

Tiempo de esterilización (Valor F_0)

Impact of experimental parameter variability on thermal food processing decisions based on microbial safety objectives



Definitions

* Processor must estimate process time (F_T) inactivating bacterial spores of the pathogen of interest

- *Clostridium botulinum*

- Heat-resistant bacterial spore of most severe safety risk

* Prediction Model

$$- F_T = SV D_T$$

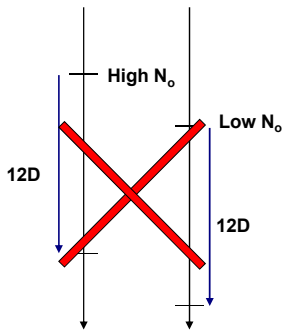
Decimal reduction time

Number of decimal reductions

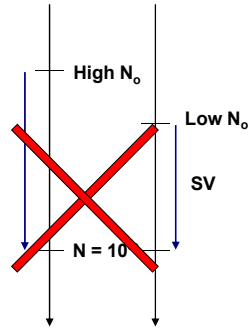


History of Sterilization Objectives

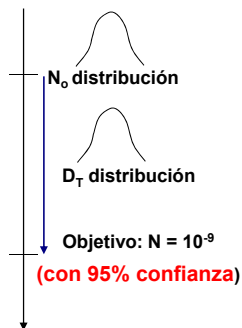
Safety objective: 12D
= Same process



Safety objective: $N = 10^{-9}$
(1 in 1 billion containers)
= Same final load



Tiempo de esterilización (Valor F_o)



$$F_T = SV D_T$$

$$SV = \text{Log} \left(\frac{N_o}{N} \right)$$

Variabilidad

Valor fijo
Objetivo del proceso
 10^{-9} espora/unidad



Tiempo de esterilización (Valor F_0)

Natural prevalence of *C. botulinum* in fresh mushrooms

Sample	Mean N_0 log CFU/g
Fresh mushrooms	-1.36 ± 0.87

Decimal reduction time ($D_{110^\circ\text{C}}$), *C. botulinum* in commercially canned mushrooms

Sample	Temp (°C)	Mean D_T (min)
Commercial canned mushroom	110	0.78 ± 0.17



Data generation:

- * Sample size of each generated dataset must be equal to sample size reported in the literature⁽¹⁾
 - N_0^* refers to N_0 from simulation (12 values)
 - D_T^* refers to D_T from simulation (9 values)

Determination of data acceptability:

- * Metric helps to determine if the distribution of the generated data is approximately equivalent to the distribution of the data reported in the literature

(1) Chu, 2009



Evaluate metric for N_o and D_T values:

$$metric = \left| \frac{(\mu_o - \varepsilon \mu_o)}{\mu_o} \right| f_1 + \left| \frac{(a_o - \varepsilon a_o)}{a_o} \right| f_2 + \left| \frac{(z_o - \varepsilon z_o)}{z_o} \right| f_3$$

- Simulation set accepted if the metric value was less than

$$metric = |1 - \varepsilon| (f_1 + f_2 + f_3) \begin{cases} metric_{N_o} \leq 393 \\ metric_{D_T=110C} \leq 0.45 \end{cases}$$

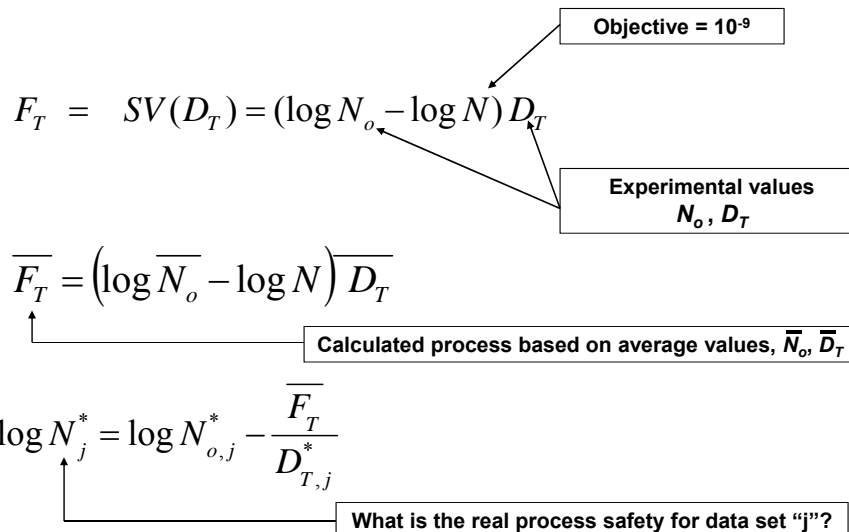
- If rejected, a new set was generated.

This test was a necessary but not a sufficient test as the original experimental dataset was not available to determine its true distribution

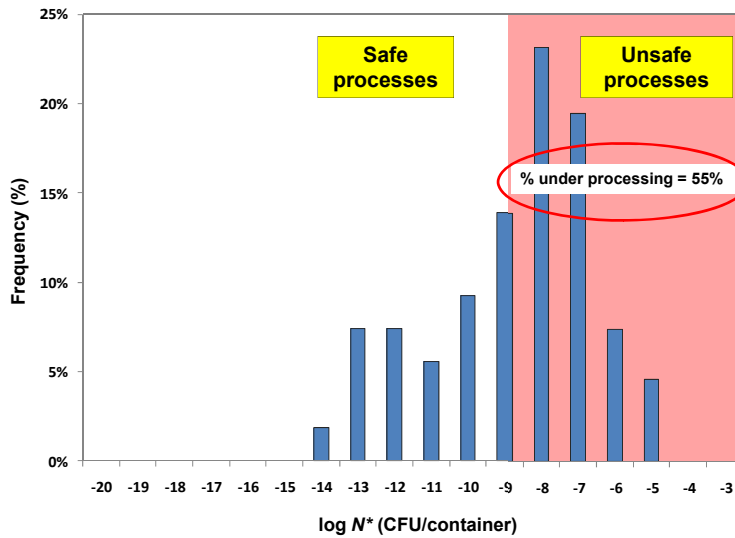
Statistical consulting group. Summer 2008. Dept of Statistics, OSU



Microbial Inactivation



An example of the probability distribution (n = 108) for the estimation *C. botulinum* spore survival ($\log N^*$, CFU/container) based on one accepted N^*_o and $D^*_{110^\circ C}$ dataset



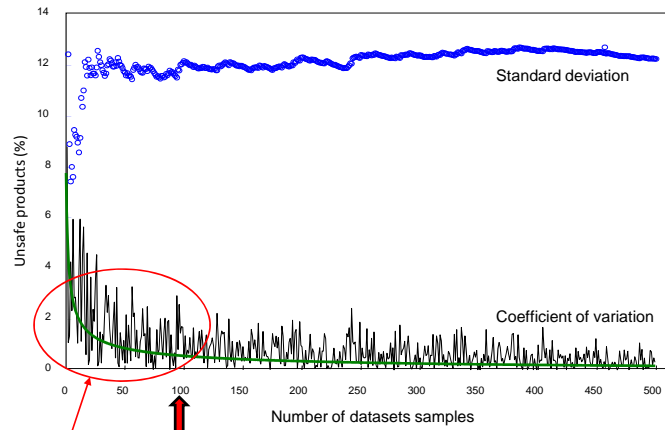
Determination of the recommended number of datasets for the Monte Carlo simulations:

- * Estimate the standard deviation (SD) and coefficient of variation (CV) for the % under processing for 2, 3, 4..., s = 500 datasets
 - The recommended sample size S is the one with an acceptable CV value

Efron & Tibshirani, 1986; Almonacid-Merino & Torres, 2009



Determination of the recommended number of Monte Carlo simulations for the estimation of thermal processing time (F_T)



a sample size of 100 was considered sufficient



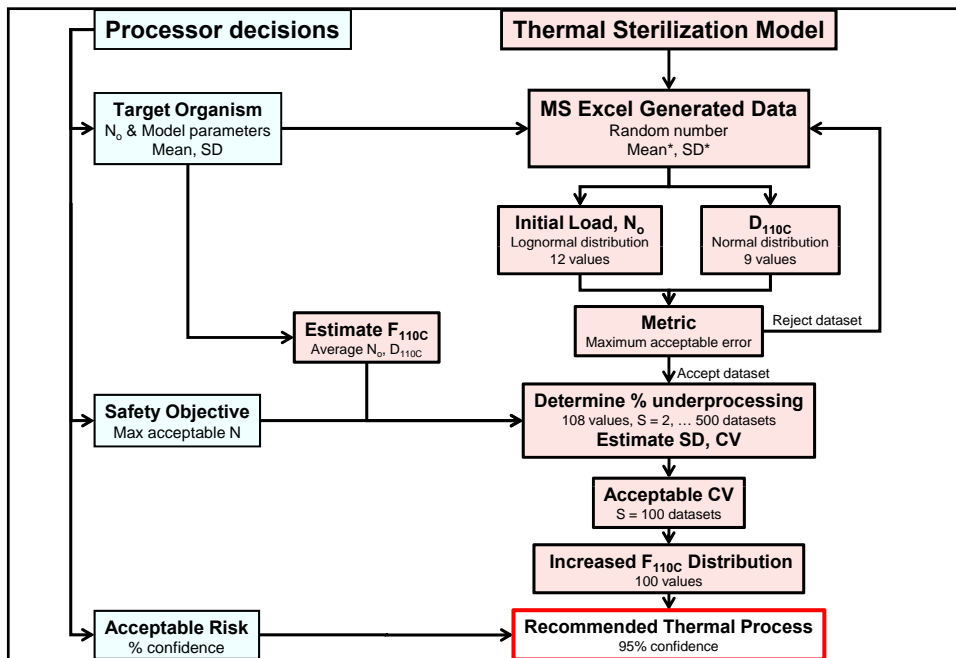
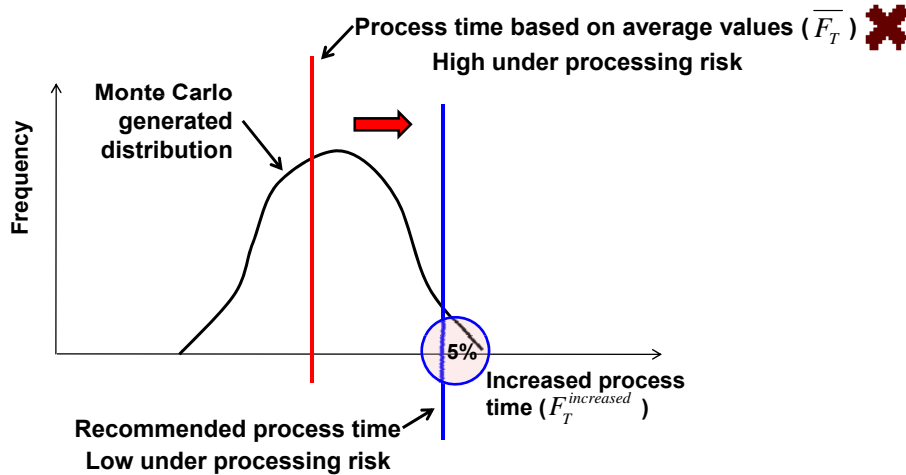
Based on the acceptable 100 datasets:

- * Calculate $\text{Log } N^*$ ($= 12 \times 9 = 108$ values) for each dataset based on the process time F_T calculated using average values (\bar{N}_o & \bar{D}_T)
- * Use the predictive model to estimate an increased processing time $F_T^{increased}$ for each unique N_o^* & D_T^* combination ($n = 100$ sets) to achieve $N = 10^{-9}$ spores/container with 95% confidence
- * Recommend a process time ($F_T^{recommended}$) larger than 95% of the increased processing time values estimated ($F_T^{increased}$, $n = 100$)



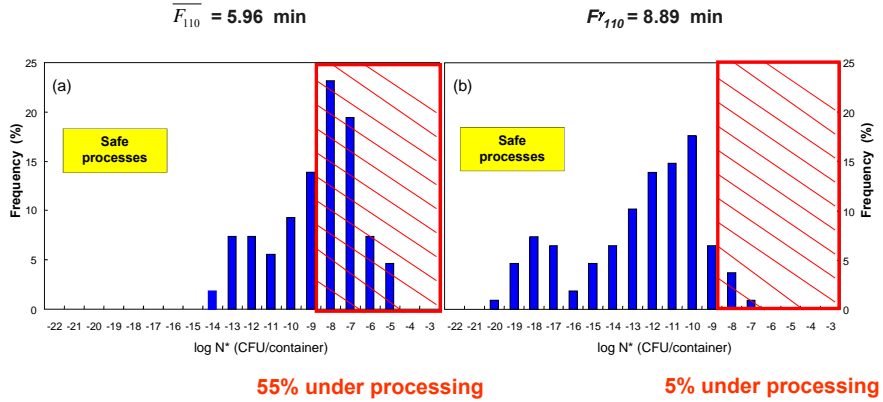
Definition of a recommended process time

Monte Carlo generated distribution ($F_T^{increased}$, $n = 100$ values)



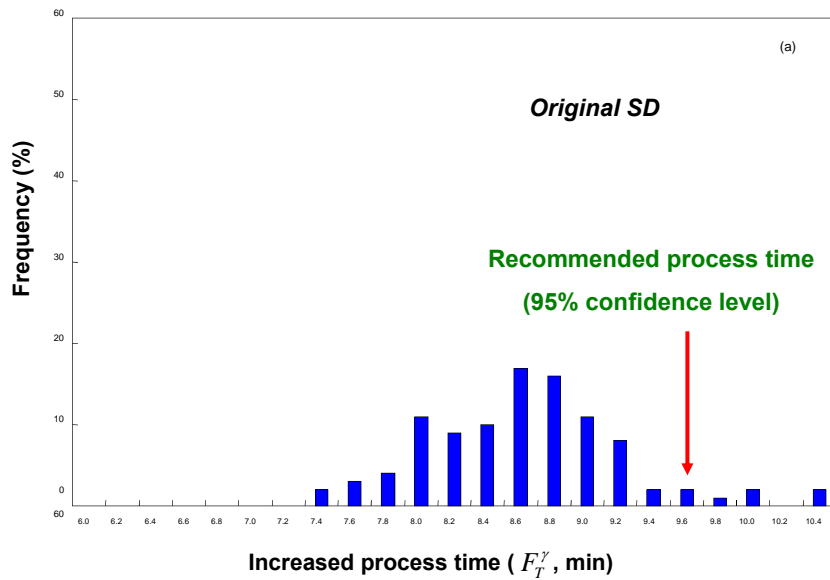
Microbial inactivation

Effect of increasing the processing time to reduce the risk of under processing
(an example)



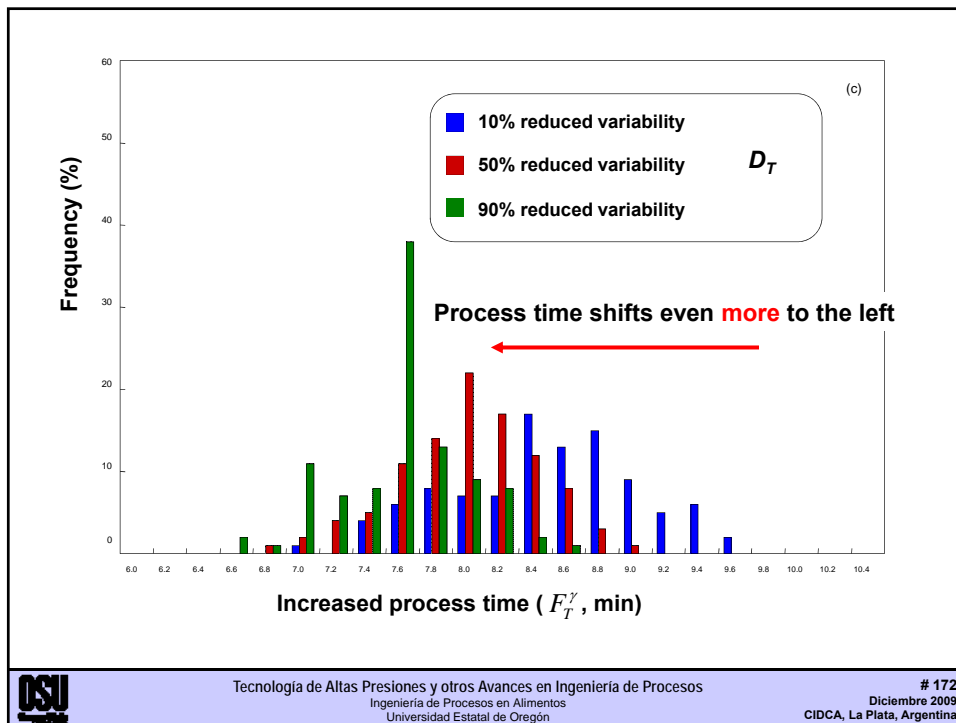
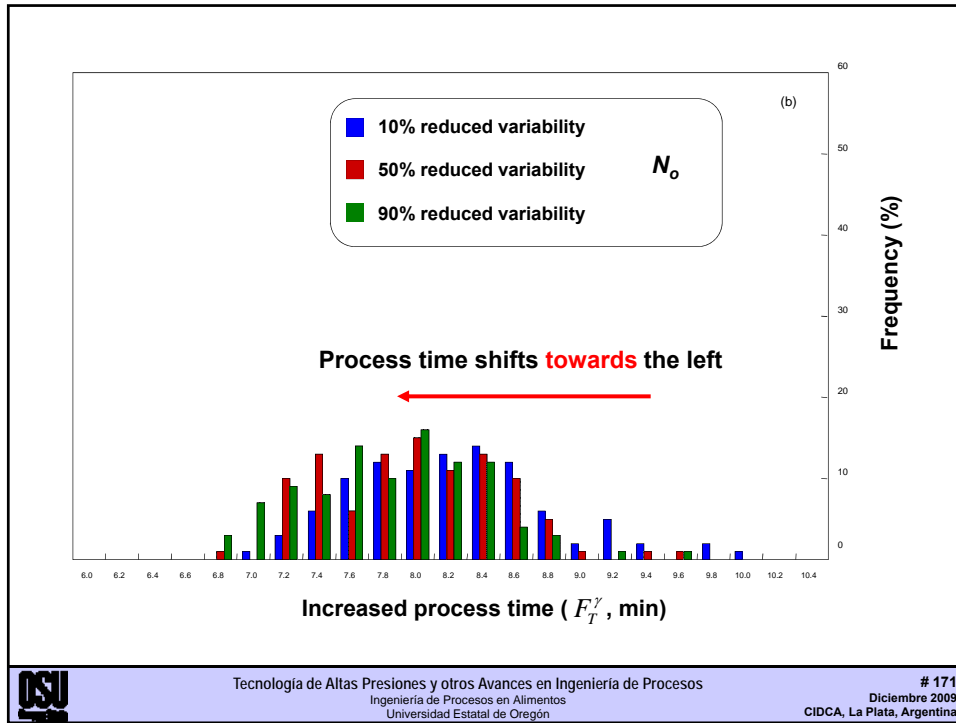
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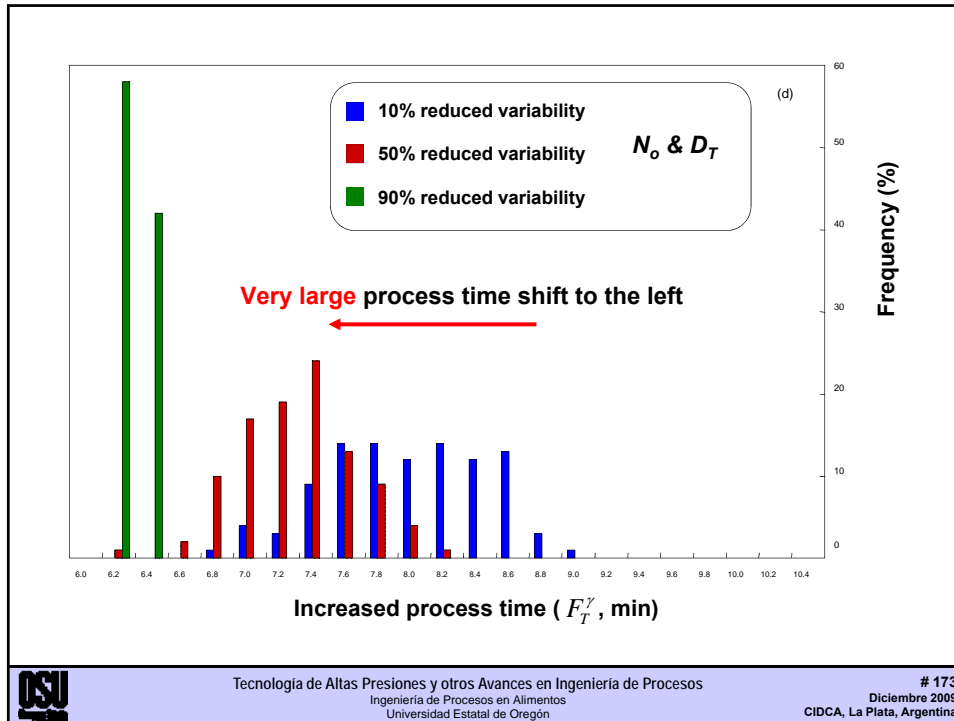
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Results:

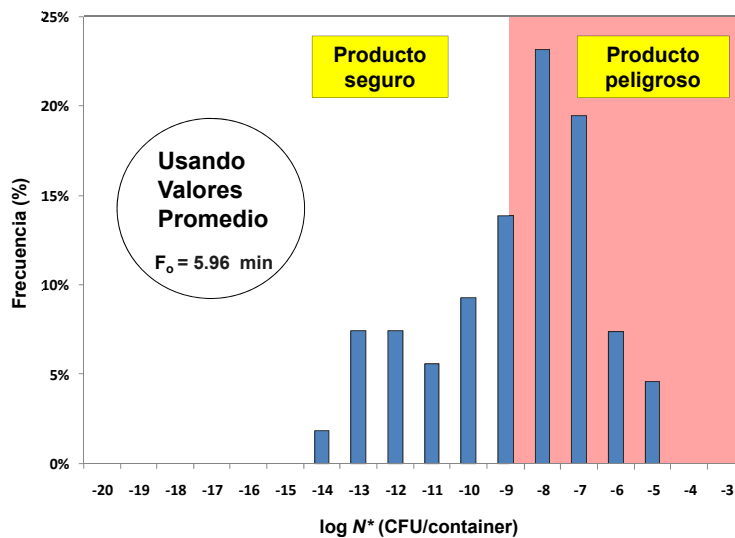
Effect of the reduction in parameter variability			
Average value based estimation (min) = 5.96 (55% under processing)			
Monte Carlo based estimation (min) = 9.60 (5% under processing)			
Recommended process time (min), 95% confidence			
SD reduction	10%	50%	90%
N_o	9.2	8.8	8.6
D_T	9.4	8.6	8.2
N_o & D_T	8.6	7.8	6.4 A 3.2 min reduction!

Conclusions

- * Monte Carlo-type simulations allowed estimation of thermal processing times achieving a desirable surviving spore load probability with a specified confidence level.
- * Monte Carlo-type simulations allowed estimations of the reduction in process time gained by the reduction in the variability of process design parameters.



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Overall conclusions

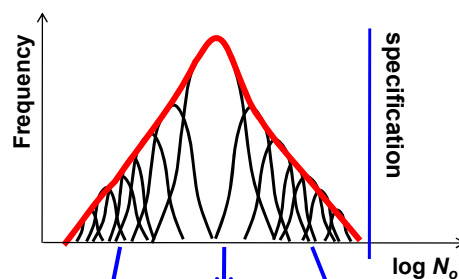
- * All procedures here presented were implemented on MS Excel spreadsheets and used concepts suitable for inclusion in an FST undergraduate program
- * Procedures developed emphasized the importance of minimizing variability, but how can a processor achieve this goal?
- * What are the sources of this variability?
 - If the variability source is the measurement itself, you can reduce it by training, improvement of methods and multiple sampling
 - If the source of the variability is the true heterogeneity of the population (i.e., it is irreducible by measurement improvements), one must find the variability source



Reducing true heterogeneity: an example

Thermal processing, initial microbial load, $\log N_0$

- * Larger processors need a larger number of suppliers and one can expect microbial load differences among them



- * Solution: segregate your suppliers which will reduce variability and reduce processing intensity



¿Preguntas?



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Buen Camino!
Have a good Journey!



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