Effect of Temperature on Drying Kinetics and Quality of Partially Deoiled Chia Flour Wheat Pasta

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Abstract: After chia oil extraction, a by-product, partially-deoiled chia flour (PDCF) with a high nutritional value (high content of fiber, protein, and antioxidant properties) remains. The effect of drying temperature on the quality and drying kinetics of wheat pasta enriched with PDCF was evaluated. Wheat pasta was prepared with different proportions of PDCF (0, 5 and 10%, wheat based) and dried at 45, 55 and 65 °C for 24 hours. Experimental data were fitted to different empirical models. Drying kinetics, drying rate (DR), effective moisture diffusivity (Deff), pasta cooking parameters, microstructure, textural and color were analyzed. The Midilli model had the most suitable performance to describe pasta drying kinetics behaviour. Pasta enriched with PDCF required higher energy input for drying. High drying temperatures had a great impact on enriched pasta quality. Pasta was darker, opaque, irregular, and porous. Cooking time and cooking losses decreased. Hardness was not affected. The strength and microstructure properties of enriched pasta improved at 65 °C, while the non-enriched pasta properties improved at 55 °C.

Keywords: kinetics, chia flour, modeling, pasta quality, drying heat treatment

Abbreviations

- PDCF: partially-deoiled chia flour
- MC: moisture content
- MR: moisture ratio
- t: drying time
- $M_t$: moisture content at any point of drying process
- $M_e$: equilibrium moisture content
- $M_0$: initial moisture
- RMSE: root mean square error
- MR\textsubscript{pre,i}: predicted moisture ratio
- MR\textsubscript{exp,i}: experimental moisture ratio
- N: number of observations
- DR: drying rate
1. Introduction

Chia seed (Salvia hispanica L.) originated in Central and South America. Its present interest to researchers is mainly due to their nutritional composition and their benefits to the human-health. After chia oil extraction, a by-product, partially-deoiled chia flour (PDCF) with a high nutritional value (high content of fiber, protein, and antioxidant properties) remains [1]. Recent years there has been a special growing appeal in the use of this residue to fortified different food products as maize tortillas [2], cookies [3], gluten-free premixes [4] and highly nutritional breads [5].

Pasta is consumed all over the world and is preferred in all age groups. It is a good source of energy, high in complex carbohydrates but low in protein, minerals, vitamins, bioactive compounds and dietary fiber [6]. Pasta can be used as a functional food through the incorporation of healthy components in its formulation. Several studies have shown the viability of fortifying pasta with berries [7], and spirulina [6] beside others. We used PDCF as a source of antioxidants to enhance the nutritional value in pasta and muffins [8-9]. Other researchers [10-11] have used crushed chia seeds for the production of pasta products.

The application of heat air treatments is one of the most important processes in pasta manufacture since it provides the final characteristics of texture and stability during storage. During drying, a phenomenon of mass and energy transport in an unsteady state occurs [12], the initial moisture content (MC) of fresh pasta (31% w.b.) reduces to an appropriate final moisture for preservation (11% w.b.), in order to restructure the protein network that embeds around the starch granules. Pasta drying conditions (temperature, drying time and moisture content) could affect its final quality [13]. In this regards, the study of its optimal drying conditions allows in to manufacture good quality pasta, optimizing drying times, the use of energy and reducing costs. Although these investigations have reported the effect of drying temperature on pasta making process, to the best of our knowledge the effect of drying temperature on PDCF-enriched pasta quality has not been studied so far.

In previous studies we have found that the incorporation of PDCF in the pasta manufacture process impacts on technological quality [8-9]. In general, the amount of material to be added to a food formulation is a compromise between the intended improvement of nutritional quality and the sensory and functional properties of the final product. In this context, the aim of this work was to study the drying kinetics of pasta and the effect of drying temperature on pasta structural characteristics.

2. Materials and methods

2.1 Obtaining PDCF and manufacturing of pasta

Commercial wheat flour (Triticum aestivum) was purchased from Molino San José, José Minetti & CIA Ltda. (Córdoba-Argentina). Chia seeds (Salvia hispanica L.) were purchased in a local market. PDCF was obtained by a screw pressing [14]. A small-scale standardized laboratory procedure was used for pasta manufacture [1]. Pasta was prepared with different concentrations of PDCF (0, 5 and 10% PDCF, respectively, weight flour basis). The 0.0% PDCF sample corresponds to a 100% wheat flour pasta. For each formulation pasta flour, water, and salt (50 g, 22.5 g, and 1.0 g, respectively) were mixed in a Hobart bench top mixer (Hobart Inc., Troy, OH, USA) until the dough had an adequate
consistency for lamination. Dough was divided by hand in appropriate size and was laminated using a pasta home scale size lamination machine (Drago, Inc., China) using a 3-step procedure: hand lamination, up to approximately 10-mm thickness; roll lamination, up to a 5-mm thickness; and final roll lamination to a 2-mm thickness (final pasta thickness). Laminated pasta sheets were cut using a cutting roll (2-mm wide) obtaining the pasta.

2.2 Drying kinetics study

Pasta samples (10 g) were dried on homemade aluminum perforated trays in two steps. The first stage, termed pre-drying, (30 min, 30 °C, forced air ventilation) in a laboratory drier (Memmert Model 600 D060602, Germany). Then, a second stage (termed drying) was used (FAC model CDH4060, Argentina) with moisture control (75%) and at 45, 55 and 65 °C for 24 hours. The weight of samples (quadruplicates) was taken during the pre-drying stage at intervals of 10 min. During the drying stage (second drying step) samples were taken each 15 minutes during the first hour and every 60 min until the 24 hours maximum of drying was reached. Moisture content (MC) was determined according to the official method 44-19 of the American Association of Cereal Chemists (AACC) [15]. The results were expressed in g of water lost/100 g of sample. Dried pasta was stored in airtight bags at room temperature until needed.

2.3 Mathematical modeling

Moisture ratio (MR) as a function of drying time (t) was computed and plotted using MC values for each temperature and determined according equation 1.

\[ MR = \left( M_f - M_e \right)/M_0 - M_e \]  

Where \( M_r, M_f \) and \( M_0 \) are the moisture content (MC, kg water/kg dry mater) at any point of drying process, equilibrium and initial time, respectively. A linear and non-linear regression method was used. The experimental data (MR vs drying time) were fitted to seven mathematical models for thin layer drying curves (Table 1).

Table 1. Parameters and statistical analysis of control pasta experimental models

<table>
<thead>
<tr>
<th>Model</th>
<th>T °C</th>
<th>Constants and coefficients</th>
<th>R²</th>
<th>RMSE</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>a</td>
<td>k</td>
<td>c</td>
</tr>
<tr>
<td>Henderson and Pabis</td>
<td></td>
<td>1.0692</td>
<td>0.0054</td>
<td></td>
</tr>
<tr>
<td>Logarithmic</td>
<td></td>
<td>1.0765</td>
<td>0.0053</td>
<td>0.0099</td>
</tr>
<tr>
<td>Midilli</td>
<td></td>
<td>1.0191</td>
<td>0.0011</td>
<td>6.0437E-06</td>
</tr>
<tr>
<td>Wang and Singh</td>
<td>45</td>
<td>1.4053E-06</td>
<td>-0.0027</td>
<td></td>
</tr>
<tr>
<td>Lewis</td>
<td></td>
<td>0.0051</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Modified page</td>
<td></td>
<td>1.3172</td>
<td>1.3172</td>
<td>0.9952</td>
</tr>
<tr>
<td>Page</td>
<td></td>
<td>0.0009</td>
<td>1.3172</td>
<td>0.9952</td>
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</table>
Table 1. (cont.)

<table>
<thead>
<tr>
<th>Model</th>
<th>Parameter</th>
<th>Constants and coefficients</th>
<th>R²</th>
<th>RMSE</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>T °C</td>
<td>a</td>
<td>k</td>
</tr>
<tr>
<td>Henderson and Pabis</td>
<td>MR = a<em>EXP(-k</em>t)</td>
<td>1.0275</td>
<td>0.0104</td>
<td>0.9869</td>
</tr>
<tr>
<td>Logarithmic</td>
<td>MR = a*EXP(-kt) + b</td>
<td>1.0248</td>
<td>0.0105</td>
<td>0.0036</td>
</tr>
<tr>
<td>Midilli</td>
<td>MR = a<em>EXP(-k</em>t^n) + bt</td>
<td>1.007</td>
<td>0.0026</td>
<td>6.3331E-06</td>
</tr>
<tr>
<td>Wang and Singh</td>
<td>MR = 1 + b<em>t + a</em>t^2</td>
<td>55</td>
<td>1.7156E-06</td>
<td>-0.0032</td>
</tr>
<tr>
<td>Lewis</td>
<td>MR = EXP(-k*t)</td>
<td>0.0102</td>
<td>0.9861</td>
<td>0.0318</td>
</tr>
<tr>
<td>Modified page</td>
<td>MR = EXP(-k*t^n)</td>
<td>0.0096</td>
<td>1.2806</td>
<td>0.992</td>
</tr>
<tr>
<td>Page</td>
<td>MR = EXP(-k*t^n)</td>
<td>0.0026</td>
<td>1.2806</td>
<td>0.992</td>
</tr>
<tr>
<td>Henderson and Pabis</td>
<td>MR = a<em>EXP(-k</em>t)</td>
<td>1.0310</td>
<td>0.0117</td>
<td>0.9837</td>
</tr>
<tr>
<td>Logarithmic</td>
<td>MR = a*EXP(-kt) + b</td>
<td>1.0264</td>
<td>0.0118</td>
<td>0.0043</td>
</tr>
<tr>
<td>Midilli</td>
<td>MR = a<em>EXP(-k</em>t^n) + bt</td>
<td>1.0022</td>
<td>0.0015</td>
<td>9.1736E-06</td>
</tr>
<tr>
<td>Wang and Singh</td>
<td>MR = 1 + b<em>t + a</em>t^2</td>
<td>65</td>
<td>1.7476E-06</td>
<td>-0.0032</td>
</tr>
<tr>
<td>Lewis</td>
<td>MR = EXP(-k*t)</td>
<td>0.0114</td>
<td>0.9829</td>
<td>0.0349</td>
</tr>
<tr>
<td>Modified page</td>
<td>MR = EXP(-k*t^n)</td>
<td>0.0104</td>
<td>1.4243</td>
<td>0.9945</td>
</tr>
<tr>
<td>Page</td>
<td>MR = EXP(-k*t^n)</td>
<td>0.0015</td>
<td>1.4243</td>
<td>0.9945</td>
</tr>
</tbody>
</table>

* MR: moisture ratio; k: drying constant; n, a, b and c are equations constants

To identify the best fit model, the highest coefficient of determination (R²) and the lowest root mean square error (RMSE) were considered. Equation 2 shows the RMSE calculation.

\[
RMSE = \left( \frac{1}{N} \sum_{i=1}^{N} \left( \frac{MR_{pre,i} - MR_{exp,i}}{} \right)^2 \right)^{0.5}
\]

Where \( MR_{pre,i} \) and \( MR_{exp,i} \) are the \( i \)-th predicted and experimental moisture ratio, respectively. \( N \) is the number of observations.

2.4 Drying rate

Equation 3 was used to determine the drying rate (DR).
\[
DR = -(dM_{hs}/dt) = -(M_{t+\Delta t} - M_t)/(t_{t+1} - t_t)
\]

Where, \(M_{t+\Delta t}\) is the moisture content on dry basis (kg water/kg db) at time \((t_i) + \) time difference (min); \(DR\) is drying rate (kg water/kg min) and \(t\) is drying time (min).

2.5 Effective moisture diffusivity

The pasta moisture diffusivity \((Deff)\) was calculated by solving the diffusion equation of Fick’s law for a thin layer. Thus, it was considered that the moisture loss is uniformly distributed on both sides of pasta, the Crank equation (Equation 4) was applied [16-17].

\[
MR = \left(\frac{8}{\pi^2}\right)\sum_{n=0}^{\infty}\left(\frac{1}{(2n+1)^2}\right)\exp\left(-\left((2n+1)^2 \pi^2/4L^2\right)\right)\left(Deff * t\right)
\]

Where, \(MR\) is the moisture ratio (kg water/kg db); \(Deff\) is the effective moisture diffusivity (m²/s); \(t\) is drying time (min); \(L\) is the critical drying distance in (m) (1/2 of pasta thickness). Considering pasta as a thin layer and a drying over long periods of time, equation 4 can be simplified as equation 5.

\[
MR = \left(\frac{8}{\pi^2}\right)\exp\left(-\left(\frac{\pi^2 Deff * t}{4L^2}\right)\right)
\]

2.6 Activation energy

The relationship of temperature vs the diffusion coefficient \((Deff)\) and the drying kinetic constant \((k)\) was determined following the Arrhenius model (Equation 6).

\[
k = k_0 \exp\left(-\frac{Ae}{RT}\right)
\]

The Arrhenius equation was linearized and \(\ln(Deff)\) and \(\ln(k)\) were graphed as a function of \(1/T\), the slope obtained in the line equation was used to calculate activation energy \((A_e)\).

\[
LnDeff = LnD_0 - \left(\frac{A_e}{R}\right) * (1/T)
\]

Where, \(k\) is drying kinetic constant (min-1); \(k_0\) and \(D_0\) are pre-exponential constants (min⁻¹ and m²/s respectively); \(A_e\) is activation energy (kJ/mol); \(T\) is absolute drying temperature (ºK); \(R\) is the universal gas constant (8.314 × 10⁻³ kJ/mol K).

2.7 Pasta cooking parameters

Cooking quality parameters of pasta were evaluated according to method 16-50 of the AACC [15]. Optimum cooking time (OCT), water absorption (WA), weight gain (WG), and cooking loss (CL) were assessed.

2.8 Texture and color analysis

Texture analysis of cooked pasta was performed using an INSTRON Texturometer (Model 3,342, Norwood, MA, USA) equipped with a 500 N cell. Breaking force (BF) of each strand of uncooked pasta was evaluated by the three-point bending test [15]. Firmness and gumminess of cooked optimum time pasta were calculated using Application Study Ref N002/P35 (Stable Micro System, Surrey, UK). A cylindrical probe (AP/35) was used to compressed twice 4
strands of cooked pasta (5 cm long each) at fixed 60% strain; the results were expressed in Newtons. Pasta color was
determined using a Minolta 508d spectrophotometer (Ramsey, NJ, USA) according to method 14-22 [15], which details
each color in the CIE Lab coordinates color space: L* (lightness), a* (red-green) and b* (yellow-blue).

2.9 Textural analysis of pasta images obtained by microscopy

The microstructural characteristics of the inner structure of raw pasta were evaluated using an Olympus LEXT
OLS4000 3D confocal laser scanning microscope (CLSM). The Image J program (National Institutes of Health, USA)
using a second order statistical algorithm called Gray-Level Co-occurrence Matrix (GLCM) was used for textural
analysis of the images. The images obtained in JPEG were corrected to 8 Bit and later binarized in two colors, black
and white, allowing quantified the microstructural differences between the samples. The following dimensionless
parameters were obtained from the analysis of the images: Energy also called Uniformity or Angular second moment
which measures the textural uniformity, entropy which measures the disorder or complexity of an image, homogeneity
also called as Inverse Difference Moment and fractal dimension using the Fractal Box count algorithm in the Image J
software, which provided a numerical parameter that described the morphology and texture of the images with irregular
and complex structures.

2.10 Statistical analysis

A linear and non-linear regression method was used to fit pasta drying kinetics data. The adjusted coefficient of
determination (R²), was used to evaluate the quality of the fit. The results were analyzed using the InfoStat Statistical
Software (Facultad de Ciencias Agropecuarias, UNC, Argentina), significant differences between samples were
evaluated with Analysis of Variance (ANOVA) (p < 0.05). The results were compared using the DGC method [18].
The statistical analysis was executed using SigmaPlot software (v.12.5, SigmaPlot Software, Chicago, IL, USA). Each
analysis was accomplished in three replicates.

3. Results and discussion

3.1 Pasta drying conditions

![Graph showing pasta drying conditions](image-url)
Figure 1. (a) Moisture ratio (MR) variation as a function of drying time at different temperatures (b) Relationship between MR and drying time at different temperatures (c) Experimental vs predicted MR by the Midilli et al. (2002) model at different temperatures (d) Pasta drying rate and drying time relationship at different temperatures.

Figure 1a shows that the highest moisture loss occurred during the first six hours of drying process (360 min). As there is more water available in the pasta at the beginning of the process, the water is more available to be evaporated. Contrary to this, at the end of the process the decrease in moisture content was increasingly slower until reaching equilibrium moisture content after 24 h of the process. Moreover, the MC reduction during the drying time of each sample (0, 5 and 10% PDCF) was influenced by different drying temperatures (45, 55 and 65 °C) (Figure 1b). The higher the drying air temperature was, a greater loss of moisture occurred as a result of the greater heat transfer from the pasta surface to the inner. High temperatures decreased the drying time, since after the first 5 h of drying at 45 °C, after the first 2 hours of drying at 55 °C and after the first hour and a half of drying at 65 °C, the pasta reached acceptable storage moisture values (<12.5%), at which point the moisture content began to decrease at a lower rate.

3.2 Pasta drying mathematical models

The adjusted determination coefficient values (R²) of the models evaluated had a great variability (ranging from 0.258 to 0.999). Root mean square error (RMSE) values were in the range of 0.007 to 0.264 (Table 1 as representative of control sample-other data not shown). The Midilli and Page models showed the best performance, and also high capacity to describe pasta behavior, showing the highest values of the R² and the lowest values of RMSE. However, we did not choose the Page model because the R² values were lower than in Midilli. The Page model has been already considered by other authors as the one with the best fit for a thin layer drying process of pepper slices [17] and yacón [19] among others. However, López-Mejía et al. [20] showed that the Henderson and Pabis and the Logarithmic models described in a better way the pumpkin enriched pasta drying kinetics and the Page model was chosen to represent spinach pasta drying kinetics [21].

The parameters and statistics related to the best fit model (Midilli) were grouped and are detailed in Table 2. The drying air temperature and rate are the variables that significantly affect the “k” parameter in a thin layer drying equation, and is usually related with Deff coefficient. On the other hand, the “n” parameter behaves according to: the relative moisture of the air, the product initial moisture content, the nature of the product and the drying conditions, which shows the internal resistance of the product to the drying process.
Table 2. Constants calculated by Midilli model

<table>
<thead>
<tr>
<th></th>
<th>Constants and coefficients</th>
<th>T°C</th>
<th>a</th>
<th>k</th>
<th>c</th>
<th>n</th>
<th>R²</th>
<th>RMSE</th>
</tr>
</thead>
<tbody>
<tr>
<td>MR = a<em>EXP(-k</em>t^n) + bt</td>
<td></td>
<td>45</td>
<td>1.0191</td>
<td>0.0011</td>
<td>6.0437E-06</td>
<td>1.2978</td>
<td>0.9957</td>
<td>0.0228</td>
</tr>
<tr>
<td>0% PDCF</td>
<td></td>
<td>55</td>
<td>1.007</td>
<td>0.0026</td>
<td>6.3331E-06</td>
<td>1.2816</td>
<td>0.9923</td>
<td>0.0262</td>
</tr>
<tr>
<td></td>
<td></td>
<td>65</td>
<td>1.0022</td>
<td>0.0015</td>
<td>9.1736E-06</td>
<td>1.4323</td>
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<td>0.0205</td>
</tr>
<tr>
<td></td>
<td></td>
<td>45</td>
<td>1.023</td>
<td>0.0011</td>
<td>9.2907E-06</td>
<td>1.2743</td>
<td>0.9951</td>
<td>0.0243</td>
</tr>
<tr>
<td>5% PDCF</td>
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<td>0.0062</td>
<td>1.1776E-06</td>
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<td>1.2367</td>
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<td>0.0237</td>
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<td>10% PDCF</td>
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<td>0.0074</td>
<td>1.1201E-06</td>
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<td>0.0009</td>
<td>6.9312E-07</td>
<td>1.58</td>
<td>0.9994</td>
<td>0.0075</td>
</tr>
</tbody>
</table>

R²: coefficient of determination, k: drying constant; n, a, b and c are equations constants; t: drying time; RMSE: lowest root mean square error

Based on the chosen model, the experimental data (MR exp) were compared with the predicted data (MR pre) by the Midilli model. Figure 1c shows the data of samples grouped in a straight line justifying the choice of the model.

3.3 Drying rate (DR)

As shown in Figure 1c, DR reported two types of behaviors: 1) a sharp increase in DR at the beginning of the process (first two hours of drying) and 2) a subsequent gradual decrease in MC. This variation is due to the fact that the MC was initially high (35% ± 3), in this first stage the pasta surface is saturated with water so there is greater evaporation accelerating the dehydration rate. On the other hand, at the end of this period, the water transfer from pasta inner was increasingly slower than the water vaporization to the air, which is why the drying rate decreased until reaching moisture equilibrium. DR was governed by temperature so higher values were obtained at higher temperatures. Previous researchers have also established a similar behavior in all conditions of temperature (30 to 90 °C) and relative humidity (50% to 80%) [22]. Pasta transformation occurs from a gummy state to a glassy state with a decrease in MC during drying [23]. There are similarities with the results described by starch [24], and strawberries [25].
<table>
<thead>
<tr>
<th>Values with the same letter are not significantly different (p &lt; 0.05) according to the DGC test. PDCF = partially deoiled chia flour.</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Parameter</td>
<td>10ºC</td>
</tr>
<tr>
<td>L*</td>
<td>4.91 ± 0.88</td>
</tr>
<tr>
<td>a*</td>
<td>0.67 ± 0.04</td>
</tr>
<tr>
<td>b*</td>
<td>0.02 ± 0.15</td>
</tr>
</tbody>
</table>

**Table 3.** Effective moisture diffusivity (Deff) of samples at different temperatures (μm²/s). Values are means and standard deviations (n = 3). Cooking characteristics of pasta samples (down).

**Figure 2.** Effective moisture diffusivity (Deff) of samples at different temperatures (μm²/s). Values are means and standard deviations (n = 3). Cooking characteristics of pasta samples (down).
3.4 Effective moisture diffusivity (Deff)

The Deff for each sample against the different temperatures evaluated (45, 55 and 65 °C) are shown in the Table 3. The values ranged from $5.99 \times 10^{-09}$ to $2.36 \times 10^{-08}$ and presented high $R^2$ values. These results were in concordance within the Deff range ($10^{10}$ to $10^{-7}$) found from other authors [8,21] for pasta samples. This variation of Deff ranges found in pasta products is given by the effect of experimental procedures conditions, model proposed, sample thickness, composition and structure [26]. A clear dependence of Deff and drying temperature was found. Deff values increased significantly when the drying temperature was higher. The increase in temperature contributed to a greater water diffusion favoring the drying process. The aforementioned studies also reported an increase in Deff as a function of drying temperature. If we analyze the samples for the addition of PDCF, the Deff values at 45 °C decreased as the PDCF was higher due to its water retention capacity. However, the PDCF behavior was antagonistic to that observed at 45°C when increasing the drying temperature (55 and 65 °C), causing a decrease in PDCF water retention capacity allowing the release of retained water and increasing the diffusion of moisture inside pasta.

3.5 Activation energy

The activation energy for non-enriched pasta (0% PDCF) was 36.64 kJ/mol, for 5% PDCF pasta was 42.45 kJ/mol and for 10% PDCF pasta was 59.06 kJ/mol. All samples obtained good fits ($R^2$: 0.92, 0.99, 0.99, respectively) (Figure 2). From this point, PDCF pasta needed more energy to remove moisture from its inner for a typical drying process (45, 55 and 65 °C). Other authors reported $A_e$ values in a range from 19.34 to 37.71 kJ/mol for spinach pasta [21] and rice noodles. In fact, the values may vary due to the conditions of processing, storage and the source of raw materials [27].

![Figure 2](image-url)  
Figure 2. Linear regression analysis for ln Deff as a function of 1/T by curve fitting. The slope obtained in the equation of the line gives the value of activation energy

3.6 Pasta cooking parameters

3.6.1 Effects on texture and color of raw pasta

To evaluate the impact of pasta drying temperatures, the main quality parameters analyses of uncooked pasta...
(color and breaking force) were carried out (Table 3). Regarding color values and analyzing each sample as a separate entity, the results indicate that as the drying temperature increases, the $L^*$ parameter (brightness) significantly ($p < 0.05$) decreases from 78.09 to 70.36 for control sample, from 73.97 to 65.77 for 5% PDCF and from 72.20 to 64.10 for 10% PDCF. This implies that drying at 65 °C leads to pasta surface looks opaque. The redness parameter $a^*$ ranged from 1.20 and 2.62, and opposite to $L^*$, the highest values were obtained by the samples dried at 65 °C. The increase in redness could be linked to non-enzymatic browning reactions as a result of drying at high temperatures [28]. Control sample dried at 65 °C pointed out negative yellowness ($b^*$) values. In case of fortified samples, as the original color of PDCF is brown, the difference in yellowness was not significantly detected ($p < 0.05$).

BF is an indication of how strong the pasta is during handling and storage. The results ranged from 1.48 to 3.85 (Table 3), control sample dried at 55 °C presented the highest values being the most resistant sample compared to the rest. Gull et al. [8] also observed that spaghetti dried at 50 °C with 70% RH showed the best BF values. In case of enriched pasta (5% PDCF), no significant differences were found between them, but a significant increase was observed for 10% PDCF sample dried at 65 °C. Another study [22] pointed out that although they have not found significant differences in the resistance to brake to their pasta samples dried at 50, 70 and 85 °C, the sample dried at very high temperature showed more resistance. They argued that drying at high temperatures favors the formation of disulfide bonds, increasing the strength of protein network and pasta structure.

### 3.6.2 Effects on quality properties of cooked pasta

Table 3 also shows the quality properties of cooked pasta dried at different temperatures. Although the BF of control pasta dried at 55 °C was the highest (Table 3), the drying temperature didn’t affect significantly ($p < 0.05$) the final hardness of cooked pasta in any sample. The firmness values obtained varied between 6.34 and 7.19 N.

Optimum cooking time (OCT) varied from 9.42 to 12.75 min, this parameter decreased as the drying temperature increased. Cooking loss (CL), weight gain (WG) and water absorption (WA) decreased as a result of increasing drying temperature, due to a higher porosity of pasta dried at higher temperatures (Figure 3). This tendency found is in concordance to those reported in Chinese noodles [29].
Figure 3. GLCM of confocal images of inner pasta dried at 45, 55 and 65 °C, magnification of 428x. The arrows indicate water evaporation pores. (a) 0%PDCF pasta (b) 5%PDCF pasta (c) 10%PDCF pasta

3.6.3 Analysis of pasta microstructure dried at different temperatures

The microphotography of the internal microstructures of dried pasta (Figure 3) revealed an open and porous...
structure with the presence of empty spaces, indicating evaporation, water transport and its replacement by air during drying. In general, it seems that the pore size of pasta seems to increase as the pasta is dried at higher temperatures (Figure 3, left to right). Drying at low temperatures (45 °C) lead to a slow evaporation and therefore the transport of water is less from pasta inner to the outside, resulting in a structure with smaller pores. This may have occurred because the high temperatures could cause an increase in water vapor pressure giving as a result in a higher mass transfer and thus pore growth during drying. However, the internal structure of control sample dried at 55 °C (Figure 3a), doesn’t appear to be porous but rather a more consolidated, compact and dense protein matrix with irregular and almost imperceptible pores. In contrast of this, microphotographs of pasta dried at 65 °C appear to be a porous and brighter structure, demonstrating the presence of a denser and firmer protein network. Other researchers [30] have also found that drying at high temperatures caused the formation of larger and more heterogeneous pores and that the level of gelatinization during the dehydration process produced a great deformation and swelling of the dispersed phase (starch granules). The level of gelatinization during this process is regulated by the denaturation of gluten proteins. These results, combined with the results of BF, firmness and OCT (Table 3), confirmed that the number and size of the pores was crucial to determine the characteristics and quality of pasta.

3.6.4 Textural analysis of confocal images using GLCM

The effect of dried temperature on pasta structure is not quite clear from the data shown in Figure 3. However, although some parameters are not statistically significant, some changes can be inferred. The incorporation of PDCF made the pasta internal structure more tortuous and irregular compared to control pasta (confocal images Figure 3). No significant differences were observed in terms of uniformity, homogeneity and entropy values between the enriched pasta dried at different temperatures, but significant differences were found in control pasta. During drying control pasta at 55 °C, its internal structure was affected becoming more compact and complex (Entropy: 0.90 to 1.07), uniformity and homogeneity decreased significantly ($p < 0.05$) from 0.44 to 0.39 and from 0.91 to 0.87, respectively. High values of entropy and low values of uniformity are related to a high state of disorder and a rough surface of the samples. Contrary to this, samples with low values of entropy and high values of uniformity are characterized by smooth textural surfaces. The images obtained by confocal microscopy (Figure 3) in the case of control pasta dried at 55 °C corroborate these results since its internal structure doesn’t appear to be so porous but rather compact with irregular and almost imperceptible pores. However, drying at 65 °C significantly decreased the DF values of all samples, resulting in less tortuous internal structures. Other authors have demonstrated that drying at high temperatures (60-100 °C) makes the protein network more continuous and dense [31].

4. Conclusions

The Midilli model presented a satisfactory fit to the experimental data to describe the pasta drying process. The $DR$ and the $Deff$ coefficients in samples with PDCF were governed by temperatures (55 and 65 °C). As the temperature increases, there is a greater transfer of heat from the pasta surface to its inner, accelerating the internal evaporation of the sample. However, control pasta obtained better drying rate values at 55 °C. Fortified pasta required higher $A_e$ (45 and 59.06 kJ/mol) for a typical drying process be occur compared to non-fortified pasta (36.64 kJ/mol). Drying temperature significantly affected OCT, CL, WG and WA, as a result, pasta dried at higher temperatures appeared to have greater internal porosity (larger pores) while the pores observed in samples dried at lower temperatures appeared to be smaller. The incorporation of PDCF weakened the pasta internal structure when dried at low temperatures. However, its strength and microstructure properties improved when was dried at 65 °C and non-enriched pasta presented better resistance properties, a more compact and dense structure when were dried at 55 °C. Nevertheless, these differences were simplified after cooking. In this sense, the drying temperature affects the strength of dry pasta and not so much the final strength of pasta after cooking.
Author contribution

C.A. was responsible for conducting the experiments, analyzing the data, writing the original draft, visualization, and editing the manuscript. A.A. contributed to the conceptualization, reviewing the manuscript, and funding acquisition. R.B. was in charge of conceptualization, experimental design, funding acquisition, supervision, and manuscript revision.

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Data availability statement

Research data of this study could be made available upon reasonable request.

Conflict of interest

The authors declare no competing financial interest.

References


