Predicting variable-temperature response of non-dormant seeds from constant-temperature germination data

STUART P. HARDEGREE, STEVEN S. VAN VACTOR, FREDERICK B. PIERSM AND DEBRA E. PALMQUIST

Abstract

The objective of many laboratory-germination experiments is to develop insight into the process of field establishment. It is relatively difficult, however, to infer potential field response from laboratory data given the enormous spatial and temporal variability in seedbed microclimate. Previous studies have attempted to survey large numbers of alternating day/night temperature regimes in order to estimate germination response to potential conditions of field microclimate. The objectives of this study were to estimate the errors associated with prediction of variable-temperature germination response from fewer, constant-temperature germination data. Non-dormant seeds of thickspike wheatgrass [Elymus lanceolatus (Scribn. and J.G. Smith) Gould], bluebunch wheatgrass [Pseudoroegneria spicata (Pursh) Löve], Sandberg bluegrass (Poa sandbergii Vasey), and bottlebrush squirreltail [Elymus elymoides (Raf.) Swezey] were germinated under constant, alternating-constant and sine-wave temperature regimes. Predicted and measured cumulative-germination response generally coincided to within a day for most temperature treatments except for the most slowly germinating subpopulations of seeds. Thermal response models can be parameterized from relatively few experimental data but provide predictive inferences relevant to a wide number of variable-temperature conditions.

Key Words: seed, model, rate

Most research concerning germination response of cool-season range grasses has been conducted under either constant or alternating-constant temperature conditions (Wester 1991). Alternating-constant temperature experiments can mimic diurnal temperature fluctuation in the seedbed but a comprehensive survey of potential germination response requires that a large number of treatment combinations be tested (Young and Evans 1982). An alternative approach is to develop predictive models for estimating variable-temperature germination response (Garcia-Huidobro et al. 1982b, Benech Arnold et al. 1990, Roundy and Biedenbender 1996). Such models can be developed from constant-temperature germination data if it is assumed that instantaneous germination rate is independent of diurnal temperature pattern (Garcia-Huidobro et al. 1982b, Benech Arnold et al. 1990). This assumption may not be correct for species that require fluctuating-temperature conditions to release dormancy (Thompson 1974, Thompson and Grime 1983, Murdoch et al. 1989, Benech Arnold et al. 1990). Application of thermal response mod-
els to rangeland grass species has received relatively little attention in the literature (Jordan and Haferkamp 1989, Roundy and Biedenbender 1996). Roundy and Biedenbender (1996) used constant-temperature germination data to predict median germination response of 8 warm-season grasses under 3 variable-temperature regimes. Predicted and measured median germination times in their study averaged within 0.6, 1.3, and 3.6 days for thermal regimes representative of summer, spring, and winter, respectively.

The purpose of this study was to evaluate variable-temperature effects on the cumulative-germination response of 4 cool-season range grasses. We hypothesized that cumulative-germination response under alternating-constant and sine-wave temperature patterns could be predicted from constant-temperature germination data. Calibration of thermal models would make it possible to predict cumulative-germination response under a wide number of potential field-temperature regimes without having to survey large numbers of alternative thermal treatments. The species in question, thickspike wheatgrass [Elymus lanceolatus (Scribn. and J.G. Smith) Gould], bluebunch wheatgrass [Pseudoroegneria spicata (Pursh) Löve], Sandberg bluegrass (Poa sandbergii Vasey) and bottlebrush squirreltail [Elymus elymoides (Raf.) Swezey], were selected because they have been identified by the United States Department of the Interior, Bureau of Land Management as high-priority species for restoration of deteriorated rangelands in the Great Basin and Columbia Plateau.

Materials and Methods

Seeds for this experiment were purchased from Stevenson seed company (Ephraim, Ut.) which collected the seeds in 1991. Air-dry seeds were stored in cloth bags at room temperature during the course of the study.

Two experiments were conducted to evaluate variable-temperature effects on germination response. Experiment 1 was conducted between Oct. 1992 and Feb. 1993 to measure germination response under constant, alternating-constant and sine-wave temperature patterns. Experiment 2 was conducted between July 1993 and Nov. 1993 to measure germination response at constant temperature and under sine-wave temperature patterns of different amplitude.

Germination experiments were conducted inside 15 programmable environmental chambers (Hardegree and Burgess 1995). These chambers consisted of small refrigerators that were enhanced with incandescent and fluorescent lights, fans, and electrical resistance heaters. An electronic data acquisition/control system maintained temperature control to within ± 0.5°C. Chamber lights were programmed to have a 12 hour photoperiod. The temperature inside each box was monitored every 10 seconds for purposes of temperature control and an average temperature calculated and recorded for every 15 minute period.

In Experiment 1, separate environmental chambers were programmed to maintain constant temperatures in 5° increments between 5 and 35°C; alternating-constant night/day temperature regimes of 5/15, 10/20, 15/25, 20/30, and 25/35°C; and sine-wave patterns with an amplitude of 10°C and mean daily temperatures of 10, 15, 20, 25, and 30°C. In Experiment 2, environmental chambers were programmed to simulate constant temperatures in 5° increments between 5 and 35°C; and sine-wave patterns with a mean temperature of either 15, 20, or 25°C and temperature amplitudes of either 5, 10, 15, or 20°C. Each thermal regime was replicated 3 times in different germinators. As there were 17 separate thermal regimes in Experiment 1 and 19 in Experiment 2, all replicate treatments could not be evaluated simultaneously. Treatment order was randomized and treatments were evaluated in groups of 15 until all had been replicated 3 times.

Seeds were germinated on a cellulose dialysis membrane at the bottom of germination cups that were in contact with a water reservoir inside a clear-plastic snap-top vial (Hardegree and Emmerich 1992). Free water on top of the membrane was kept to a minimum by controlling the height of water in the reservoir and by daily suction when necessary.

One vial of each species was arranged randomly within each of 6 replicated blocks inside each germination chamber. Thirty seeds of each species (35 seeds for Sandberg bluegrass) were placed in each germination cup and dusted with Daconil fungicide (2,4,5,6-tetrachloro-1,3-benzenedicarbonitrile, wettable powder) at the beginning of a given experimental run. Additional fungicide was applied if the seeds developed fungal growth during the experiment. Germination vials were monitored daily for 21 days and seeds counted and removed when they exhibited radicle extension ≥ 2 mm.

Characterization of Subpopulation Germination Rates at Constant Temperature

The constant-temperature germination data were used to develop regression equations for estimating germination response under the alternating-constant and sine-wave temperature regimes. Cumulative-germination percentages were calculated for every combination of species and treatment for every day. The cumulative data were numerically transformed by dividing germination percentages by a scaling factor (Ellis et al. 1986). The scaling factor was equal to the maximum-mean-germination percentage achieved in the most favorable temperature treatment for that species. This transformation adjusted germination percentages for each species to a common scale with a maximum value of 100%. The scaling factors were 0.95 for thickspike wheatgrass, 0.91 for bluebunch wheatgrass, 0.82 for Sandberg bluegrass and 0.86 for bottlebrush squirreltail. Germination counts were pooled by species within each germinator box and the 3 within-box averages considered treatment replicates for model development and analysis.

For modeling purposes, the seed population was considered to be composed of subpopulations based on rel-
ative germination rate (Garcia-Huidobro et al. 1982a). Days required to achieve 5, 10, 15, 20, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 75, 80, 85, 90 and 95% germination were calculated for each species and treatment by interpolating between points on the cumulative-germination curves (Covell et al. 1986). It was assumed that the seeds in a given population germinate in the same relative order, regardless of thermal environment (Garcia-Huidobro et al. 1982a). Inverse days required to achieve a given germination percentile, therefore, was considered to equal the per-day germination rate of the subpopulation represented by the percentile ranking (Garcia-Huidobro et al. 1982a).

Germination rates for each species and subpopulation were plotted as a function of temperature (Fig. 1). The relationship between germination rate and temperature appeared to be relatively linear for most species and subpopulations in the suboptimal temperature range of 5 to 20°C. The exception was for Sandberg bluegrass which exhibited nonlinearity in the slower germinating subpopulations above 15°C. Linear equations were calculated from the suboptimal temperature data relating germination rate (day⁻¹) to temperature for all species and subpopulations. Suboptimal linear regressions were extrapolated to the x-axis to obtain a value for the base temperature at which subpopulation germination rate was estimated to be zero (Arnold 1959).

Germination rate could not be calculated for subpopulations that exhibited zero germination during the 21-day test period. Lack of data for many subpopulations made it difficult to develop a uniform approach to characterizing supraoptimal temperature response. Supraoptimal temperature response, however, was clearly nonlinear (Fig. 1). Thermal response above 20°C (15°C for Sandberg bluegrass) was characterized by a series of linear equations used to interpolate between constant-temperature germination rates in 5°C increments.

Estimation of Variable-Temperature Response
A hypothetical example illustrates the procedure used to estimate cumulative-germination response in the variable-temperature treatment regimes. If subpopulation “a” requires 10 days to germinate at 10°C then its per-day germination rate will be 0.1 day⁻¹ and it will progress 10% of the way toward germination for every day that it spends at 10°C. If the per-day germination rate at 10°C is 0.1 day⁻¹ then the germination rate for a 15-minute period is equal to the per-day rate divided by 96 (the number of 15-minute intervals in a day). Subpopulation “a” would, therefore, progress 1.042% toward germination during a 15-minute interval at 10°C. If one assumes that the instantaneous germination rate of a given subpopulation is independent of thermal history, then progress toward germination can be predicted by the running sum of germination rate estimates for each 15-minute interval. Germination of the subpopulation should occur when the...
running sum of these values reaches 1.

All environmental chambers were programmed to repeat the same temperature pattern every day. Average temperature over a given 21-day thermal treatment was calculated for each 15-minute interval of the diurnal temperature cycle. The germination rate for each 15-minute temperature interval was calculated for each species and subpopulation from the appropriate regression equations that were derived from the constant-temperature data. These 15-minute rate estimates were aggregated to produce a total daily (per-day, day⁻¹) germination rate for each combination of species, subpopulation, and variable-temperature thermal regime. The predicted time of germination (days) was determined to be the inverse of the aggregated per-day germination rate (Roundy and Biedenbender 1996). Some of the variable-temperature treatments overlapped temperature ranges for which constant-temperature germination was equal to zero. Germination times were not predicted for subpopulation-treatment combinations that were outside of the modeled-temperature range.

Results

Per-day germination rate across all subpopulations followed a general pattern. Germination rates were low at 5°C, increased with increasing temperature to a maximum in the 20–30°C range, and showed a relatively rapid decline at supraoptimal temperatures (Fig. 1). Peak germination rates tended to occur at relatively higher temperatures for the more rapidly germinating subpopulations. All subpopulations exhibited approximately linear, suboptimal temperature response over the temperature range of 5 to 20°C except for Sandberg bluegrass. Linear regression equations were derived for the data in this temperature range (5 to 15°C for Sandberg bluegrass) and base temperatures calculated by extrapolation to the temperature at which germination rate was estimated to be zero. Base temperature estimates tended to decrease with subpopulation but became highly variable for the most slowly germinating seeds (Fig. 2).

Suboptimal linearization of temperature response resulted in a general lack of model fit for the 5°C constant-temperature treatments (Fig. 3). We believe that this lack of fit was caused by non-linearity of germination rate response at low temperature. The suboptimal-temperature models were recalculated using the same piece-wise linear interpolation procedure that was used for the supraoptimal temperature range. This approach greatly improved model fit for temperature regimes that included time spent in the 5 to 10°C range.

![Fig. 2. Base temperature of thickspike wheatgrass [Expt. 1 (O), Expt. 2 (●)], bluebunch wheatgrass [Expt. 1 (□), Expt. 2 (■)], Sandberg bluegrass [Expt. 1 (△), Expt. 2 (▲)] and bottlebrush squirreltail [Expt. 1 (○), Expt. 2 (●)] as a function of subpopulation. Base temperatures were estimated by extrapolation of suboptimal germination rates (Fig. 1) to the temperature at which germination rate was estimated to be zero.](image_url)

![Fig. 3. Measured days to germinate (△), predicted days to germinate in the constant -5°C treatment for thickspike wheatgrass from the suboptimal-linear model (●), and predicted days to germinate form the piecewise-linear model (○). Error bars represent ±1 standard error.](image_url)
Some of the variable-temperature treatments exhibited reduced germination for some of the higher average-temperature treatments (Fig. 6).

**Discussion**

Models for calculating thermal response of seed germination have been developed for many agricultural species but there has been little model validation under variable-temperature conditions (Garcia-Huidobro et al. 1982b). Thermal response models have been used to compare germination response of several rangeland species but not for the specific purpose of predicting the time course of cumulative germination under variable temperature regimes (Jordan and

<table>
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<tr>
<th>Species</th>
<th>Temp</th>
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| Thickspike      | alt  | -0.4(0.05) | 0.1(0.02) | 0.2(0.04) | 0.1(0.04) | 0.5(0.14) | 0.5(0.20) | 0.3(0.22) | 0.3(0.10) | 0.6(0.23) | 0.4(0.24) | -0.2(0.08) | 0.5(0.17) | 0.4(0.21) | -
|                 | sin  | -0.3(0.03) | -0.3(0.04) | 0.1(0.03) | 0.0(0.02) | 0.3(0.10) | 0.4(0.24) | -0.2(0.08) | 0.5(0.17) | 0.4(0.21) | -
| Bluebunch       | alt  | -0.3(0.05) | 0.1(0.06) | 0.2(0.06) | 0.1(0.01) | 0.3(0.20) | 0.8(0.22) | 0.5(0.36) | 0.0(0.09) | 0.6(0.43) | -
|                 | sin  | -0.3(0.06) | -0.3(0.05) | 0.0(0.03) | -0.1(0.02) | 0.3(0.10) | 0.4(0.16) | -0.4(0.05) | 0.2(0.17) | -0.2(0.29) | -
| Sandberg        | alt  | -0.1(0.05) | 0.4(0.02) | 0.4(0.08) | 0.3(0.03) | - | -1.7(0.71) | 1.5(0.46) | 1.2(0.22) | -
|                 | sin  | 0.1(0.09) | 0.0(0.03) | 0.4(0.04) | 0.5(0.04) | - | 0.9(0.19) | 0.6(0.23) | 1.5(0.28) | -
| Squirreltail    | alt  | -0.1(0.06) | -0.4(0.04) | 0.0(0.02) | 0.0(0.02) | - | 1.5(0.72) | 0.6(0.53) | 0.7(0.27) | 0.1(0.03) | -
|                 | sin  | -0.2(0.04) | -0.1(0.01) | 0.1(0.01) | 0.1(0.03) | - | 0.9(0.46) | -0.2(0.09) | 0.2(0.08) | 0.7(0.10) | -

**Table 2.** Mean residual (predicted - measured) values of germination time (days) for Experiment 2 as a function of species, average temperature and temperature amplitude. Residual values were averaged separately for subpopulations ≤ 50% and subpopulations ≥ 50%. Positive values indicate that measured germination was faster than predicted. Values in parentheses represent 1 standard error of the mean. An asterisk indicates a single subpopulation reporting.

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<th>Species</th>
<th>Temp</th>
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<td>(5-50% Subpopulations)</td>
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<td>Thickspike</td>
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<td>Sandberg</td>
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<td>Squirreltail</td>
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Haferkamp 1989, Romo and Eddleman 1995, Roundy and Biedenbender 1996). Wester’s (1991) survey of the range literature on seed germination documents a previous emphasis on the calculation of germination indices for treatment comparisons rather than predictive inference. Young and Evans (1982) summarized an extensive body of their previous work measuring germination response of range grass species to alternating-temperature regimes but their data emphasizes total germination percentage rather than germination rate. Thermal response models of the type described in the current study could be used to expand our level of inference to predict differences in cumulative germination under an unlimited number of potential field temperature scenarios (Garcia-Huidobro et al. 1982a).

The germination-rate response to temperature found in our study conforms to the general pattern found in previous experiments (Garcia-Huidobro 1982a, Covell et al. 1986, Benech Arnold et al. 1990). Subpopulation-germination rates tended to increase in a relatively linear fashion up to an optimal temperature, above which they declined rapidly (Fig. 1). Our data are also similar to those shown in previous studies in that optimal temperatures for maximum germination rate tended to decrease for later-germinating subpopulations (Covell et al. 1986, Garcia-Huidobro et al. 1982a).

Previous authors have accepted several assumptions about thermal-response data in order to simplify analysis and to facilitate comparisons among seedlots. Common assumptions include the following: subpopulations within a given seedlot share the same base temperature; germination rate response is linear over the suboptimal temperature range; germination rate response is linear over the supraoptimal temperature range; instantaneous germination rate is independent of temperature pattern; instantaneous germination rate is independent of thermal history; and subpopulations maintain their relative germination rate regardless of the thermal environment to which they are subjected (Garcia-Huidobro et al. 1982a, Covell et al. 1986, Benech Arnold et al. 1990, and Probert 1992).

We found significant differences in base temperature estimates as a function of subpopulation (Fig. 2). Previous studies have measured the same phenomenon (Garcia-Huidobro et al. 1982a) but most conclude that the assumption of a common base temperature is still useful (Covell et al. 1986, Carberry and Campbell 1989, Benech Arnold et al. 1990). A common base temperature makes it possible to develop a single equation for predicting the thermal-time requirements of all subpopulations in the suboptimal temperature range (Washitani 1985, Covell et al. 1986, Ellis et al. 1987, Benech Arnold et al. 1990). Equation coefficients can then be used as an objective tool for comparing seedlots (Covell et al. 1986, Ellis et al. 1986). If the base temperature assumption is false, however, thermal accumulation for a given temperature regime must be calculated separately for each subpopulation. Temperatures near the base temperature have a relatively large influence on germination times as the temperatures above the base are inversely

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Fig. 4. Residual values of predicted minus measured days required to germinate as a function of subpopulation. Symbols for Experiment 1 represent alternating-temperature regimes with an average temperature of 10 (○), 15(●), 20 (△), 25 (▼), and 30˚C (♦); and sine-wave temperatures with an average temperature of 10(●), 15(●), 20 (△), 25(▼) and 30˚C (●). Symbols for Experiment 2 represent sine-wave regimes with an average temperature of 15˚C and a temperature amplitude of 5 (○), 10 (●), 15 (△), 20 (♦); and an average temperature of 20˚C and a temperature amplitude of 5(●), 10 (●), 15 (△) and 20 (●)˚C; and an average temperature of 25˚C and a temperature amplitude of 5 (●), 10 (●), 15 (●) and 20 (●)˚C.
related to germination rate (Bierhuizen and Wagenvroot 1974). We chose not to accept the common-base-temperature assumption because model fit was more important than computational efficiency.

Many previous models assume linearity of germination rate at both sub- and supraoptimal temperatures. Acceptance of this assumption simplifies the experimental design for obtaining model data (Covell et al. 1986, Ellis et al. 1986, 1987). We did not have enough data points in the supraoptimal temperature range to test linear thermal-time models but found that a linear model for the suboptimal range tended to underestimate germination rate in some low-temperature treatments (Fig. 3). Similar results in previous studies have been attributed to errors in temperature control, and to the relative sensitivity of germination rate to temperatures near the base temperature (Covell et al. 1986, Carberry and Campbell 1989). We believe that our model-prediction errors were caused by nonlinearity in germination response at low temperature rather than random or systematic errors in the experimental design. These errors disappeared when we replaced the suboptimal linear equations over the 5–20°C range with separate equations for each 5°C temperature increment (Fig. 3).

A common assumption of thermal-germination models is that the instantaneous germination rate is independent of temperature pattern (Garcia-Huidobro et al. 1982a, Probert 1992). This assumption has been shown to be invalid for species that remain dormant in a constant temperature environment (Thompson 1974, Thompson and Grime 1983, Murdoch et al. 1989). Models derived from constant-temperature data tend to underestimate variable-temperature germination response for these species (Garcia-Huidobro et al. 1982b, Benech Arnold...
Another common assumption of germination-response models is that germination rate is independent of thermal history (Benech Arnol et al. 1990, Garcia-Huidobro et al. 1982a). This assumption is counter-intuitive and may only apply to the suboptimal temperature range. Germination shares a common temperature response with many other types of physiological processes. Suboptimal temperature response is generally attributed to the increased thermal activity of the molecules involved in metabolic reactions (Monteith 1977, Garcia-Huidobro et al. 1982a, 1982b, Probert 1992). Supraoptimal temperature response, however, is generally attributed to molecular dysfunction caused by membrane degradation, denaturation of enzymes and other, relatively irreversible, degenerative processes (Bewley and Black 1994). One would, therefore, not expect seeds to regain germinability after significant time spent at supraoptimal temperature (Garcia-Huidobro et al. 1982b, Ojeda and Trione 1990). This hypothesis is supported by our observation that ungerminated seeds in the low-temperature treatments would quickly germinate at the end of the experiment when germination vials were returned to room temperature. Ungerminated seeds in the high temperature treatments, however, did not germinate when temperatures were subsequently lowered. If the thermal-history assumption is false, one would expect total germination percentage to be lower than predicted for treatments that overlap both the suboptimal and supraoptimal temperature ranges. Our data show this phenomenon for blue-bunch wheatgrass and Sandberg bluegrass for some of the higher-temperature treatments (Fig. 6). When this occurred, the effect was greater at alternating-constant temperatures than in sine-wave regimes that had equivalent mean daily temperature. This would be expected as the seeds in alternating-constant temperature regimes spend relatively more time at the maximum treatment temperature. The decline in germination rates at supraoptimal temperature was very rapid in our study and may not have been adequately characterized by a 5°C temperature increment. Future studies using a smaller constant-temperature interval and inclusion of some treatments with a greater diurnal amplitude could be used to more rigorously test this assumption.

An additional assumption of thermal-response models is that the relative ranking of germination rate among subpopulations remains the same regardless of the thermal regime (Garcia-Huidobro et al. 1982a, Benech Arnold et al. 1990). There is no reason to expect that subpopulations adapted to germinate rapidly at low temperatures are the same subpopulations that are adapted to germinate rapidly at high temperatures. Violation of this assumption would also cause an underestimation of cumulative-germination rate in regimes that overlapped both high and low temperature ranges. This response was not apparent from our data but our experimental design did not allow for rigorous testing of the assumption.

Conclusions

Rangeland systems exhibit high spatial and temporal variability of seedbed microclimate (Pierson and Wight 1991). Field-planting trials integrate complex combinations of environmental variables, but statistical inference is limited by the unique set of conditions present during a given test (Hurlbert 1984). Laboratory experiments are easier to control and replicate but inferences are limited by the artificiality of the treatment environment. Previous studies have attempted to bridge the gap between the laboratory and field approaches by testing seeds under a large number of alternating-temperature scenarios (Young and Evans 1982). Our study showed that for the seedlots tested in this experiment, cumulative germination under a large number of variable-temperature regimes could be predicted from data generated by relatively few, constant-temperature experimental treatments. Future studies might, therefore, adopt a simplified experimental design for seedlots that do not have a dormancy response to fluctuating temperature regimes. Data from some previous constant-temperature experiments could perhaps be reevaluated and the data used to predict population-germination response to potential field-variable temperature regimes.

The variable-temperature treatments tested in this experiment differ from true field regimes in that diurnal temperature patterns were constant. Actual field regimes would vary in diurnal temperature range, temperature pattern, and longer-term trends in mean-daily temperature (Hardegree and Burgess 1995). Our experiment also ignored the effects of other abiotic and biotic seedbed variables (Hegarty 1973). Soil water availability is often the limiting factor for seedling establishment in the Intermountain West (Roundy and Call 1988). Hydrothermal modeling approaches have been developed for predicting the interactive effects of temperature and water stress on seed germination (Gummerson 1986, Bradford 1990) and these models have recently been applied to the prediction of dormancy release in the winter-annual weed Bromus tectorum L. (Christensen et al. 1996). Thermal and hydrothermal-response models, however, require additional validation under field-variable temperature and moisture conditions.


