

**Research Paper**

# **Thermal Inactivation of *Salmonella* Agona in Low-Water Activity Foods: Predictive Models for the Combined Effect of Temperature, Water Activity, and Food Component**

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**ABSTRACT**

*Salmonella* can survive in low-moisture, high-protein, and high-fat foods for several years. Despite nationwide outbreaks and recalls due to the presence of *Salmonella* in low-moisture foods, information on thermal inactivation of *Salmonella* in these products is limited. This project evaluated the impact of water activity ( $a_w$ ), temperature, and food composition on thermal inactivation of *Salmonella enterica* serovar Agona in defined high-protein and high-fat model food matrices. Each matrix was inoculated with *Salmonella* Agona and adjusted to obtain a target  $a_w$ , ranging from 0.50 to 0.98. Samples were packed into aluminum test cells and heated (52 to 90°C) under isothermal conditions. Survival of *Salmonella* Agona was detected on tryptic soy agar with 0.6% yeast extract. Complex influences by food composition,  $a_w$ , and temperature resulted in significantly different ( $P < 0.05$ ) thermal resistance of *Salmonella* for the conditions tested. It was estimated that the same point temperatures at which the  $D$ -values of the two matrices at each  $a_w$  (0.63, 0.73, 0.81, and 0.90) were identical were 79.48, 71.28, 69.62, and 38.42°C, respectively. Above these temperatures, the  $D$ -values in high-protein matrices were larger than the  $D$ -values in high-fat matrices at each  $a_w$ . Below these temperatures, the inverse relationship was observed. A correlation between temperature and  $a_w$  existed on the basis of the level of fat or protein in the food, showing that these compositional factors must be accounted for when predicting thermal inactivation of *Salmonella* in foods.

**Key words:** *D*-value; Low water activity; Pet food; *Salmonella*; Thermal inactivation

*Salmonella enterica* serovars are estimated to cause 1.2 million illnesses and 450 deaths per year in the United States (8). Several recent recalls and outbreaks involved *Salmonella* Agona, the organism investigated in this study (6, 7, 17, 24), which has relatively high thermal resistance (37, 39). In recent years, there have been multiple outbreaks and recalls in low-water activity ( $a_w$ ) products, such as cereal, chocolate, dog food, flour, nuts, nut butters, and spices (6, 7, 17). Foods are considered to be low  $a_w$  if below the  $a_w$  of 0.83. Low- $a_w$  foods do not typically support the growth of microorganisms. However, it is well documented that *Salmonella* can survive for an extended length of time in a variety of low- $a_w$  foods and cause illness if a low number of cells are present (15, 22, 23, 25, 26, 33, 37). Also, *Salmonella* has an increased thermal resistance as  $a_w$  decreases, which poses a significant challenge to processors

of low- $a_w$  foods (2, 3, 14, 16, 22, 27, 41, 44). In addition, low- $a_w$  food products, such as flour and peanut butter, may form microenvironments with increased pathogen resistance (27, 29).

Commodity-specific research has been conducted on thermal inactivation of *Salmonella* in low-moisture foods, such as powdered whey protein, oregano and mint oil, wheat flour, and other food products with a wide temperature range from 56 to 140°C (11, 36, 37, 41, 44). Rachon et al. (36) found that *Salmonella* was more resistant in a meat powder with high-protein content than in a confectionary product with high-carbohydrate content. Although studies reveal that *Salmonella* survival and inactivation in commodity-specific foods is influenced by product composition,  $a_w$ , and temperature, it does not provide insight into *Salmonella* inactivation in complex food products formulated with multiple ingredients.

Thermal inactivation models and predictive models such as log-linear and Weibull models have been developed for some *Salmonella* serovars in various foods (5, 11, 18, 20, 32, 41). These models provide reliable data to predict the thermal kinetics of different *Salmonella* serovars. However, thermal inactivation models for *Salmonella* in complex food

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products with multiple ingredients examining the influence of temperature and  $a_w$  are limited.

In summary, recalls and outbreaks of *Salmonella* in low- $a_w$  foods have identified a knowledge gap pertaining to the processing safety of these low- $a_w$  commodities. Thermal inactivation data on *Salmonella* in complex products with multiple ingredients, such as biscuits, crackers, cookies, or other low- $a_w$  foods, are limited. Further research on *Salmonella* thermal inactivation models during low- $a_w$  food processing is required. In addition, evaluation of the potential use of isothermal data to predict microbial inactivation under dynamic processing conditions is necessary (2, 30, 42). The objective of this project was to develop thermal inactivation models for *Salmonella*, as influenced by temperature and  $a_w$ , in high-protein and high-fat low- $a_w$  food matrices with formulations based loosely on dog biscuits and snack crackers, respectively.

## MATERIALS AND METHODS

**Salmonella strain.** *Salmonella enterica* serovar Agona (strain 447967, associated with a puffed cereal outbreak, Arkansas Regional Laboratory, Jefferson, AR) was stored as a frozen stock. Working cultures were streaked on tryptic soy agar with 0.6% yeast extract (TSAYE; Difco, BD, Franklin Lakes, NJ) and incubated for 24 h at 37°C. Working cultures were kept at 4°C for the duration of the study and transferred monthly to fresh media.

**Preparation of inoculum.** Inoculum was prepared by using the methodology of Keller et al. (22). An isolated colony was aseptically harvested from the working culture and inoculated into 10 mL of tryptic soy broth supplemented with 0.6% yeast extract (TSBYE; Difco, BD) and incubated for 24 h at 37°C. A 100-µL aliquot of culture was spread plated on the surface of TSAYE plates and incubated for 24 h at 37°C. Bacterial cells were harvested by adding 1 mL of buffered peptone water (BPW; Difco, BD) per plate to the agar surface, which was then gently scraped with a disposable sterile "L-shaped" spreader (Fisher Scientific, Pittsburgh, PA) to suspend cells. The cell suspension was pipetted into a sterile conical tube (Fisher Scientific, Fair Lawn, NJ). Each plate yielded approximately 0.5 mL of inoculum with a concentration of approximately 11 log CFU/mL.

**Background microflora evaluation.** Whole wheat flour (Pillsbury, General Mills, Minneapolis, MN), soy protein (GNC Holdings, Inc., Pittsburgh, PA), and soybean oil (Gordon Food Service, Wyoming, MI) were used as ingredients to create the model food matrices. Background levels of microflora in untreated ingredients were determined prior to formulation. Ten-gram samples of each ingredient, flour, protein, or oil, were independently weighed into 710-mL (24-oz) sterile Whirl-Pak bags (Nasco, Fort Atkinson, WI) with 90 mL of TSBYE. Samples were soaked for 1 h at ambient conditions before being serially diluted in BPW and spread plated in duplicate on TSAYE and xylose lysine desoxycholate agar (XLD; Difco, BD) plates. A composite matrix of the three ingredients was weighed out into a sterile 710-mL (24-oz) Whirl-Pak bag and thermally treated at 85°C to evaluate spore background. After heating, 10-g samples were aseptically sampled and enumerated, as previously stated, on TSAYE and XLD.

**Formulation of food matrices.** Ingredients were stored in a 37% relative humidity chamber at ambient temperature conditions

throughout the study to ensure a consistent starting  $a_w$ . The ingredients listed previously were used to formulate the two model food matrices. To create the high-protein matrix, 24 g of flour, 10 g of protein, and 2.4 mL of oil were combined in a 710-mL (24-oz) Whirl-Pak bag. To obtain the high-fat matrix, 24 g of flour, 2.4 g of protein, and 10 mL of oil were combined in a 710-mL (24-oz) Whirl-Pak bag. Formulations were mixed via alternately hand massaging for 1 min and stomaching (model IUL, Neutec Group Inc, Farmingdale, NY) for 30 s, three times each. A representative sample of each model food was sent out to a commercial test laboratory for proximate analysis.

**Inoculation of matrices.** Inoculum, diluted with additional BPW (0.4 to 20 mL total), was added to each food matrix to achieve a target  $a_w$  of 0.50, 0.60, 0.70, 0.80, 0.90, and 0.98. The amount of BPW needed to achieve the target  $a_w$  was determined in a separate experiment. The inoculated matrices were mixed by alternating, three times each, hand massaging for 1 min, and stomaching 30 s, to obtain a homogenous dough mixture. Each preparation was then allowed to equilibrate in a sealed Whirl-Pak bag at ambient conditions (23 ± 2°C) for 1 h to simulate a resting period that is common for doughs.

**Sample preparation and initial product measurements.** To minimize sample  $a_w$  changes due to environmental relative humidity that may have differed from the target  $a_w$ , any postinoculation sample manipulation occurred in a humidity controlled glove box (41) preequilibrated to the target sample  $a_w$ . The  $a_w$  of the samples was determined by using the  $a_w$  meter (AquaLab 4TE Duo, Decagon Devices, Pullman, WA), which was calibrated prior to use with appropriate  $a_w$  standards. Three  $a_w$  measurements were made throughout the sample preparation period to verify that sample  $a_w$  remained at the target level. Three initial 1-g samples were serially diluted by using BPW, spread plated onto TSAYE and XLD plates, which were then incubated at 37°C for 48 h. Samples (0.7 to 1 g) for thermal analysis were packed into sterile aluminum test cells (Washington State University, Pullman) (9) to eliminate any headspace in the cell and sealed with antiseize lubricant (Anti-Seize Technology, Franklin Park, IL) to achieve uniform and consistent heating of inoculated samples.

**Thermal treatment and microbial enumeration.** Test cells were heated in an oil bath (Neslab RTE-211, Thermo Fisher Scientific, Newington, NH) at isothermal temperatures (52 to 90°C) on the basis of the  $a_w$  of the samples to obtain  $D$ -values. Four temperatures were evaluated at each  $a_w$ . During treatment, a Fluke 50-Series II model 52 thermometer (Fluke, Singapore), calibrated against a National Institute for Standards and Technology traceable thermometer, was used to monitor the temperature of the oil and the product using a thermocouple positioned in the oil and a thermocouple fed into the center of a filled test cell, respectively. This test cell was also used to measure the come-up time, defined as the time required for sample temperature to reach -0.5°C of the target set point temperature. Triplicate subsamples were used for each dwell time ( $n \geq 6$ ). After treatment, samples were immediately cooled in an ice water bath for at least 1 min. A sterile spatula (Fisherbrand, Fisher Scientific, Waltham, MA) was used to aseptically transfer the sample from the test cell into a 207-mL (7-oz) sterile Whirl-Pak filter bag. The sample was diluted with BPW to create an initial 1:10 dilution and hand massaged for approximately 30 s before further serial dilutions were performed. Aliquots were spread plated on TSAYE, in duplicate, and

incubated at 37°C for 48 h. Counts were expressed as log CFU per gram.

**Model regression.** Prior to model regression, the come-up time was subtracted from the total processing time. Log-linear and Weibull (43) models were fit to *Salmonella* inactivation data sets for each temperature- $a_w$  combination. The shape parameter of the Weibull model was tested to determine if the value was significantly different from a value of 1 (which would represent a log-linear case). Details of the methodology and results of Weibull model are presented in the supplemental material.

The log-linear primary inactivation model (equation 1) was used to describe the isothermal inactivation kinetics

$$\log\left(\frac{N}{N_0}\right) = \frac{-t}{D_{T, a_w}} \quad (1)$$

where  $N$  and  $N_0$  (CFU/g) are the *Salmonella* populations at times  $t$  and 0, respectively,  $t$  (min) is the isothermal treatment time, and  $D_{T, a_w}$  (min) is the decimal reduction time at temperature  $T$  (°C) and  $a_w$  is  $a_w$ . The survivor ratio,  $\log N/N_0$ , was first normalized so that the data would have the same beginning point at time zero.

For each matrix and target  $a_w$  combination, a Bigelow secondary model (equation 2) was used to describe the effect of temperature on *Salmonella* thermal resistance

$$D_{T, a_w} = D_{T_{ref}, a_w} \times 10^{\left(\frac{T_{ref}-T}{z_T}\right)} \quad (2)$$

where  $D_{T_{ref}, a_w}$  (min) is the decimal reduction time at the reference temperature  $T_{ref}$  (°C),  $a_w$  is the mean  $a_w$  for the data set, and  $z_T$  (°C) is the temperature required for a log cycle change in the  $D_{T, a_w}$ -value. The reference temperature was fixed at 75°C because 75°C was frequently used as an isothermal treatment temperature throughout this study. Any variation between  $a_w$  within a target  $a_w$  was assumed to be negligible.

A modified Bigelow secondary model (equation 3), similar to a model described by Gaillard et al. (13), was used to account for the effect of temperature and  $a_w$  for the high-protein and high-fat food matrices

$$D_{T, a_w} = D_{ref} \times 10^{\left[\left(\frac{T_{ref}-T}{z_T}\right) + R_{aw} (a_{wref} - a_w)\right]} \quad (3)$$

where  $D_{ref}$  (min) is the decimal reduction time at the reference temperature and  $a_w$ ,  $T_{ref}$  (°C) and  $a_{wref}$ , respectively, and  $R_{aw}$  is total potential log cycles in the  $D_{T, a_w}$ -value contributable to  $a_w$ . The reference temperature and  $a_w$  were fixed at 75°C and 0.90, respectively. For model estimation, the measured  $a_w$  for each replication was used.

Qualitative and quantitative measurements of model fit were used to evaluate the efficacy of the model fit. The root mean square error (RMSE; equation 4) was used to estimate model error

$$RMSE = \sqrt{\frac{\sum (\log(N/N_0)_{predicted} - \log(N/N_0)_{observed})^2}{n-p}} \quad (4)$$

where  $n$  is the number of samples,  $N$  and  $N_0$  (CFU/g) are the *Salmonella* populations at times  $t$  and 0, respectively, and  $p$  is the number of model parameters. The Durbin-Watson autocorrelation test was applied to the residuals of the primary model regressions. To evaluate the adequacies of secondary models,  $D_{T, a_w}$  residuals were inspected for biases correlated with time, temperature, and  $a_w$ . Models with a normal distribution of residuals and low RMSE were considered to have good fit.

All model estimations were performed by using a one-step regression; that is, all primary and secondary model parameters were estimated simultaneously by using MATLAB nonlinear fitting tools (nlinfit; version 2015a; MathWorks, Natick, MA). Statistical differences between model parameters were tested by using the Welch's  $t$  test ( $\alpha = 0.05$ ).

## RESULTS AND DISCUSSION

Background levels of native microflora were lower than detectable limits in oil and protein powder ( $\leq 0.7$  log CFU/g), and were 2.74 log CFU/g in flour. The background microflora in the composite high-protein matrix after the 85°C thermal treatment was 1.40 log CFU/g on TSAYE and consisted of white, dry colonies resembling typical *Bacillus* species. There were no colonies detected on XLD ( $\leq 0.7$  log CFU/g). Proximate analysis results ( $n = 3$ ) showed that the high-protein and high-fat matrices had carbohydrate:protein:fat (weight ratio) compositions of  $46.8 \pm 2.4:33.8 \pm 1.0:9.3 \pm 0.4$ ; and  $49.0 \pm 0.3:16.7 \pm 0.1:26.7 \pm 0.3$ , respectively.

The temperature come-up time of the sample within an aluminum test cell was approximately 90 to 120 s. Postinoculation and postresting *Salmonella* populations in high-protein and high-fat matrices were  $>7$  log CFU/g on both TSAYE and XLD.

In the Weibull model, more than half of the resultant estimated shape parameters for the high-fat and high-protein survival curves were not significantly different from a value of 1 ( $P < 0.05$ ), indicating constant slope (i.e., linearity). In addition, a visual inspection of the log-linear model residuals, looking for random and normally distributed residuals, did not preclude the log-linear model. Therefore, only log-linear models were further evaluated and reported in this article. Results of Weibull model parameter estimation are presented in the supplemental material.

The isothermal inactivation data obtained in this research was fitted to the log-linear model. Figure 1 shows the thermal inactivation data fitted by the log-linear model under different  $a_w$  and temperatures in the high-protein and high-fat matrices. The RMSEs of primary model fits for the high-protein matrix ranged between 0.24 and 0.91 log CFU/g, with 20 of 24 RMSEs  $\leq 0.60$  log CFU/g (Supplemental Table S1). The RMSEs of primary model fits for the high-fat matrix ranged between 0.29 and 1.15 log CFU/g, with 13 of 24 RMSEs  $\leq 0.60$  log CFU/g (Table S2). Higher temperatures led to a shorter  $D$ -value at each  $a_w$  level in both matrices. For example, in the high-protein matrix formulated at  $a_w$  0.90, isothermal inactivation treatments at 60, 62.5, 65, and 70°C, resulted in  $D$ -values of 59.72, 29.33, 10.86, and 2.75 min, respectively (Table S1). Similar trends were also found in other thermal studies in literature (12, 31, 34, 36, 44). In addition, a correlation was found between  $a_w$  and  $D$ -value in both high-protein and high-fat matrices; that is, a higher  $a_w$  resulted in a smaller  $D$ -value under the same treatment temperature. For example, the  $D$ -values were 15.46, 3.95, and 1.07 min at 80°C in 0.50, 0.63, and 0.73  $a_w$  high-fat matrices, respectively (Table S2). Results were consistent with the findings of others (12, 21, 44), exhibiting the influence of  $a_w$  on thermal resistance.

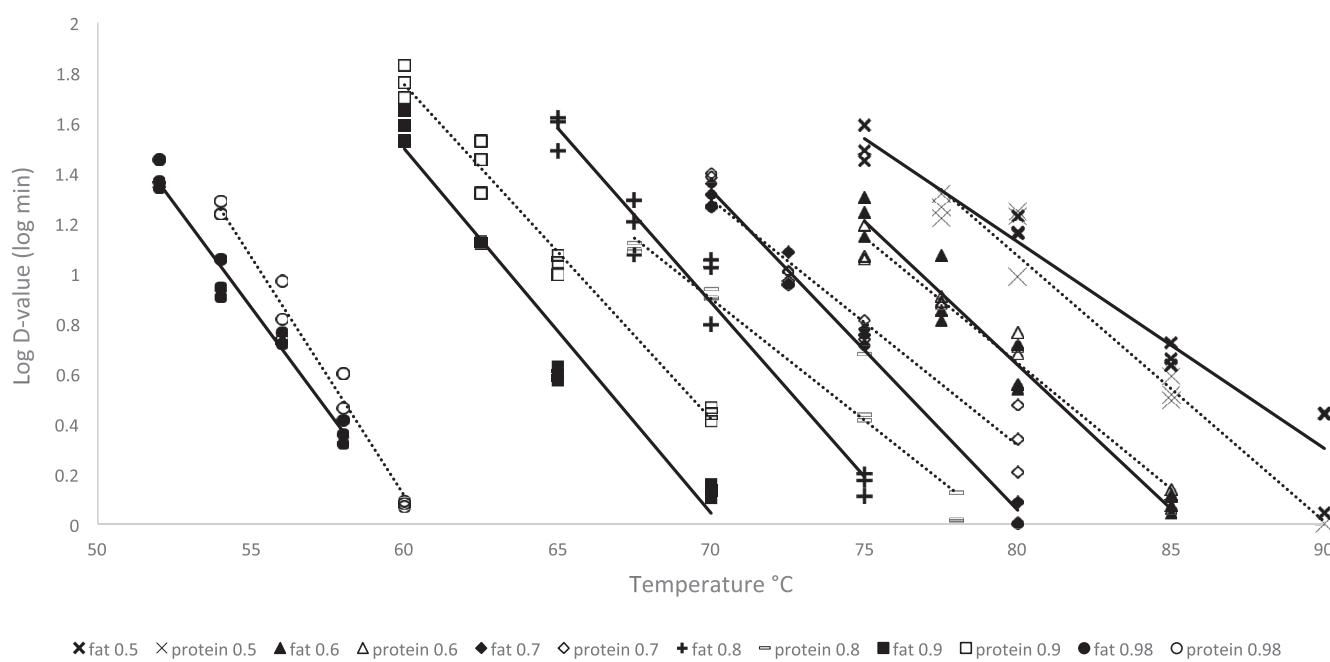


FIGURE 1. Comparison of *Salmonella Agona* survival between high-protein and high-fat matrices at  $a_w$  0.50, 0.63, 0.73, 0.81, 0.90, and 0.98; dotted line, high-protein matrices; solid line, high-fat matrices.

The effect of temperature, quantified through the  $z_T$ -value, was not uniform across the tested  $a_w$  range (Table 1). The  $z_T$ -value was largest at the lowest  $a_w$  and gradually decreased with the increase of  $a_w$  in the high-fat matrix (Table 1). However, a similar clear trend was not found in the high-protein matrix (Table 1). With the exceptions of  $a_w$  0.50 and 0.98,  $z_T$ -values in the high-protein matrix were larger than the  $z_T$ -values in the high-fat matrix at the same  $a_w$ .

Additionally, the  $z_T$ -values within a target  $a_w$ , but across products, were significantly different. As a result, the secondary models are not parallel, and there is an intersecting temperature in which the  $D$ -values are the same (Fig. 1). The implication of this result is that the thermal resistance of *Salmonella* is not universally greater in one product over the other, and there are practical considerations when the intersecting temperature occurs during the normal thermal inactivation range. At 77.43, 79.48, 71.28, 69.62, 38.42, and 62.90°C, the predicted  $D$ -values of the two matrices were the same at  $a_w$  0.50, 0.63, 0.73, 0.81, 0.90, and 0.98, respectively. At temperatures above these values, predicted  $D$ -values in the high-protein matrix were larger than the  $D$ -values in the high-fat matrix for  $a_w$  0.63 to 0.90; however, at  $a_w$  of 0.50 and 0.98, predicted  $D$ -values for the high-fat matrix were larger than  $D$ -values for the high-protein matrix at points greater than the intersection temperatures. At a temperature below the intersection points, the opposite relationships were observed. Because the treatment temperatures in this study ranged from 52 to 90°C, caution should be observed in assuming that the  $D$ -values in the high-protein matrix are always larger than in the high-fat matrix at  $a_w$  from 0.50 to 0.81. However, it is justifiable to conclude that high-protein matrix had larger  $D$ -values than high-fat at  $a_w$  0.9 from 52 to 90°C, because the

same point temperature of 38.42°C was below any practical inactivation temperature.

The inconsistency of trends may be attributed to the formulations of the two model food matrices. As described by Li et al. (29), high-fat content may have a protective effect on *Salmonella* present in local microenvironments during thermal inactivation. Also, Juneja and Eblen (19) described that increased fat levels in beef could result in greater  $D$ -values at a specific temperature. Although a number of studies focused on the protective effect of fat on microorganism thermal inactivation (1, 31, 40), other publications have demonstrated conflicting reports with respect to the protective effect of fat (21, 28, 32). The results in this study suggest that a higher fat content may affect the  $D$ -values and  $z_T$ -values in food matrices, but this trend was

TABLE 1.  $z_T$ -values calculated by Bigelow model explaining temperature in high-protein and high-fat matrices

Actual $a_w$ (avg)	$D_{75}$ (min) <sup>a</sup>	$z_T$ -value (°C) <sup>a</sup>	RMSE (log CFU/g)
<b>High protein</b>			
0.50	35.34 (0.87)	9.78 (0.13)	0.41
0.63	15.02 (0.44)	9.46 (0.17)	0.57
0.73	6.09 (0.09)	10.4 (0.20)	0.65
0.81	2.41 (0.03)	9.85 (0.12)	0.57
0.90	0.57 (0.02)	7.42 (0.12)	0.34
0.98	0.001 (0.0002)	5.08 (0.09)	0.76
<b>High fat</b>			
0.50	37.7 (1.56)	11.46 (0.32)	0.59
0.63	16.16 (0.39)	8.72 (0.14)	0.77
0.73	4.77 (0.06)	7.8 (0.09)	0.55
0.81	1.47 (0.03)	7.17 (0.08)	0.65
0.91	0.24 (0.01)	7.01 (0.08)	0.49
0.98	0.004 (0.0005)	6.03 (0.10)	0.83

<sup>a</sup> Standard error is reported in parentheses.

TABLE 2.  $z_T$  and  $R_{a_w}$  calculated by global model (Bigelow-type secondary model) in high-protein and high-fat matrices

	High-protein matrices	High-fat matrices
$T_{ref}$ (°C)	75	75
$a_w$ ref	0.9	0.9
$D_{ref}$ (min)	0.99	0.61
$z_T$ (°C)	8.99	8.27
$R_{a_w}$	4.38	5.41
$D$ error (min)	0.02	0.02
$z_T$ error (°C)	0.08	0.07
$R_{a_w}$ error	0.06	0.06
RMSE (log CFU/g)	0.65	0.76

not consistent. These conflicting reports, as well as results here, indicate that more studies should be performed to further investigate effects of high-fat and high-protein formulations on thermal inactivation parameters.

The global model that included parameters accounting for the effect of time, temperature, and low  $a_w$  range of 0.50 to 0.91 (Table 2 and Figs. 2 and 3), was constructed by using equation 3. The authors chose to modify a similar model by Gaillard et al. (13) to include a new term,  $R_{a_w}$ , in place of  $z_{a_w}$  (13).  $R_{a_w}$  describes the total potential log cycles in the  $D_{T, a_w}$ -values attributable to  $a_w$ . It is more appropriate to use  $R_{a_w}$  to interpret the variable describing  $a_w$  as a total potential effect because, unlike temperature that has no upper limit,  $a_w$  is defined as a finite value ranging from 0 to 1. From Table 2, it can be seen that the total potential log cycles in  $D_{T, a_w}$ -values are 4.38 min for the high-protein matrix. Because the  $D_{75^\circ\text{C}, 0.9} = 1.0$  min, the theoretical  $D_{75^\circ\text{C}, 0.0} \sim 10,000$  min, but in no event higher, because the lower  $a_w$  boundary (zero) has been reached.

When data from the  $a_w$  of 0.98 were incorporated into the global model extending the  $a_w$  from 0.50 to 0.98, predicted  $D$ -values at 75°C were exceedingly larger (~303 times larger) than the corresponding measured  $D$ -values at

0.98  $a_w$  for the high-protein and the high-fat matrices. This finding suggests that there is a potential transitional  $a_w$  point between a high-moisture model and a low-moisture model. Further, because the 0.98  $a_w$  level would be considered high  $a_w$ , this result indicates that caution should be observed before applying thermal inactivation models developed for low- $a_w$  foods to those with high  $a_w$ .

Because tertiary models describing the  $z_T$ -value as an effect of  $a_w$  would only have five data points, one for each  $a_w$  included in the global model, there were insufficient temperature- $a_w$  combinations tested to characterize the effects of temperature and  $a_w$  ( $R_{a_w}$ -value) as anything but linear, despite the evidence that the effect of temperature ( $z_T$ -value) was nonuniform across the tested  $a_w$  range and without risking overparameterization of resultant models. Nevertheless, the fit of the current model form was satisfactory because RMSE values for the high-protein (0.65 log CFU/g) and high-fat (0.76 log CFU/g) matrices were relatively low (Table 2).

Each parameter for the global high-protein and high-fat models were significantly different. Not only was the relative scale of *Salmonella* thermal resistance different, but the models suggest that the effect of temperature and  $a_w$  on *Salmonella* thermal resistance is characteristically different. This phenomenon would be expected if two completely different low-moisture products were examined, for example, dehydrated apples and orange jam, but the two matrices examined within this experiment were composed of identical ingredients, tested at identical conditions, and only varied with the relative composition of ingredients. This suggests that differences in product formulation may result in substantially different inactivation kinetics. Although this study does not provide insight on the sensitivity of formulation changes, it does indicate that a high-protein and a high-fat characterization of a product consisting of identical ingredients should be treated as two completely

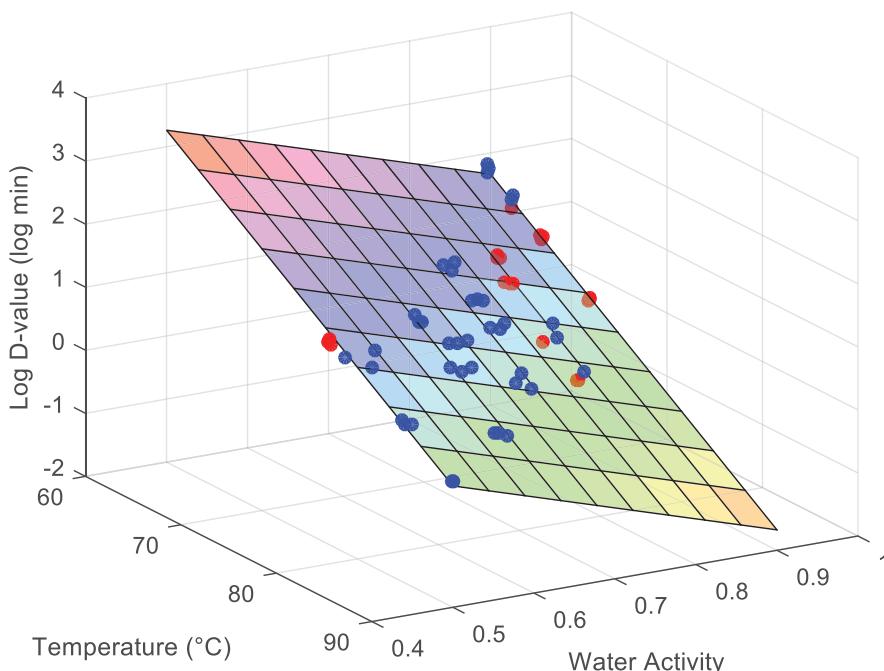
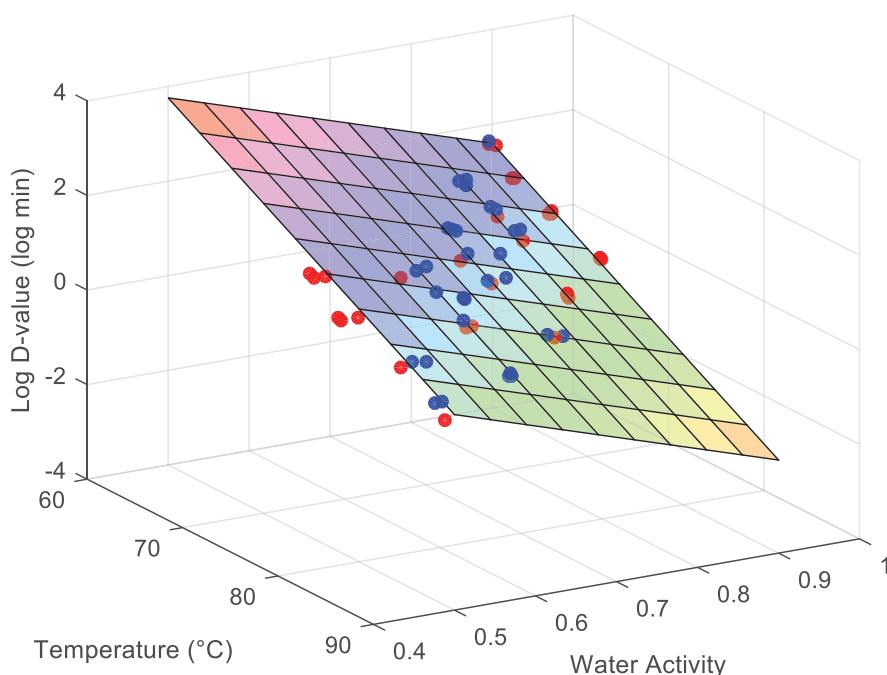


FIGURE 2. Predicted  $D$ -values with respect to temperature and  $a_w$  in high-protein matrices (residuals =  $\log D_{measured} - \log D_{predicted}$ ). Blue circles, positive log  $D$  residuals; red circles, negative log  $D$  residuals.

**FIGURE 3.** Predicted D-values with respect to temperature and  $a_w$  in high-fat matrices (residuals =  $\log D_{measured} - \log D_{predicted}$ ). Blue circles, positive log D residuals; red circles, negative log D residuals.



different food matrices when applying thermal inactivation models.

This research shows that both temperature and  $a_w$  have significant effects on the thermal inactivation rate of *Salmonella*. Determination and control of  $a_w$  in the production of high-protein and high-fat foods is needed to obtain microbiologically safe products. The relative component of a food matrix has a complex role in thermal resistance of *Salmonella* as D-values obtained in the high-fat and high-protein matrices were not universally larger in one product over the other and also depended on temperature and  $a_w$  of the product. To the authors' knowledge, this is the first time such a complex relationship related to food components, temperature and  $a_w$  has been described in the literature. The isothermal models generated in this project may be used to provide the D-value of any combination of  $a_w$  and temperature in the ranges of foods tested. The comprehensive D-values and z-values data may be of use in estimating the thermal inactivation of *Salmonella* in high-protein or high-fat foods during dynamic thermal processes. Further work should be performed to validate the efficacy of the isothermal models on foods processed in pilot- and commercial-scale fields prior to implementation. Further work is needed to develop thermal inactivation models in foods of variable product formulations.

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#### SUPPLEMENTAL MATERIAL

Supplemental material associated with this article can be found online at: <https://doi.org/10.4315/0362-028X.JFP-18-041.s1>.

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