




Strain Variability of *Listeria monocytogenes* under NaCl Stress Elucidated by a High-Throughput Microbial Growth Data Assembly and Analysis Protocol

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ABSTRACT *Listeria monocytogenes* causes the severe foodborne illness listeriosis and survives in food-associated environments due to its high stress tolerance. A data assembly and analysis protocol for microbial growth experiments was compiled to elucidate the strain variability of *L. monocytogenes* stress tolerance. The protocol includes measurement of growth ability under stress (step 1), selection of a suitable method for growth parameter calculation (step 2), comparison of growth patterns between strains (step 3), and biological interpretation of the discovered differences (step 4). In step 1, *L. monocytogenes* strains ($n = 388$) of various serovars and origins grown on media with 9.0% NaCl were measured using a Bioscreen C microbiology reader. Technical variability of the growth measurements was assessed and eliminated. In step 2, the growth parameters determined by Gompertz, modified-Gompertz, logistic, and Richards models and model-free splines were compared, illustrating differences in the suitability of these methods to describe the experimental data. In step 3, hierarchical clustering was used to describe the NaCl tolerance of *L. monocytogenes* measured by strain-specific variation in growth ability; tolerant strains had higher growth rates and maximum optical densities and shorter lag phases than susceptible strains. The spline parameter area under the curve best classified “poor,” “average,” and “good” growers. In step 4, the tested *L. monocytogenes* lineage I strains (serovars 4b and 1/2b) proved to be significantly more tolerant toward 9.0% NaCl than lineage II strains (serovars 1/2a, 1/2c, and 3a). Our protocol provides systematic tools to gain comparable data for investigating strain-specific variation of bacterial growth under stress.

IMPORTANCE The pathogen *Listeria monocytogenes* causes the foodborne disease listeriosis, which can be fatal in immunocompromised individuals. *L. monocytogenes* tolerates several environmental stressors and can persist in food-processing environments and grow in foodstuffs despite traditional control measures such as high salt content. Nonetheless, *L. monocytogenes* strains differ in their ability to withstand stressors. Elucidating the intraspecies strain variability of *L. monocytogenes* stress tolerance is crucial for the identification of particularly tolerant strains. To enhance reliable identification of variability in bacterial stress tolerance phenotypes, we compiled a large-scale protocol for the entire data assembly and analysis of microbial growth experiments, providing a systematic approach and checklist for experiments on strain-specific growth ability. Our study illustrated the diversity and strain-specific variation of *L. monocytogenes* stress tolerance with an unprecedented scope and dis-

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covered biologically relevant serovar- and lineage-dependent phenotypes of NaCl tolerance.

KEYWORDS *Listeria*, Bioscreen, osmotic, salt, stress tolerance, growth model, model-free spline, Gompertz, modified-Gompertz, logistic, Richards, kinetic parameter, osmotic stress, stress response

The foodborne pathogen *Listeria monocytogenes* causes listeriosis, a severe illness to which the elderly, infants, and immunocompromised or pregnant individuals are particularly susceptible (1, 2). Preventing the occurrence of *L. monocytogenes* is difficult due to its flexible lifestyle, which is supported by its ability to survive and adapt to stressors (3–5) in various environments, including foods, soil, water, sewage, and mammalian hosts (6, 7). *L. monocytogenes* is a problematic contaminant of the food chain, as it has the ability to persist in processing facilities, contaminate foodstuffs, and cause infections predominantly via ready-to-eat foods (7–13). Stress tolerance contributes to the survival of *L. monocytogenes* within the food chain (14, 15) and facilitates its transition from a saprophyte to a pathogen (16–21). The genetic lineages and sublineages of *L. monocytogenes* have been described to be unevenly represented among isolates from different environments and hosts (22–26) and to contain differing stress resistance genes (27). *L. monocytogenes* serovars 1/2a, 1/2c, and 3a belong to the traditionally food-associated lineage II, whereas serovars 1/2b and 4b are assigned to lineage I (28). *L. monocytogenes* strains possess variable stress tolerance qualities (29, 30), and the overall intraspecies diversity of *L. monocytogenes* stress tolerance, including lineage-associated traits, remains to be elucidated.

Phenotypic strain variability, if not considered, may cause bias in microbiological investigations (31). To capture overall intraspecies diversity, the investigation of the variability of *L. monocytogenes* stress tolerance should employ large strain collections of different origins and genotypes. However, large experiments are labor-intensive, necessitating methodologies to perform numerous experiments in bulk. Turbidity via microplate absorbance (optical density [OD]) measurement technologies has been widely used for decades in the study of microbial growth patterns and stress tolerance (32–37). Researchers have considered quantitative approaches, such as comparison with traditional viable counts and calibration of OD measurements to estimate cell counts and kinetic parameters (36–42). Kinetic parameters calculated from OD measurements systematically deviate from parameters obtained using viable counts (36, 37, 39, 40), and since the deviation is systematic, OD is deemed suitable for the relative comparison of growth patterns between strains (41). However, execution of the entire data collection and analysis protocol in a reliable manner, including a stepwise compilation of considerations and tools that yield comparable, qualitative high-throughput data on bacterial stress tolerance phenotypes, has received less attention.

Microbial susceptibility and tolerance toward stressors can be quantified by growth parameters derived from mathematical models (43–47). Commonly used kinetic parameters include lag time (lag phase), maximum specific growth rate, asymptotic (maximum) growth level (44), and area under the curve (48, 49). The importance of selecting a suitable growth model for parameter estimation has been emphasized, as some models assume relationships between parameters or are unsuited for certain types of data (41, 46, 47, 50, 51). Although numerous models with differing underlying principles have been developed and used, one generally has no inherent superiority over another (44). Therefore, determining a parameter estimation approach that suits the purposes of the experiment is relevant for each study design.

The aims of our study were to establish a high-throughput data assembly and analysis protocol for OD measurements of microbial growth and to explore the reliability of the protocol for investigating the intraspecies diversity of *L. monocytogenes*

TABLE 1 Summary of the data assembly and analysis protocol for strain variability of growth ability under stress conditions

Protocol step	Specific consideration(s)	Method(s) or example(s) ^a
1. Measurement of growth ability under stress	Optimize stress condition by piloting strain variability. Randomize pilot strains and the testing order of strains.	Fig. S1 Functions to generate random numbers without repeats, e.g., in R or Excel
	Minimize potential technical variation and batch effects: Use the same growth medium throughout the entire experiment. Have one individual perform all experiments, if possible.	Prepare the required amount all at once Fig. S4, S5
	Utilize within-experiment technical and biological replicates.	Fig. S11
	Utilize between-experiment control strains.	Fig. S11
	When using Bioscreen, leave the outermost rim of the honeycomb plate blank to avoid test broth evaporation.	Fig. S3, S11
	Detect potential contamination and empty wells; retest, if necessary.	Visualize growth curves during or after experiment; culture honeycomb plate wells with unusual growth and no growth Fig. S2, Text S10: volume 1 Fig. S6–9, Text S10: volumes 2 and 3
2. Selection of a suitable method for growth parameter calculation	Inspect and deal with outliers. Inspect and normalize potential technical variation and batch effects.	Text S10: volume 4 Text S10: volume 5
	Calculate growth parameters using several methods. Compare the fit and values of the parameter calculation methods.	Text S10: volume 6
3. Comparison of growth patterns between strains	Visualize growth parameters to see overall parameter variation.	Text S10: volume 6
	Visualize growth curves to see entire growth patterns. Combine statistical methods and intuitive reasoning to determine a suitable way to quantify strain variability.	Text S10: volumes 3 and 7 Fig. S12–15
4. Biological interpretation of the discovered differences	Classify strain variability and interpret it via growth parameters.	Text S10: volumes 6 and 7
	Investigate strain variability with biological background variables and draw conclusions.	Data exploration and statistical tests (see Materials and Methods for examples)

^aMethods or examples described in the supplemental material published with this article are indicated by their number. Text S10 includes R codes for the data analyses and is divided into volumes 1 to 7 according to their content.

stress tolerance. Osmotic (salt) stress was chosen as an example due to its relevance at several phases of the bacterial ecology (52–58).

RESULTS

Our data assembly and analysis protocol incorporates the following steps that are essential for the identification of variability in growth ability, i.e., stress tolerance, among bacterial strains: measurement of growth ability under stress (step 1), selection of a suitable method for growth parameter calculation (step 2), comparison of the growth patterns between strains (step 3), and biological interpretation of the discovered differences (step 4) (Table 1).

Measurement of growth ability under stress (step 1). (i) 9.0% NaCl distinguished strain variability of *L. monocytogenes* growth ability under salt stress. A pilot study using a Bioscreen C microbiology reader determined the sufficient NaCl concentration for identifying differences of stress tolerance between *L. monocytogenes* strains. At 6.5% NaCl, most pilot strains (Table 2) reached the stationary phase within 7 h and grew to relatively high ODs, displaying little variation in growth patterns between strains (see Fig. S1 in the supplemental material). At 7.5% NaCl, the pilot strains reached the stationary phase within approximately 10 h, and a slight decrease of maximum OD and an increased variability of growth patterns appeared (Fig. S1). Variation between the growth patterns of the pilot strains was apparent at 8.5% NaCl, but the overall decrease in their maximum optical density remained moderate (Fig. S1). To ensure that the selected NaCl concentration would distinguish strain variability, the final test concentration of brain-heart infusion (BHI) broth was set at 9.0% NaCl.

TABLE 2 *Listeria monocytogenes* strains used in this study

Strain	Immediate source ^a	Isolation source	Isolation yr ^b	Serovar	Pilot (P) or control (C) ^c
LE57E	Finland	Processing environment	2000	1/2a	–
LU12/2	Finland	Pig, feces	2003	1/2a	–
AE30E	Finland	Processing environment	1999	1/2a	–
LK126	Finland	Fish	2000	1/2a	–
LM86	Finland	Fish	NA	1/2a	–
MJL41	Finland	Meat	2009	1/2a	–
MJL47	Finland	Meat	2009	1/2a	–
TT82E	Finland	Fish	1997	1/2a	–
PE4E/1	Finland	Processing environment	NA	1/2a	–
HT45E	Finland	Poultry	1998	1/2a	–
LK132	Finland	Fish	2000	1/2a	–
TT107E	Finland	Fish	1998	1/2a	–
LM210	Finland	Fish	NA	1/2a	–
MJL43	Finland	Meat	2009	1/2a	–
LMML90	Finland	Meat	2004	1/2a	–
LMML100	Finland	Meat	2004	1/2a	–
L34-s	Finland	Meat	1999–2001	1/2a	–
LU26/1	Finland	Pig, feces	2003	1/2a	–
LU103/1	Finland	Meat	2003	1/2a	–
MJL14	Finland	Processing environment	2009	1/2a	–
HL34E/1	Finland	Processing environment	1999	1/2a	–
HT65E/1	Finland	Meat	1998	1/2a	–
LK133	Finland	Fish	2000	1/2a	–
RL22E/1	Finland	Processing environment	2000	1/2a	–
LL16/2	Finland	Wild bird, feces	1998	1/2a	–
L12s	Finland	Poultry	1999–2001	1/2a	–
LK36	Finland	Fish	1993	1/2a	–
LK129	Finland	Fish	2000	1/2a	–
TR47E	Finland	Fish	NA	1/2a	–
L47s	Finland	Poultry	1999–2001	1/2a	–
LM70	Finland	Dairy product	1988	1/2a	–
LL82/1	Finland	Wild bird, feces	2001	1/2a	–
LT3/1	Finland	Meat	1999	1/2a	–
IR17V	Finland	Milk	1998	1/2a	–
LT26E	Finland	Vegetable	2000	1/2a	–
LM69	Finland	Dairy product	1988	1/2a	–
LT14/1	Finland	Meat	1999	1/2a	–
LT4/1	Finland	Meat	1999	1/2a	–
LA40	Finland	Bird	1989	1/2a	–
LT16/1	Finland	Meat	1999	1/2a	–
LK134	Finland	Fish	2000	1/2a	–
LE27E	Finland	Processing environment	1999	1/2a	–
JHM70	Finland	Milk	1987	1/2a	–
HE1E	Finland	Processing environment	1997	1/2a	–
LK60/1	Finland	Fish	1999	1/2a	–
LE55E	Finland	Processing environment	2000	1/2a	P
LU74/1	Finland	Meat	2003	1/2a	–
RE46E	Finland	Processing environment	1997	1/2a	–
JHM339	Finland	Silage	1988	1/2a	–
JHM71	Finland	Cow, feces	1987	1/2a	–
LU49/1	Finland	Meat	2003	1/2a	–
LMML10	Finland	Meat	2003	1/2a	–
LA35	Finland	Sheep	1998	1/2a	–
MJL34	Finland	Meat	2009	1/2a	–
LL51/2	Finland	Wild bird, feces	2001	1/2a	–
L10s	Finland	Meat	2000	1/2a	–
LL18/3	Finland	Wild bird, feces	1998	1/2a	–
LU102/1	Finland	Meat	2003	1/2a	–
LT9/1	Finland	Meat	1999	1/2a	–
LU118/1	Finland	Meat	2003	1/2a	–
LT13/1	Finland	Meat	1999	1/2a	–
JHM3	Finland	Cow, feces	1987	1/2a	–
JHM110	Finland	Cow, feces	1987	1/2a	–
LM104/1	Finland	Vegetable	1999	1/2a	–
JHM35	Finland	Cow, feces	1987	1/2a	–

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TABLE 2 (Continued)

Strain	Immediate source ^a	Isolation source	Isolation yr ^b	Serovar	Pilot (P) or control (C) ^c
LT33/1	Finland	Meat	2000	1/2a	–
LML46	Finland	NA	NA	1/2a	–
JHM349	Finland	Silage	1988	1/2a	–
10B	Finland	Fish	1999–2001	1/2a	–
JHM387	Finland	Silage	1988	1/2a	–
LL24/2	Finland	Wild bird, feces	1998	1/2a	–
IR18V	Finland	Milk	NA	1/2a	–
LM110	Finland	Vegetable	1999	1/2a	–
LA68	Finland	Animal	1999	1/2a	–
TE9/1	Finland	Meat	NA	1/2a	–
MJL2	Finland	Meat	2009	1/2a	–
LT12/1	Finland	Meat	1999	1/2a	–
LK130	Finland	Fish	2000	1/2a	–
LL65/1	Finland	Wild bird, feces	2001	1/2a	–
LA61	Finland	Meat	1993	1/2a	–
LU123/1	Finland	Meat	2003	1/2a	–
LK121/1	Finland	Fish	2001	1/2a	–
LK51	Finland	Fish	1999	1/2a	–
LMK1	Finland	Fish	1996	1/2a	–
LU56/3	Finland	Meat	2003	1/2a	–
JHM229	Finland	Cow	1975	1/2a	P
TE27/1	Finland	Meat	NA	1/2a	–
LA42	Finland	Cow	1989	1/2a	–
MJL1	Finland	Meat	2009	1/2a	–
IR16V	Finland	Milk	1998	1/2a	–
LA48	Finland	Sheep	1990	1/2a	–
HR5E	Finland	Meat	1998	1/2a	–
LK54/1	Finland	Fish	1999	1/2a	–
LU120/1	Finland	Meat	2003	1/2a	–
LT5/1	Finland	Meat	1999	1/2a	–
LL85/1	Finland	Wild bird, feces	2001	1/2a	–
LM298	Finland	Salad	2002	1/2a	–
TL1E	Finland	Processing environment	NA	1/2a	–
LT32/1	Finland	Meat	2000	1/2a	–
TE12/1	Finland	Animal	NA	1/2a	–
LM111	Finland	Processing environment	NA	1/2a	–
LL22/8	Finland	Wild bird, feces	1998	1/2a	–
LL52/2	Finland	Wild bird, feces	2001	1/2a	–
LMML101	Finland	Meat	2004	1/2a	–
13M	Finland	Poultry	2000	1/2a	–
MJL21	Finland	Meat	2009	1/2a	–
TE4/1	Finland	Meat	NA	1/2a	–
RE1E	Finland	Processing environment	1997	1/2a	–
HR19E/1	Finland	Meat	2000	1/2a	–
18B	Finland	Fish	1999–2001	1/2a	–
L96s	Finland	Poultry	1999–2001	1/2a	–
LK89/1	Finland	Fish	2001	1/2a	–
LM136	Finland	Fish	NA	1/2a	–
LA64	Finland	Cow	1994	1/2a	–
LMML117	Finland	Meat	2004	1/2a	–
LT2/1	Finland	Meat	1999	1/2a	–
LK131	Finland	Fish	2000	1/2a	–
LL83/1	Finland	Wild bird, feces	2001	1/2a	–
LMK28	Finland	Fish	1996	1/2a	–
LM84	Finland	Ready-to-eat food	1999	1/2a	–
MJL45	Finland	Meat	2009	1/2a	–
L83s	Finland	Meat	2001	1/2a	–
LL66/3	Finland	Wild bird, feces	2001	1/2a	–
LU44/1	Finland	Meat	2003	1/2a	–
LML48	Finland	Meat	1996	1/2a	–
LT22/1	Finland	Meat	1999	1/2a	–
LT23/1	Finland	Meat	2000	1/2a	–
LE52E	Finland	Processing environment	2000	1/2a	–
JHM15	Finland	Milk	1987	1/2a	–
LA33	Finland	Cow	1998	1/2a	–

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TABLE 2 (Continued)

Strain	Immediate source ^a	Isolation source	Isolation yr ^b	Serovar	Pilot (P) or control (C) ^c
HT101E/1	Finland	Vegetable	2000	1/2a	–
HL90E/1	Finland	Processing environment	2000	1/2b	–
RE74E	Finland	Processing environment	1997	1/2b	–
L94s	Finland	Poultry	1999–2001	1/2b	–
LML44	Finland	Meat	1996	1/2b	P
LM115	Finland	Vegetable	1995	1/2b	–
LSO885/1	Finland	Processing environment	1996	1/2b	–
LML36	Finland	Processing environment	NA	1/2b	–
LL31/1	Finland	Wild bird, feces	1998	1/2b	C, P
LM103/1	Finland	Vegetable	NA	1/2b	–
LA46	Finland	Sheep	1990	1/2b	–
LM116	Finland	Vegetable	1999	1/2b	–
L25s	Finland	Meat	1999–2001	1/2b	–
2904	Finland	Processing environment	1990	1/2b	–
2919	Finland	Processing environment	1996	1/2b	–
3129	Finland	Processing environment	1994	1/2b	P
2920	Finland	Processing environment	1996	1/2b	–
LM105	Finland	Vegetable	NA	1/2b	–
HT69E	Finland	Meat	1999	1/2c	P
L125s	Finland	Meat	NA	1/2a	–
MJL40	Finland	Meat	2009	1/2c	P
HE161E/1	Finland	Processing environment	1999	1/2c	–
RE70E	Finland	Processing environment	1997	1/2c	–
AR5E	Finland	Poultry	1999	1/2c	–
LM89	Finland	Processing environment	NA	1/2c	–
HL6E	Finland	Processing environment	1998	1/2c	–
HE152E	Finland	Processing environment	1999	1/2c	–
AT3E	Finland	Meat	1995	1/2c	–
HT93E/1	Finland	Ready-to-eat food	2000	1/2c	–
HT100E/1	Finland	Poultry	2000	1/2c	–
HE28E	Finland	Processing environment	1997	1/2c	–
L51s	Finland	Meat	1999–2001	1/2c	–
LL40/2	Finland	Wild bird, feces	1998	1/2c	–
LMMML65	Finland	Meat	2003	1/2a	–
LMMML67	Finland	Meat	2003	1/2c	–
KE1E	Finland	Processing environment	1998	1/2c	–
E7	Finland	Dairy product	1999	3a	–
LK42/1	Finland	Fish	1998	3a	P
LK55/1	Finland	Fish	1999	3a	–
LM128	Finland	Vegetable	1999	3a	–
LK127	Finland	Fish	2000	4b	–
JHM230	Finland	Cow	1984	4b	–
LL17/3	Finland	Wild bird, feces	1998	4b	–
JHM331	Finland	Silage	1988	4b	–
LT25E	Finland	Vegetable	2000	4b	–
LL91/1	Finland	Wild bird, feces	2001	4b	–
LL78/1	Finland	Wild bird, feces	2001	4b	–
LK43/3	Finland	Fish	1998	4b	–
LL67/1	Finland	Wild bird, feces	2001	4b	–
LA22	Finland	Sheep	1986	4b	–
LA30	Finland	Sheep	1987	4b	–
44M	Finland	Meat	1999–2001	4b	–
LL72/1	Finland	Wild bird, feces	2001	4b	–
LE21E	Finland	Processing environment	1999	4b	P
LA56	Finland	Poultry	1992	4b	–
LL49/2	Finland	Wild bird, feces	2001	4b	–
AE18E	Finland	Processing environment	1998	4b	–
JHM270	Finland	Cow, feces	1987	4b	–
LL87/3	Finland	Wild bird, feces	2001	4b	–
LL4/2	Finland	Wild bird, feces	1998	4b	P
LL1/3	Finland	Wild bird, feces	1998	4b	–
LU97/4	Finland	Meat	2003	4b	–
LL71/1	Finland	Wild bird, feces	2001	4b	–
Lm 217	Switzerland	Meat	2000	1/2a	P
Lm 57	Switzerland	Meat	1999	1/2b	–

(Continued on next page)

TABLE 2 (Continued)

Strain	Immediate source ^a	Isolation source	Isolation yr ^b	Serovar	Pilot (P) or control (C) ^c
Lm 60	Switzerland	Human	2006	1/2a	–
Lm 69	Switzerland	Human	2006	1/2a	–
LmE240	Switzerland	Carcass	2011	1/2a	P
LmE61	Switzerland	Carcass	2011	1/2a	–
LmS1	Switzerland	Environment	2011	1/2a	–
LmS9	Switzerland	Environment	2011	1/2a	–
N11-1218	Switzerland	Poultry	2011	1/2a	–
N11-1255	Switzerland	Human	2011	1/2a	–
N11-1285	Switzerland	Human	2011	1/2a	–
N11-1346	Switzerland	Human	2011	1/2a	–
N11-1415	Switzerland	Human	2011	1/2a	–
N11-1515	Switzerland	Milk	2011	1/2a	–
N11-1546	Switzerland	Human	2011	1/2a	–
N11-1617	Switzerland	Meat	2011	1/2a	–
N11-1649	Switzerland	Meat	2011	1/2a	–
N11-1653	Switzerland	Meat	2011	1/2a	–
N11-1696	Switzerland	Meat	2011	1/2a	–
N11-1734	Switzerland	Ham	2011	1/2a	–
N11-1845	Switzerland	Dairy	2011	1/2a	–
N11-1905	Switzerland	Poultry	2011	1/2a	–
N11-2183	Switzerland	Salad	2011	1/2a	–
N11-2215	Switzerland	Dairy	2011	1/2a	–
N11-2272	Switzerland	Maize	2011	1/2a	–
N11-2345	Switzerland	Meat	2011	1/2a	–
N11-2509	Switzerland	Meat	2011	1/2a	–
N11-2538	Switzerland	Milk	2011	1/2a	–
N11-2542	Switzerland	Cheese	2011	1/2a	–
N11-2543	Switzerland	Environment	2011	1/2a	–
N11-2554	Switzerland	Meat	2011	1/2a	–
N11-2662	Switzerland	Meat	2011	1/2a	–
N12-0088	Switzerland	Fish	2012	1/2a	–
N12-0275	Switzerland	Milk	2012	1/2a	–
N12-0299	Switzerland	Quorn	2012	1/2a	–
N12-0303	Switzerland	Quorn	2012	1/2a	–
N12-0373	Switzerland	Seafood	2012	1/2a	–
N12-0402	Switzerland	Seafood	2012	1/2a	–
N12-0435	Switzerland	Seafood	2012	1/2a	–
N12-0459	Switzerland	Seafood	2012	1/2a	–
N12-0494	Switzerland	Meat	2012	1/2a	–
N12-0561	Switzerland	Environment	2012	1/2a	–
N12-0571	Switzerland	Meat	2012	1/2a	–
N12-0677	Switzerland	Sausage	2012	1/2a	–
N12-0762	Switzerland	Meat	2012	1/2a	–
N12-0796	Switzerland	Meat	2012	1/2a	–
N12-0823	Switzerland	Vegetable	2012	1/2a	–
N12-0906	Switzerland	Milk	2012	1/2a	–
N12-0999	Switzerland	Environment	2012	1/2a	–
N12-1002	Switzerland	Milk	2012	1/2a	–
N12-1024	Switzerland	Meat	2012	1/2a	–
N13-0796	Switzerland	Human	2013	1/2a	–
N12-1436	Switzerland	Meat	2012	1/2a	–
N12-2549	Switzerland	Human	2012	1/2a	–
N12-1273	Switzerland	Human	2012	1/2a	–
N12-1641	Switzerland	Environment	2012	1/2a	–
N12-1667	Switzerland	Environment	2012	1/2a	–
N12-1731	Switzerland	Environment	2012	1/2a	–
N12-2118	Switzerland	Meat	2012	1/2a	–
N12-2188	Switzerland	Meat	2012	1/2a	–
N12-2229	Switzerland	Food	2012	1/2a	–
N12-2236	Switzerland	Meat	2012	1/2a	–
N12-2313	Switzerland	Environment	2012	1/2a	–
N12-2320	Switzerland	Food	2012	1/2a	–
N12-2329	Switzerland	Food	2012	1/2a	–
N12-2389	Switzerland	Milk	2012	1/2a	–
N13-0225	Switzerland	Meat	2013	1/2a	–

(Continued on next page)

TABLE 2 (Continued)

Strain	Immediate source ^a	Isolation source	Isolation yr ^b	Serovar	Pilot (P) or control (C) ^c
N13-0228	Switzerland	Milk	2013	4b	P
N13-0245	Switzerland	Meat	2013	1/2a	–
N13-0254	Switzerland	Meat	2013	1/2a	–
N13-0288	Switzerland	Meat	2013	1/2a	–
N13-0369	Switzerland	Meat	2013	1/2a	–
N13-0406	Switzerland	Goat	2013	1/2a	–
N11-2039	Switzerland	Human	2011	1/2a	–
N13-0474	Switzerland	Meat	2013	1/2a	–
N13-0714	Switzerland	Human	2013	1/2a	–
N11-1547	Switzerland	Human	2011	1/2a	–
N11-1837	Switzerland	Human	2011	1/2c	–
N11-2036	Switzerland	Human	2011	1/2a	–
N11-2134	Switzerland	Human	2011	1/2a	–
N11-2474	Switzerland	Human	2011	1/2a	–
N11-2553	Switzerland	Human	2011	1/2a	–
N12-0123	Switzerland	Human	2012	1/2a	–
N12-0258	Switzerland	Human	2012	1/2a	–
N12-0560	Switzerland	Human	2012	1/2a	–
N12-0588	Switzerland	Human	2012	1/2a	–
N12-0922	Switzerland	Human	2012	1/2a	–
N12-0935	Switzerland	Human	2012	1/2a	–
N12-1107	Switzerland	Human	2012	1/2a	–
N12-1872	Switzerland	Human	2012	1/2a	–
N12-1873	Switzerland	Human	2012	1/2a	–
N12-1917	Switzerland	Human	2012	1/2a	–
N12-2031	Switzerland	Human	2012	1/2a	–
N12-2082	Switzerland	Human	2012	1/2a	–
N12-2169	Switzerland	Human	2012	1/2a	–
N13-0048	Switzerland	Human	2013	1/2a	–
N13-0094	Switzerland	Human	2013	1/2a	–
N13-0119	Switzerland	Human	2013	1/2a	–
N13-0281	Switzerland	Human	2013	1/2a	–
N11-1584	Switzerland	Human	2013	1/2a	–
N13-0698	Switzerland	Human	2013	1/2a	–
N13-0287	Switzerland	Human	2013	1/2a	–
N13-0733	Switzerland	Human	2013	1/2a	–
N13-0739	Switzerland	Human	2013	1/2a	–
N12-0367	Switzerland	Human	2012	1/2a	–
Lm 19	Switzerland	Human	2005	1/2b	–
Lm 45	Switzerland	Human	2005	1/2b	–
Lm 5	Switzerland	Meat	1999	1/2b	–
Lm 9	Switzerland	Meat	1999	1/2b	–
LmE188	Switzerland	Carcass	2011	1/2b	–
LmE212	Switzerland	Carcass	2011	1/2b	–
LmS2	Switzerland	Environment	2011	1/2b	–
LmSB1	Switzerland	Seafood	2011	1/2b	–
N11-1251	Switzerland	Sausage	2011	1/2b	–
N11-1252	Switzerland	Salad	2011	1/2b	–
N11-2675	Switzerland	Human	2011	1/2a	–
N12-1307	Switzerland	Meat	2012	1/2b	P
N12-1608	Switzerland	Human	2012	1/2b	–
N12-1609	Switzerland	Human	2012	1/2b	–
N12-1914	Switzerland	Milk	2012	1/2b	–
N12-2025	Switzerland	Food	2012	1/2b	–
N12-2387	Switzerland	Meat	2012	1/2b	–
N12-2441	Switzerland	Milk	2012	1/2a	–
N12-2449	Switzerland	Milk	2012	1/2b	–
N12-2532	Switzerland	Human	2012	1/2a	–
N12-2533	Switzerland	Human	2012	1/2a	–
N13-0402	Switzerland	Fish	2013	1/2b	–
N13-0581	Switzerland	Human	2013	1/2b	–
N13-0762	Switzerland	Human	2013	1/2a	–
Lm 22/3A	Switzerland	Human	2004	1/2c	–
Lm 25/9	Switzerland	Meat	2002	1/2c	–
Lm 28	Switzerland	Human	2005	1/2c	–

(Continued on next page)

TABLE 2 (Continued)

Strain	Immediate source ^a	Isolation source	Isolation yr ^b	Serovar	Pilot (P) or control (C) ^c
Lm 760	Switzerland	Meat	2001	1/2c	–
N11-1514	Switzerland	Meat	2011	1/2c	–
N12-0486	Switzerland	Human	2012	1/2c	–
N12-0563	Switzerland	Meat	2012	1/2c	P
N12-0644	Switzerland	Meat	2012	1/2c	–
N12-0710	Switzerland	Meat	2012	1/2c	–
N12-0822	Switzerland	Meat	2012	1/2c	–
N12-1773	Switzerland	Meat	2012	1/2c	–
N12-1921	Switzerland	Rice	2012	1/2c	P
N12-2151	Switzerland	Poultry	2012	1/2c	–
N12-2271	Switzerland	Meat	2012	1/2c	–
N12-2386	Switzerland	Environment	2012	1/2c	–
N13-0001	Switzerland	Human	2013	1/2c	–
N12-0318	Switzerland	Quorn	2012	3a	P
Lm 49	Switzerland	Meat	1999	4b	–
Lm 58	Switzerland	Human	2006	4b	–
Lm 72	Switzerland	Human	2006	4b	–
Lm 8	Switzerland	Meat	1999	4b	–
Lm LL195	Switzerland	Human	1983–87	4b	C, P
Lm LL201	Switzerland	Human	1983–87	4b	–
LmE131	Switzerland	Carcass	2011	4b	–
LmE153	Switzerland	Carcass	2011	4b	–
LmE154	Switzerland	Carcass	2011	4b	–
LmE162	Switzerland	Carcass	2011	4b	–
N11-1846	Switzerland	Meat	2011	4b	–
N11-1850	Switzerland	Dairy product	2011	4b	–
N11-2292	Switzerland	Human	2011	4b	–
N11-2618	Switzerland	Human	2011	4b	–
N11-2747	Switzerland	Human	2011	4b	–
N11-2801	Switzerland	Human	2011	4b	–
N12-0160	Switzerland	Meat	2012	4b	–
N12-0320	Switzerland	Human	2012	4b	–
N12-0341	Switzerland	Human	2012	4b	–
N12-0432	Switzerland	Meat	2012	4b	–
N12-0466	Switzerland	Meat	2012	4b	–
N12-0529	Switzerland	Meat	2012	4b	P
N12-0551	Switzerland	Human	2012	4b	–
N12-0570	Switzerland	Human	2012	4b	–
N12-0575	Switzerland	Human	2012	4b	–
N12-0605	Switzerland	Sausage	2012	4b	–
N12-0794	Switzerland	Human	2012	4b	–
N12-0869	Switzerland	Human	2012	4b	–
N12-0973	Switzerland	Meat	2012	4b	–
N12-1338	Switzerland	Human	2012	4b	–
N12-1339	Switzerland	Meat	2012	4b	–
N12-1387	Switzerland	Human	2012	4b	–
N12-1655	Switzerland	Human	2012	4b	–
N12-1665	Switzerland	Environment	2012	4b	–
N12-1730	Switzerland	Human	2012	4b	–
N12-1772	Switzerland	Milk	2012	4b	–
N12-1796	Switzerland	Human	2012	4b	–
N12-1859	Switzerland	Food	2012	4b	–
N12-1916	Switzerland	Environment	2012	4b	–
N12-1996	Switzerland	Environment	2012	4b	–
N12-2185	Switzerland	Environment	2012	4b	–
N12-2378	Switzerland	Human	2012	4b	–
N12-2447	Switzerland	Meat	2012	4b	–
N13-0047	Switzerland	Milk	2013	4b	–
N13-0177	Switzerland	Human	2013	4b	–
N13-0677	Switzerland	Human	2013	4b	–
N13-0771	Switzerland	Human	2013	1/2a	–
N13-0772	Switzerland	Human	2013	4b	P
N12-0036	Switzerland	Human	2012	1/2a	–

^aFinland: University of Helsinki, Department of Food Hygiene and Environmental Health strain collection; Switzerland: University of Zürich, Institute for Food Safety and Hygiene strain collection.

^bNA, not available.

^cUsed also as pilot (P) or control (C) strain. –, not applicable.

TABLE 3 Summary of growth parameter estimation for *Listeria monocytogenes* strains at 9.0% NaCl

Method used for strain fitting	% of strains (<i>n</i> = 388) fitted	Best model relative to others used for % of strains (<i>n</i> = 388) ^b	% of $\lambda^c < 0$ of fitted strains	Range for fitted strains ^d			
				μ	λ	MaxOD	AUC
Spline ^a	100	NA	3.1	0.012–0.18	–4.3–12	0.25–1.0	2.7–8.8
Gompertz	88	0.52	13	0.011–0.19	–8.0–57	0.41–265	3.0–8.7
Modified Gompertz	41	39	0	–0.08–0.09	3.9–15	0.034–1.4	3.7–8.2
Logistic	97	19	15	0.011–0.65	–7.7–33	0.36–19	2.7–8.7
Richards	77	42	8.3	0.013–0.13	–2.9–8.2	0.26–2.1	2.7–8.8

^aModel-free splines according to Kahm et al. (49).

^bThe relative quality of models Gompertz, modified Gompertz, logistic, and Richards to fit each of the strains was assessed by using the Akaike information criterion.

NA, not applicable.

^c λ , lag phase.

^d μ , maximum growth rate; MaxOD, maximum OD₆₀₀; AUC, area under the curve.

(ii) Outliers and batch effects were assessed to remove technical variability.

During the Bioscreen experiments at 9.0% NaCl, a slight oscillation of OD₆₀₀ values appeared due to the measurement technology (Fig. S2), and therefore OD₆₀₀ values at 1-h intervals were used in the analyses. The replicates of the test strains (Table 2) placed in the outermost wells of the Bioscreen honeycomb plate exhibited reduced growth, which increased the variation in the data set; mean coefficient of variance (CV) was higher for replicates from all wells (4.4%) than for those from inner wells (3.2%). Similarly, reduced growth was observed for the replicates of the two control strains (Table 2) placed in the outermost wells throughout all experimental runs (Fig. S3). In addition, variation between the replicates of the control strains was larger in the outermost wells than in the inner wells for Bioscreen runs performed by both laboratory technician A (CV 3.9% versus 3.6%, respectively) and laboratory technician B (CV 6.0% versus 3.4%, respectively), indicating that conditions in the outermost and inner wells differed. The outermost well data were therefore removed as outliers, after which an average of four replicate growth measurements remained per strain to calculate mean OD₆₀₀ values. Furthermore, both control strains grew systematically more poorly in the experiments performed by laboratory technician A than by laboratory technician B, who performed the experiments for 90 and 298 of the test strains, respectively (Fig. S4 and S5). Although both laboratory technicians used the same growth media and calibrated pipettes, differences could have arisen from different working techniques, speed, and the “signature” of each laboratory technician, even between these two highly experienced individuals. Removal of this batch effect was achieved through a normalization of the test strain mean OD₆₀₀ values at the measured time points (Fig. S6 to S10). To avoid excessive technical variation, we propose (i) pipetting blank test broth into the outermost wells of a Bioscreen honeycomb plate (see the example of a loading map in Fig. S11) and (ii) assigning one individual to perform all Bioscreen experiments of a particular study.

Selection of a suitable method for growth parameter calculation (step 2). (i) Kinetic growth parameters for the experimental data set varied significantly between different calculation methods. The proportion of strains fitted by each parameter calculation method and the ranges of their parameter values varied notably between the methods used (Table 3 and Fig. 1). Gompertz and logistic models provided negative and excessively large parameter values and included a larger amount of negative values for λ than the other methods (Table 3). All parameter calculation methods differed significantly from one another in pairwise comparison of their respective parameter values for at least two of the parameters λ , μ , MaxOD, and AUC (Friedman’s two-way analysis of variance by ranks for the strains fitted by all methods, *n* = 128, *P* ≤ 0.001).

(ii) Model-free splines provided the most suitable growth parameters for this data set. When assessing the quality of each growth model relative to the other models using the Akaike information criterion (AIC), Richards and modified Gompertz provided the best fit for the largest number of strains, but neither could fit all of the

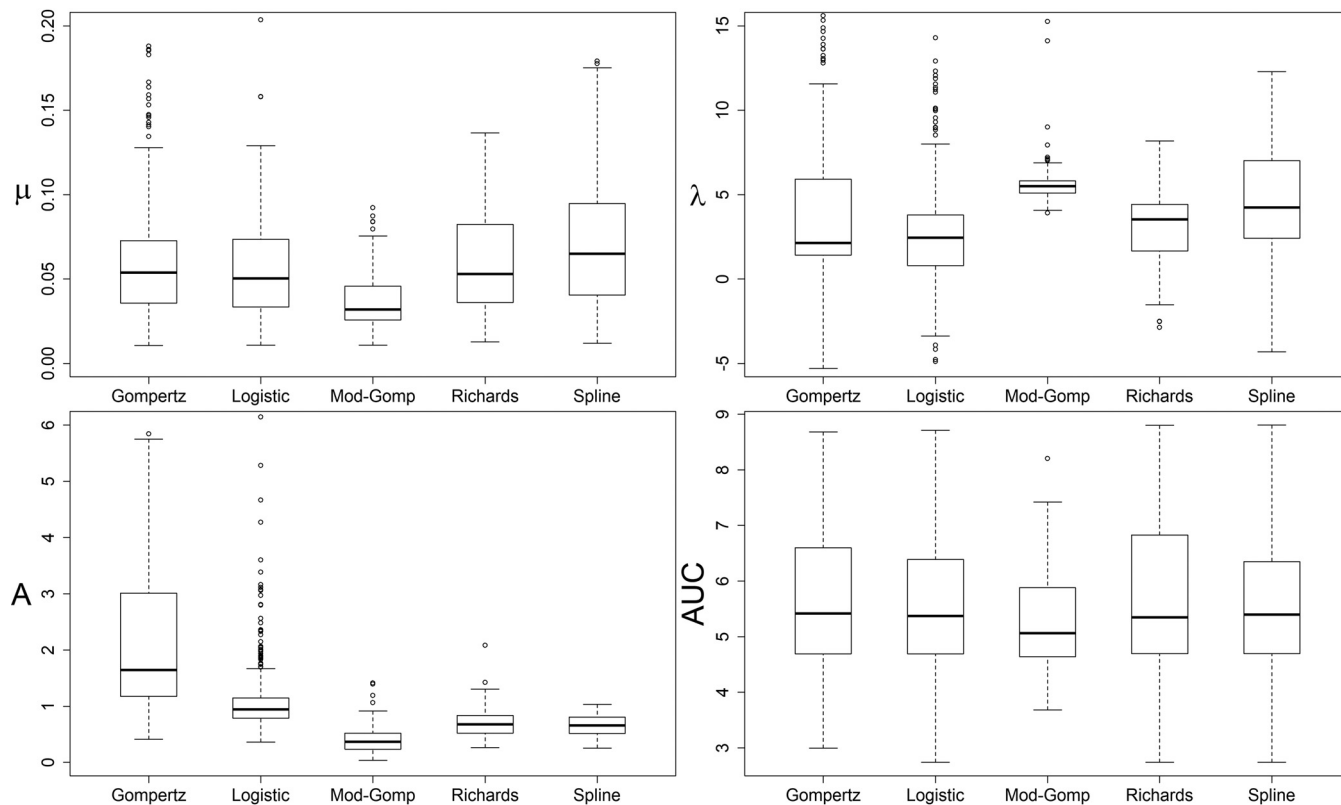


FIG 1 Distribution of growth parameter values (μ = maximum growth rate, λ = lag phase, MaxOD = maximum OD₆₀₀, AUC = area under curve) among the fitted *L. monocytogenes* strains by Gompertz, logistic, modified Gompertz, and Richards models and model-free spline. Whiskers depict the highest and lowest quartiles, and the box depicts the middle quartiles divided by median line. y axes have been cut to focus on quartiles and remove the most extreme individual outliers.

strain growth curves (Table 3). With 77% of the curves fitted, Richards appeared to be a more suitable model for most strains than modified Gompertz. Conversely, the model-free spline provided growth parameters for all tested strains, included relatively few negative λ values, and matched the experimental data points well upon visual inspection. Each spline parameter also correlated significantly with its equivalent Richards model parameter (Spearman's rho 0.933, 0.722, 0.927, and 1.0 for μ , λ , MaxOD, and AUC, respectively; $P < 0.001$). Spline parameters were therefore deemed adequate to describe the biological variation of our data set and were chosen for the comparison of strain growth patterns, for the purpose of which the negative spline λ values were changed to 0 after the calculation of all parameters.

Comparison of growth patterns between strains (step 3). (i) Growth curve and parameter visualization facilitated the identification of growth patterns.

Upon visual inspection of strain growth curves, 9.0% NaCl resulted in evident variation of growth patterns between *L. monocytogenes* strains: tolerant strains displayed higher growth rates and maximum optical densities and shorter lag phases than susceptible strains (Fig. S9). This intraspecies variability indicated that NaCl tolerance of *L. monocytogenes* could be characterized as a continuous phenotype, measured by gradually increasing growth rate and maximum optical density and decreasing lag phase duration between strains (Fig. S9). Thereby, a classification of the strains into poor, average, and good growers was *a priori* deemed appropriate to summarize this ordinal variability in the strain-specific growth patterns. The variability between the strain spline parameters was smallest for μ and MaxOD, and highest for λ and AUC (Table 3 and Fig. 1), and all four parameters correlated with each other (Spearman's rho > 0.8 ; $P < 0.001$). As the integral of OD₆₀₀ measurements, AUC consolidates the other three parameters and was thus hypothesized to be a suitable parameter for strain comparison.

(ii) The parameter AUC differentiated *L. monocytogenes* growth patterns in salt stress. Four alternative hierarchical clustering outcomes using different variables were investigated to select the best clustering method. The alternative clusterings (numbered 1 to 4) included the following sets of variables: OD₆₀₀ values from 0 to 17 h (alternative 1), spline growth parameters λ , μ , MaxOD, and AUC (alternative 2), spline growth parameters λ , μ , and MaxOD (alternative 3), and spline growth parameter AUC (alternative 4). For each alternative clustering outcome, dendrograms based on Ward linkage suggested a division of the entire data set into two clusters, both of which could be further divided into two clusters. The alternative clustering outcomes 1 to 4 were initially visualized by their strain growth curves in Microsoft Office Excel 2013 (Fig. S12 to S15) and assessed for their ability to recognize the *a priori* deemed ordinal growth patterns. Only alternative 4, which utilized AUC as the clustering variable, resulted in ordinal division of the growth clusters (Fig. S15) and thus was chosen for the growth pattern classification of *L. monocytogenes* strains.

Biological interpretation of the discovered differences (step 4). (i) The growth of *L. monocytogenes* strains varied from poor to good at 9.0% NaCl. Clusters of strains displaying poor, average, and good growth at 9.0% NaCl, including 182, 116, and 90 strains, respectively, were labeled accordingly (Fig. 2). Variation between the strains within each cluster (CVs “good” [52%], “average” [62%], and “poor” [63%]) was higher than variation between the individual replicates of a strain (CVs 0.0 to 36%). The cluster “poor” was characterized by the lowest values of all growth parameters, whereas “good” included strains with the highest MaxOD and AUC values (Fig. 3). The “average” cluster included strains with highly varying μ and λ , but intermediary MaxOD and AUC (Fig. 3). The spline growth parameters differed significantly from one another between the three growth clusters (Kruskal-Wallis test, $P < 0.001$), except for μ of clusters “average” and “good” (Fig. 4).

(ii) Strain variability at 9.0% NaCl uncovered serovar- and lineage-dependent salt stress tolerance of *L. monocytogenes*. The best-growing *L. monocytogenes* strains belonged to the serovars 1/2b and 4b of lineage I, and the poorest growers to serovars 1/2a, 1/2c, and 3a of lineage II (Fig. 2). Serovar 4b was significantly overrepresented in the cluster “good,” and 1/2b in the “good” and “average” clusters, whereas serovar 1/2a was significantly underrepresented in cluster “good” and 1/2c in clusters “good” and “average” (z-test, $P < 0.05$; Table 4). In our data set, the serovar 1/2a strains were characterized by a large variation of their growth parameter values (Fig. 3). Serovars 1/2b and 4b exhibited the highest and 1/2c and 3a the lowest growth parameter median values (Fig. 3). Each serovar differed significantly from another serovar in the pairwise comparison of their growth parameter values for at least two of the parameters λ , μ , MaxOD, and AUC (Kruskal-Wallis test, $P \leq 0.03$, Fig. 5).

DISCUSSION

This growth data assembly and analysis protocol for *L. monocytogenes* salt stress tolerance studies provides a systematic tool and checklist for the investigation of strain variability of stress tolerance (Table 1). Strain variability of growth ability is an important practical factor to consider, for instance, prior to conducting quantitative microbiological risk assessments (QMRA), as it has implications on the selection of strains for challenge tests estimating shelf-lives of food products (31, 59). Our protocol compiles a systematic approach to estimate the scope of intraspecies stress tolerance variability and identify strains that grow better than others under examined stress conditions, aiding the selection of suitable strains for QMRAs. Phenotypic knowledge gained by our approach can also be used to identify exceptionally tolerant or susceptible strains for further analysis of underlying dynamics and mechanisms of stress tolerance.

Although tools for data generation and handling depend upon the preferences of each researcher, the considerations to ensure the reliability of extensive growth studies are the same. To begin with, our results highlight the importance of incorporating an appropriate level of stress when aiming to examine strain variability of stress tolerance. Although NaCl concentrations of 8% and below have been used in *L. monocytogenes* stress experiments to illustrate strain variability (33, 35, 60, 61), our results imply that a

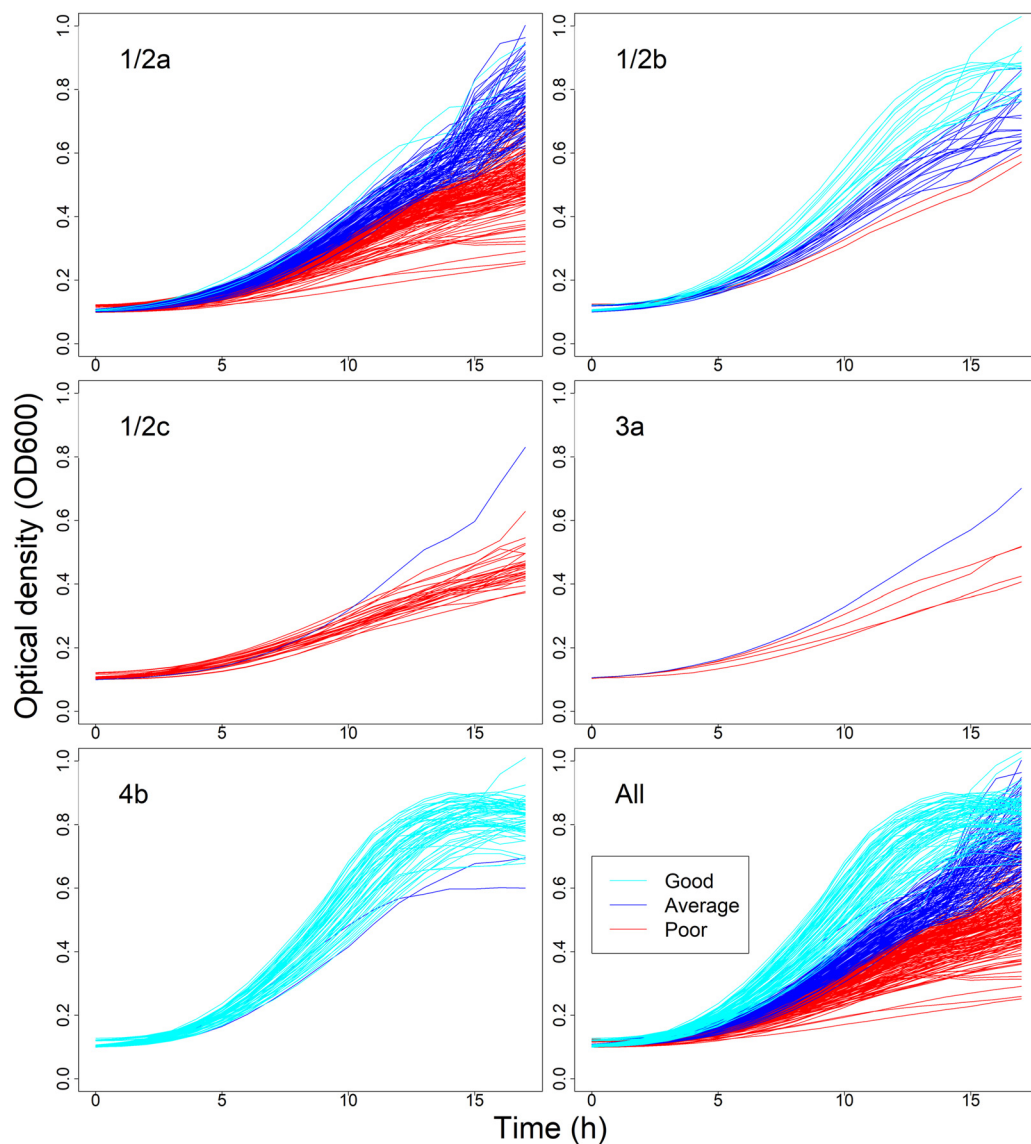


FIG 2 *L. monocytogenes* strains ($n = 388$) by serovar and growth cluster in 9.0% NaCl. The panels depict strain growth curves belonging to serovars 1/2a, 1/2b, 1/2c, 3a, 4b, and all serovars. Division to growth clusters is visualized by color, as indicated in legend.

high NaCl concentration was required to differentiate notable variation in strain growth ability. The ability of our study to affirm, on an unprecedentedly large scale, some previously described serovar- and lineage-dependent osmotic stress phenomena (30, 61, 62, 77) emphasizes the relevance and reliability of our data assembly and analysis protocol for *L. monocytogenes* stress tolerance investigations.

When assessing the strain variability of growth ability, we must initially make sure to draw conclusions on biological, not technical, variability (63) by the use of technical and biological controls and careful data exploration. Technical variation and data exploration steps should be discussed in growth experiment publications to review the reproducibility and reliability of the results. Measurement oscillation due to the machinery and differences between strain replicates depending on their placement in the Bioscreen honeycomb plate were resolved by removing outliers. The reduced growth of the replicates placed alongside the outermost rim of the plate might have been due to, for instance, a higher rate of evaporation of liquid from the 300 μ l of culture dilution placed in the outermost wells than in the inner wells, which would have caused an

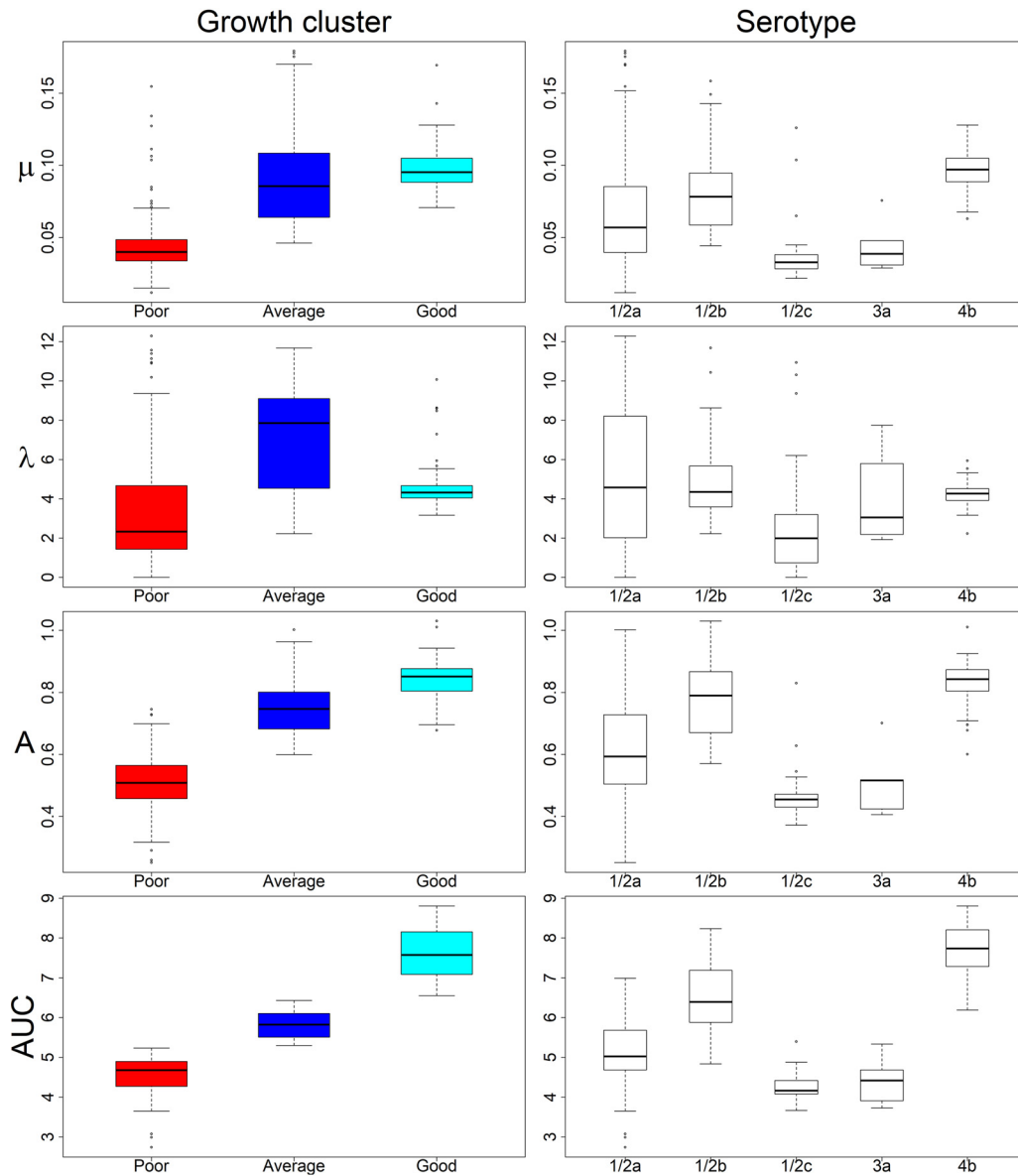


FIG 3 Variation of spline growth parameters of *L. monocytogenes* strains ($n = 388$) in 9.0% NaCl within identified growth clusters and serovars. Whiskers indicate the highest and lowest quartiles, and the box depicts the middle quartiles divided by median line. Dots indicate outliers. Growth parameters are calculated from spline fit: μ corresponds to growth rate (OD_{600} U/h), λ to lag time (h), MaxOD to maximum growth (OD_{600} units), and AUC to the area under curve calculated by an integral.

irregular increase of the NaCl concentration in the outermost wells. Although the overall variation between the control strain replicates was negligible, the systematic well effect indicated a biologically noteworthy phenomenon, which should be considered when designing loading maps for the Bioscreen honeycomb plate (Fig. S11). Due to the extreme precision required in the labor-intensive pipetting of the Bioscreen honeycomb plates, individual differences in, e.g., pipetting techniques may result in systematic variation, even between highly experienced professionals. Following standard statistical procedures, the observed batch effect between the two laboratory technicians was removed by setting one group as a reference and normalizing the other by adding the calculated difference of the mean OD_{600} values to its OD_{600} values. The emphasis of our experiments on phenotypic variability was on the comparison of growth curves and kinetic parameters proportionate to one another, not on their

μ			λ		
	Average	Good		Average	Good
Poor	<0.001*	<0.001*	Poor	<0.001*	<0.001*
Average		0.2	Average		<0.001*
MaxOD			AUC		
	Average	Good		Average	Good
Poor	<0.001*	<0.001*	Poor	<0.001*	<0.001*
Average		<0.001*	Average		<0.001*

FIG 4 Kruskal-Wallis test *P* values for the associations of each spline growth parameter between the growth clusters “poor,” “average,” and “good” determined for *L. monocytogenes* strains (*n* = 388) at 9.0% NaCl. Significant associations are indicated with an asterisk. μ corresponds to growth rate (OD₆₀₀ U/h), λ to lag time (h), MaxOD to maximum growth (OD₆₀₀ units), and AUC to the area under curve calculated by an integral.

absolute values. Thereby, the performed normalization was adequate, as well as imperative, to create comparable distributions and enable the identification of strains with dissimilar growth patterns.

The second step of our data assembly and analysis protocol was to choose a suitable method to quantify growth ability for the comparison of strains, a purpose which did not require substantial assumptions on the underlying kinetics of growth. Our analysis revealed marked differences in the ability of the tested growth models and model-free splines to fit all growth curves, as well as significant variation in the respective parameter estimates by the different methods. Unlike in previous studies (38, 41, 64), none of the tested growth models were adequate to describe our entire data set, but model-free splines (49) fitted all the curves. The spline parameters correlated with those derived from the Richards model which has been indicated to suit absorbance measurement data (38, 64). When utilizing absorbance to evaluate strain differences, it is valuable to note that an apparent increase in cell size without increase in cell numbers may increase the OD at highly stressful conditions (42). Under NaCl stress, *L. monocytogenes* cells may elongate to form filaments which are divided by septa (65–67) and consist of several normal-sized cells on the verge of division (67). Consequently, a potential increase in OD caused by such filaments would still reflect cell numbers.

The third step of our data assembly and analysis protocol included the comparison of alternative hierarchical clustering outcomes to facilitate the strain comparison and classification of growth patterns. In foodstuffs, the delay at the beginning of growth or the maximum growth level reached may sometimes be more important than growth rate, which can be similar for curves with differing λ and MaxOD (44). While μ may be

μ					λ				
	1/2b	1/2c	3a	4b		1/2b	1/2c	3a	4b
1/2a	0.03*	<0.001*	1.0	<0.001*	1/2a	1.0	<0.001*	1.0	1.0
1/2b		<0.001*	0.1	0.1	1/2b		<0.001*	1.0	1.0
1/2c			1.0	<0.001*	1/2c			1.0	0.001*
3a				0.002*	3a				1.0
MaxOD					AUC				
	1/2b	1/2c	3a	4b		1/2b	1/2c	3a	4b
1/2a	<0.001*	<0.001*	1.0	<0.001*	1/2a	<0.001*	<0.001*	0.9	<0.001*
1/2b		<0.001*	0.007*	0.6	1/2b		<0.001*	0.001*	0.03
1/2c			1.0	<0.001*	1/2c			1.0	<0.001*
3a				<0.001*	3a				<0.001*

FIG 5 Kruskal-Wallis test *P* values for the associations of each spline growth parameter between the tested serovar 1/2a, 1/2b, 1/2c, 3a, and 4b *Listeria monocytogenes* strains (*n* = 388) at 9.0% NaCl. Significant associations are indicated with an asterisk. μ corresponds to growth rate (OD₆₀₀ U/h), λ to lag time (h), MaxOD to maximum growth (OD₆₀₀ units), and AUC to the area under curve calculated by an integral.

TABLE 4 Numbers and proportions of *Listeria monocytogenes* strains of each serovar belonging to three 9.0% NaCl growth clusters^a

Serovar	No. (%) of strains by growth cluster		
	Poor	Average	Good
1/2a	144 (59.8) ^A	93 (38.6) ^A	4 (1.7) ^B
1/2b	2 (5.3) ^A	19 (50.0) ^B	17 (44.7) ^B
1/2c	32 (97.0) ^A	1 (3.0) ^B	0 (0.0) ^B
3a	4 (80.0) ^A	1 (20.0) ^A	0 (0.0) ^A
4b	0 (0.0) ^A	2 (2.8) ^A	69 (97.2) ^B

^aThe numbers of *L. monocytogenes* strains in each serovar (1/2a, $n = 241$; 1/2b, $n = 38$; 1/2c, $n = 33$; 3a, $n = 5$; 4b, $n = 71$) belonging to the 9.0% NaCl growth clusters "poor," "average," and "good" are presented. The same superscript capital letter indicates proportions that do not differ significantly at the 0.05 level within rows (z-test with Bonferroni correction).

an often-used practically convenient metric, the variation of both λ and MaxOD within our data set of 388 *L. monocytogenes* strains advocated for the consideration of the entire growth pattern in strain comparison. AUC was deemed an appropriate variable to differentiate between *L. monocytogenes* strain growth patterns under salt stress, as it suits to summarize and compare entire growth patterns in cases where λ increases while μ and MaxOD decrease in an ordinal fashion between the different strains. However, growth curves with a high μ and MaxOD could reach a similar AUC to curves with a low μ and MaxOD, if the λ were long for the former and short for the latter. Consequently, no single all-purpose solution exists, and both statistical tests and intuitive reasoning should be used to classify the overall growth patterns present in a given data set, keeping in mind the suitability of the chosen approach with respect to the data and purpose of the study. The identified ordinal growth clusters "poor," "average," and "good" displayed, on average, significantly different growth parameters, but overlapped slightly. Thereby, a strict distinction into groups of differing growth patterns was not possible, which indicates that *L. monocytogenes* growth under salt stress should be approached as a continuous rather than a categorized phenotype.

The final step of our data assembly and analysis protocol was to draw biological interpretations on stress tolerance based on the strain differences observed in the growth experiments. The clusters of *L. monocytogenes* strains displaying poor, average, and good growth appeared to provide a meaningful representation of the differing growth patterns, as they also captured biological differences of salt stress tolerance among the studied *L. monocytogenes* serovars. In our experiments, *L. monocytogenes* lineage I strains (serovars 1/2b and 4b) grew significantly better than lineage II strains (serovars 1/2a, 1/2c, and 3a) at 9.0% NaCl, which indicates that the ability to grow under NaCl osmotic stress is connected to the phylogenetic composition of the species. Serovar 4b is frequently associated with human listeriosis, while occasional human outbreaks by serovar 1/2b have been reported (23, 68), and serovar 1/2a is common in human cases in the Nordic countries (69, 70). The apparent overrepresentation of 1/2a and 1/2c among isolates from food-related sources and their underrepresentation in human cases in some countries have been hypothesized to be related to attenuated virulence of many lineage II strains (22, 27, 71) or their endurance of environmental conditions in food-associated environments (28, 61). Our results are particularly congruent with the former hypothesis, since *L. monocytogenes* virulence has been linked to its osmotolerance (17, 52, 53), but do not conflict with the latter hypothesis either, as 9.0% NaCl is a condition rarely encountered in the food chain, except for brining solutions.

In our experiments, many *L. monocytogenes* strains were able to grow relatively well in saline conditions that are among the most extreme found in the food chain. However, food matrices and establishments form complex systems that involve multiple varying stress conditions. Although *L. monocytogenes* is highly tolerant toward several conditions used for controlling bacterial growth during food production and storage (14, 15), the overall stress exposure and response depend on all environmental

conditions, such as temperature, pH, water activity, nutrients, and sanitizing agents (3, 4, 30, 35). The growth patterns described here for NaCl osmotic stress may therefore be altered by other factors present in food systems. Nevertheless, the notable strain variability during salt stress uncovered by the present study implies that *L. monocytogenes* has a large reservoir of growth ability. This may help the population adapt to prevailing environmental conditions and hence conquer various niches, a strategy referred to as “bet-hedging” (72). Based on our observations, high salinity environments could provide a growth niche particularly for *L. monocytogenes* strains of lineage I.

MATERIALS AND METHODS

Strain collection. *L. monocytogenes* wild-type strains ($n = 388$, Table 2) isolated from animals, raw materials, food, processing environment, and patients were used. Stock cultures were stored at -70°C in bead tubes. Serotyping was performed using the Listeria Antisera Set with O- and H-factor antisera (Denka Seiken, Tokyo, Japan) according to the manufacturer's instructions. Serovars of 14 strains yielding uncertain results were confirmed using multiplex-PCR as described elsewhere (73). Strains represented serovars 1/2a ($n = 241$), 1/2b ($n = 38$), 1/2c ($n = 33$), 3a ($n = 5$), and 4b ($n = 71$).

Optimization of the stress condition. To select an appropriate osmotic stress condition, separate pilot study broths were prepared by adding NaCl to final concentrations of 6.5, 7.5, and 8.5% into BHI broth (Oxoid, Cheshire, England). In the pilot, 20 *L. monocytogenes* strains (Table 2) were chosen by generating random numbers for each strain in R and selecting 10 strains with the highest random number values from both lineage I and II while ensuring even distribution of serovars. A single colony of each pilot strain was grown in duplicate (technical replicates) in each NaCl concentration using a Bioscreen C microbiology reader (Growth Curves, Helsinki, Finland) following the protocol described for the growth experiments at salt stress below. Growth curves were visualized in Microsoft Office Excel 2013 to determine which condition displayed increased variation among strain growth ability, i.e., sufficient stress to distinguish differences between strains (Fig. S1).

Growth experiments of NaCl tolerance. Stock cultures plated on blood agar were incubated at 37°C for 24 h. Three separate colonies (biological replicates) were individually inoculated into 5 ml of BHI broth and incubated at 37°C with shaking (250 rpm) for 18 h to reach an average cell count of 10^9 CFU ml^{-1} . Overnight cultures were diluted (1:100) in 1.5 ml of fresh BHI broth including 9% NaCl (test broth), and $300\ \mu\text{l}$ of each diluted culture was loaded into separate wells of a 100-well honeycomb plate in duplicate (technical replicates). Four wells were left with blank test broth as controls for no growth. The three biological replicates of the same strain were loaded at scattered positions on the plate. A cover was used on the plate to prevent evaporation.

Growth experiments were performed in the Bioscreen C microbiology reader at 37°C for 17 h, and the optical densities at 600 nm (OD_{600}) were automatically measured at 15-min intervals. Two control strains were used in each Bioscreen run throughout the study; three individual isolates of the two controls were grown in the same positions of the honeycomb plate to exclude systematic technical variation between separate Bioscreen runs. The order in which the 388 strains were tested throughout the experiments was determined via random numbers generated in R version 3.4.3 (74). During each experiment the growth curves were routinely visualized at the computer connected to the Bioscreen machine. At the end of the experiment, the wells showing no growth or exceptional shape of growth curve were sampled and cultured on blood agar (37°C for 24 h) to detect lack of sample or potential contaminants. If this culture displayed no growth (empty well) or mixed growth (contaminated well) based on visual inspection, the data from such wells were disregarded and the strains concerned retested.

Raw data exploration. Data exploration was performed in Microsoft Office Excel 2013 and R version 3.4.3. To mitigate the slight oscillation of the measured OD_{600} values, which appeared due to the measurement technology (Fig. S2), the OD_{600} values at 1-h intervals were used in the analyses. The replicate growth curves of the two control strains were visually inspected to evaluate potential systematic differences within the experiments. As differences were observed between the replicates grown in the inner and outermost wells of the Bioscreen honeycomb plate (Fig. S3 to S5), the outermost-well replicates were excluded as outliers. Initially, three biological and two technical replicate sets of OD_{600} values (six in total) were obtained for every strain. The technical and biological replicate curves of each strain were visualized, and their similarity was evaluated. The strain was retested if individual replicates displayed deviations that could not be explained by contamination or their position in an outermost well.

Identification and normalization of batch effects. The presence of potentially confounding batch effects, i.e., systematic nonbiological differences between sample groups, was estimated using the control strain replicates included in each Bioscreen experiment. Laboratory technicians A and B conducted the experiments for 90 and 298 strains, respectively, and an evident batch effect was observed between them (Fig. S3 to S6). A nonparametric approach was employed in the batch effect normalization to avoid assuming any specific parametric shape of the growth data over time, as the observed batch effect itself was nonlinear (Fig. S7 to S9). At each time point t , the mean ODs— mean_{L_A} and mean_{L_B} —of control samples belonging to each group were calculated. When L_A and L_B are the two groups of control samples defined by the batch (laboratory technician), the difference— $\Delta_t = \text{mean}_{L_B} - \text{mean}_{L_A}$ —is the estimated batch effect at time point t . The normalization was performed by adding Δ_t to all samples belonging to batch L_A . The custom R scripts for normalization are provided as supplemental material (Text S10 in the supplemental material).

Growth parameters and data visualization. Growth modeling and growth curve visualization were performed with R version 3.4.3 (74). Using the normalized mean OD₆₀₀ values, kinetic growth parameters were determined for each strain with the R package *grofit* (49) (archived from the CRAN repository on 17 June 2018) using default settings. The parameters were calculated with logistic, Gompertz, modified-Gompertz, and Richards models, as well as model-free splines based on cubic spline interpolation, for which the smoothing parameter, *smooth.gc*, was set to default (NULL), allowing the program to select an optimal value via cross-validation techniques (49). The resulting five sets of kinetic growth parameters included λ (lag phase duration), μ (growth rate), MaxOD (maximum OD₆₀₀), and AUC (area under the curve). In addition, the program assesses and selects the best-fitting growth model for each individual strain by the comparison of nonlinear least-squares regression of each model via Akaike information criterion (AIC) (75). Since the same parameter calculation method had to be used for all strains to enable their comparison, we selected the best-fitting parameter calculation method based on the following criteria: (i) the ability to calculate parameters for all growth curves, (ii) a low proportion of negative values of λ (76), and (iii) a good fit on visual inspection of the modeled curve and data points. After the selection of the best method, the negative λ values were transformed to 0, because negative measurements of time duration are not meaningful. The custom R scripts for growth curve and parameter visualization are provided as supplemental material (Text S10).

Statistical analyses. Statistical analyses were performed with R version 3.4.3 and IBM SPSS Statistics for Windows version 25.0 (IBM, Armonk, NY). The variability between the replicates of each strain and the strains within each growth cluster was estimated using the coefficient of variation (CV, i.e., ratio of standard deviation to mean) calculated from their respective OD₆₀₀ values. Nonnormal distribution of the growth parameter values was confirmed by Kolmogorov-Smirnov test and visualization of normal Q-Q plots. Friedman's two-way analysis of variance by ranks and Spearman's rho were used to compare the growth parameters of the different models and model-free splines, while removing strains with missing values from the calculation of these statistics. Hierarchical clustering with Ward's method by squared Euclidian distance was utilized to classify the strain growth patterns. Four alternative clustering outcomes were formed using either the strain mean-OD₆₀₀ values from 0 to 17 h at 1-h intervals (alternative 1) or a selection of the best-fitting growth parameters (alternatives 2 to 4 described in Results). Utilizing the visualization of the cluster growth curves in Microsoft Office Excel 2013, we chose the clustering alternative that best captured the growth patterns emerging from the data set. A Kruskal-Wallis test was used to investigate the associations between serovar, growth cluster, and growth parameters.

SUPPLEMENTAL MATERIAL

Supplemental material is available online only.

SUPPLEMENTAL FILE 1, PDF file, 1.2 MB.

SUPPLEMENTAL FILE 2, PDF file, 0.02 MB.

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