the first two terms of the general series solution of Fick's second law of diffusion to correct the shortcomings of the Henderson-Pabis model. The Modified Henderson-Pabis model improves previous models by adding the third term of the general solution from Fick's second law of diffusion [2].

The Weibull, Wang-Singh, Vega-Lemus and Proposed models, are all empirical models.

Fig. 1 shows the drying curves and their fit to the Wang-Singh model for 5 of the 17 experiments. The first number in the code corresponds to the relative humidity (20 and 50%), the second is the thickness (2 and 4 mm), and the last is the air velocity (0.8 and 1.2 m/s).

The relative humidity and air velocity affect the drying time; a lower relative humidity and higher velocity provide faster drying times regardless of the sample thickness. Krokida [3] found temperature to be the most important factor in the drying rate, while the velocity and humidity have lesser effects. In that study, the evaluated temperatures were higher (65, 75 and 85 °C).



Figure 1. Drying curves modeling by using the Wang-Singh model Source: The authors

The effective diffusivity  $(D_{eff})$  was determined using eq. (6) - (7) and the least squares analysis for each experiment. Fig. 2 shows the response surfaces for the effective diffusivity,  $D_{eff}$ . Larger effective diffusivities were obtained at lower relative humidities, higher material thicknesses and high air velocities. The desirability function confirms these results. The  $D_{eff}$  values were between 2.01E-10 and 4.38E-9 m<sup>2</sup>/s. Doymaz [19] determined the effective diffusivity values for the convective drying of carrots ranged from 7.76E-10 to 9.34E-10 m<sup>2</sup>/s and argues that this variable increases when air flow and temperature are increased; these are similar to this study's observations (Fig. 2), which account for the decreased RH at an increased temperature (Table 1). Kaya et al. [14] also found that decreasing the RH (temperature increase for a closed system) increased the effective diffusivity values. However, Phoungchandang et al. [20] determined that the effective diffusivity values ranged from 8.34E-11 to 2.77E-10 m<sup>2</sup>/s for dried carrots in a heat pump dryer at 40, 50 and 60 °C with an air velocity of 0.5 m/s. Panagiotou et al. [21] reported Deff values (m<sup>2</sup>/s) ranging from 2.20E-12 to 7.46E-9 for moisture contents between 0.10 and 15.0 on a dry basis and temperatures between 20 and 100 °C. Torres et al. [22] also obtained similar results working with yam in a convective drier.

The *p* value obtained from the ANOVA for the effective diffusivity,  $D_{eff}$ , was less than 0.05 for the regression term. This means that there is a statistically significant relationship between the variables on a 95% confidence level, specifically between the quadratic and interaction terms. The linear terms can be omitted from the model because they would have no significance at this confidence level. This is corroborated by the *p*-values for the regression coefficients in which only the



Figure 2. Effect of RH (%), sample thickness and air velocity on effective diffusivity  $D_{eff}$  (m<sup>2</sup>/s) Source: The authors

second order term (RH \* e) was significant and is effectively an interaction. The *p* value is the specific probability that the observed value of the test statistic, together with all other possible values of the test statistic that are at least as unfavorable as the null hypothesis, will occur [17].

Validation of the model via a residual analysis only happens when the probability curve that shows normal error behavior is observed. Residuals are not far from the line; the variance is homogeneous, and the largest deviation occurred in experiment 9.

The activation energy obtained was 35.50 kJ/mol, and the literature reports activation energies for carrot drying using different equipment and different processing conditions.

According to the energy levels involved and collision theory for reactive molecules, enough energy must be generated to provide the required activation energy and facilitate the reaction. The activation energy itself does not provide information on the reactivity of a given system, only on the temperature dependence of the reaction. Activation Energy activation is also related to the moisture content. The diffusion activation energy increased at lower moisture contents because the interaction between the moisture and solid is generally stronger at lower moisture contents [23].

# 4. Conclusions

All of the thin layer models evaluated could fit the experimental data. The Wang-Singh model exhibited the best fit for the experimental data, based on the criteria of having the highest correlation coefficient ( $R^2$ ) and lowest Chi-squared statistic ( $\chi^2$ ), standard error of the estimate (*SEE*) and root mean square error (*RMSE*) values. A significant effect from the *HR*\**e* interaction was found during the ANOVA for effective diffusivity. The conditions that provide the best evaluated parameters (low *MR* and high  $D_{eff}$ ) are: 20% *HR*, 4 mm thickness and 1.2 m/s air velocity.

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