Overwintering of Wheat Stripe Rust Under Field Conditions in the Northwestern Regions of China

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Abstract

Stripe rust, caused by *Puccinia striiformis* f. sp. *tritici* Eriks., is an important disease of wheat worldwide. Identification of pathogen survival hot spots is important for predicting disease onset and subsequent spread to other regions, and such understanding is essential for developing integrated management strategies. We conducted field studies to determine the relationship of *P. striiformis* f. sp. *tritici* overwintering potential to winter temperatures in 10 sites in the northwest of China in three growing seasons (2011–12, 2012–13, and 2013–14). The fungus survived better in wheat cultivars with stronger winter hardiness than in those with weaker winter hardiness. Windowpane analysis suggested that *P. striiformis* f. sp. *tritici* winter survival is related to temperatures in the coldest period from mid-December to late January. Regression showed that *P. striiformis* f. sp. *tritici* winter survival was negatively related to the number of days with daily average temperature <−2°C for cultivars with weak winter hardiness and <−4°C for cultivars with both moderate and strong winter hardiness. Models developed under constant temperature experiments greatly underestimate the *P. striiformis* f. sp. *tritici* overwinter potential under field conditions. This underestimation probably results from the possibility that *P. striiformis* f. sp. *tritici* in wheat tissues may have increased tolerance to low temperatures because of acclimation to gradual temperature changes under field conditions. The present models can be used to predict quantitative risks of *P. striiformis* f. sp. *tritici* overwintering in winter-wheat-growing regions for timely implementation of regional disease management strategies.

Stripe rust, caused by *Puccinia striiformis* f. sp. *tritici* Eriks., is an important disease of wheat worldwide (Chen 2005). Stripe rust is a polycyclic disease and *P. striiformis* f. sp. *tritici* produces urediniospores in multiple cycles. Airborne urediniospores can be dispersed over a long distance and may cause local, regional, and interregional epidemics (Wang et al. 2010). Annual cycles of stripe rust on winter wheat can be divided into four stages: oversummering, infection of seedlings in autumn, overwintering, and spring epidemics. *P. striiformis* f. sp. *tritici* has to overwinter in living host tissues (Li and Zeng 2002; Sharp and Hehn 1963) and its survival over the winter is a key determinant of spring epidemics that directly result in crop losses (Gladders et al. 2007; Li and Zeng 2002; Sharma-Poudyal and Chen 2011). In cold high-altitude regions, *P. striiformis* f. sp. *tritici* is likely to overwinter in young green leaves as latent infection established in late autumn (Ma et al. 2016). *P. striiformis* f. sp. *tritici* mycelia inside green leaves may also be killed by low temperatures rather than through death of green leaves (Ma et al. 2015).

The ability to predict pathogen survival potential may enable timely implementation of integrated disease management strategies to reduce inoculum sources and prevent the formation of disease foci. Therefore, pathogen oversummering or overwintering areas have large epidemic consequences on local, regional, and interregional scales for disease prediction and management. For instance, summer and winter survival regions for *P. striiformis* f. sp. *tritici* have been used to devise stripe rust management strategies in China. Year-round survival areas such as southern Gansu and northern Sichuan are considered stripe rust hot spots because these regions were associated with evolution of new races and high pathogenic variability (Li and Zeng 2002).

Temperature is the main limiting factor for *P. striiformis* f. sp. *tritici* survival in summer and winter because the pathogen cannot survive extreme cold or hot conditions (Sharma-Poudyal et al. 2014). Warm winters favor *P. striiformis* f. sp. *tritici* survival in infected leaves whereas severe winters do not allow the pathogen’s survival (Grabow et al. 2016). The low limit of air temperature for *P. striiformis* f. sp. *tritici* to overwinter (November to February in China and North America) is believed to be around −6 to −7°C (monthly mean) (Sharma-Poudyal et al. 2014; Shen and Wang 1962; Zeng and Luo 2006). Potential *P. striiformis* f. sp. *tritici* overwintering areas in China have been identified (Ma et al. 2005; Shen and Wang 1962) based on the data from field observations. However, recent field observations suggest that the overwintering boundary may have been gradually pushed toward the northwest (Pan et al. 2011; Yao et al. 2014), which may partially result from climate changes in recent times (Fang et al. 2012). In addition to low temperatures, wind chill may also result in reduced *P. striiformis* f. sp. *tritici* overwintering potential (Ma et al. 2016). Many models have been developed to describe the relationship between winter temperature and stripe rust epidemics based on field observations (Coakley and Line 1981; Coakley et al. 1982, 1988) and were further revised (Sharma-Poudyal and Chen 2011). A model based on winter climatic conditions has been developed to predict *P. striiformis* f. sp. *tritici* overwintering potential on a categorical scale for winter-wheat-growing regions in the United States (Sharma-Poudyal et al. 2014). The negative effect of low temperature on *P. striiformis* f. sp. *tritici* overwintering may be alleviated to a certain extent by snow: a minimum snow cover of 7.6 cm may be sufficient to protect winter wheat from winter kill. The importance of environment during the winter months on *P. striiformis* f. sp. *tritici* survival was also noted in Europe (Papastamati and van den Bosch 2007; Te Beest et al. 2008).

We recently established a quantitative relationship in *P. striiformis* f. sp. *tritici* overwintering potential to temperature, based on the data obtained from constant and fluctuating temperature regimes under controlled conditions (Ma et al. 2015, 2017). *P. striiformis* f. sp. *tritici* survival in cultivars with strong winter hardiness at 0°C and −5°C was greater than in those with weak winter hardiness. However, such

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differences in *P. striiformis* f. sp. *tritici* survival among cultivars were negligible when temperatures were ≤−10°C. Logistic models satisfactorily described the relationship of *P. striiformis* f. sp. *tritici* survival in attached leaves to temperature and winter hardiness. The relationship of *P. striiformis* f. sp. *tritici* survival to temperature is nonlinear, particularly in the range of −5 to −10°C (Ma et al. 2015), which was further confirmed by controlled fluctuating temperature studies (Ma et al. 2017). In parallel to these controlled studies, we monitored *P. striiformis* f. sp. *tritici* overwintering on cultivars with different winter hardiness as well as temperature during the winter in three seasons (2010–11, 2011–12, and 2012–13). This study identified the relationship of *P. striiformis* f. sp. *tritici* overwintering to winter temperature in winter wheat under field conditions.

**Materials and Methods**

**Sites and wheat varieties.** Field experiments were conducted in three seasons—2011–12, 2012–13, and 2013–14—at seven, eight, and eight sites (fields), respectively, in Gansu and Qinghai Provinces, China (Fig. 1; Table 1). At each site, a USB logger (Lascar EL-USB-2+; Martel Electronics Corporation) was used to record hourly temperature and humidity from 1 December to 10 March of the next year. The logger was placed at 5 cm above the ground. Several loggers failed to log the data for the complete period every year. For these sites, daily average temperature values recorded by a local meteorological station were used instead. For 8 of the 23 site–year combinations, complete data were recorded by the USB loggers. For all other sites, we had to use data recorded by local weather stations.

In the 2011–12 season, *P. striiformis* f. sp. *tritici* was only assessed in commercial winter-wheat fields whereas, in the 2012–13 and 2013–14 seasons, *P. striiformis* f. sp. *tritici* was monitored in experimental plots at each site where wheat plants were inoculated with *P. striiformis* f. sp. *tritici* in the autumn (Table 1). The experimental design was the same as that previously reported (Ma et al. 2016). At each site, four varieties (Table 1) were sown within a single plot (10 by 20 m), with each variety occupying 50 m² (2.5 by 20 m). Two common varieties were used in all years at all sites: Xiaoyan 22 (XY22, weak winter hardiness) and Xinong 979 (XN979, moderate winter hardiness). The other two varieties were from the local regions and, hence, could differ between sites. At each site, two strips (0.15 by 20 m) of cultivar Mingxian 169 (MX169, weak winter hardiness, highly susceptible to stripe rust) were planted. MX169 plants were inoculated in early autumn to act as an inoculum source for other varieties to initiate autumn infection.
Field inoculation of MX169 by *P. striiformis* f. sp. *tritici*. *P. striiformis* f. sp. *tritici* inoculum was multiplied in a controlled environment on MX169 seedlings in pots (10 by 12 cm) containing cow dung compost and soil (1:2, vol/vol). When the first two leaves were fully expanded (8 to 10 days after sowing), seedlings were inoculated with fresh urediospores of the strain Chinese Yellow Rust (CYR) 32 (Li and Shang 1989), immediately placed in a dew chamber at 8 to 10°C in dark for 24 h, then moved out to a greenhouse compartment (10 h/day at 17°C and 14 h/night 14°C). Fresh urediospores were collected from these inoculated leaves and stored in a desiccator at 4°C.

In mid-October, MX169 plants of two strips at each experimental plot were inoculated with the collected fresh urediospores of *P. striiformis* f. sp. *tritici* CYR32. Plants were first sprayed with water to wet the leaves. Then, urediospores were mixed with talcum powder (1:20, vol/vol) in a glass testing tube (0.5 by 12 cm, 15 ml) that was covered with three layers of cotton gauze, and scattered over plants through gently tapping the testing tube (Li and Shang 1989). Inoculated seedlings were immediately covered with black polythene to maintain high humidity conducive for infection. The cover was removed after 24 h.

*P. striiformis* f. sp. *tritici* assessment in the autumn and spring. From late November to early December 2011, five points were randomly selected for each cultivar, and 40 plants at each point were then assessed for the presence of actively sporulating *P. striiformis* f. sp. *tritici* lesions. In the spring (late March), a number of plants (10 to 60) of each cultivar (Table 1) were randomly selected for assessment of actively sporulating *P. striiformis* f. sp. *tritici* lesions. Any sporulating lesion was unlikely to have resulted from new infections in the early spring and, hence, most likely originated from infections in the autumn.

In autumn 2012 and 2013, randomly selected plants with sporulating *P. striiformis* f. sp. *tritici* lesions at each site were labeled (Table 1). Immediately after *P. striiformis* f. sp. *tritici* assessment in the following spring, we dug taggged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring, we dug tagged plants, took them back to the laboratory (Northwest A&F University), transplanted the plants in the following spring.

**Data analysis.** There were two measures for the proportion of plants with overwintered *P. striiformis* f. sp. *tritici* (POw): OW1 = PSy/PSx, where PSx and PSy were the *P. striiformis* f. sp. *tritici* incidences in the autumn and spring, respectively; and OW2 = PSy/TPS, where PSy was the number of tagged plants with sporulating *P. striiformis* f. sp. *tritici* when assessed in the glasshouse after transplanting, and TPS was the total number of tagged plants transplanted in the spring.

For the 2011–12 season, POw was estimated as OW1; for the 2012–13 and 2013–14 seasons, POw was estimated as the average of OW1 and OW2.

We initially used windowpane analysis (Kris et al. 2010; Pietravalle et al. 2003) first to determine the strength of correlation of POw with temperature-derived variables and then to select appropriate time periods over which temperature was to be summarized for β regression analysis.

**Windowpane analysis.** We constructed windowpane data from 1 December to 10 March of the following year to approximate the overwinter period. Summary temperature variables were calculated for windows of lengths 15, 30, 45, and 60 days. Daily average temperature in each window was summarized into the following nine variables:

(i) Average temperature (Mean_T, °C)
(ii) Number of days with daily average temperature ≥0°C (DGT0)
(iii) Degree days >0°C (MGDT0)
(iv) Number of days with daily average temperature <0°C (DLT0)
(v) Degree days <0°C (MLT0)
(vi) Number of days with daily average temperature ≤-2°C (DLT2)
(vii) Degree days ≤-2°C (MLT2)
(viii) Number of days with daily average temperature ≤-4°C (DLT4)
(ix) Degree days ≤-4°C (MLT4)

Temperature ≤-4°C was not further subdivided because there were not many days with daily average temperature ≤-4°C.

The association between POw and weather variables was quantified with the usual parametric correlation (r) for each category of cultivar winter hardiness, pooling data across the 3 years. Thus, each correlation was determined from 23 data points. There is no simple solution to the fact that the test statistics are highly correlated for assessing the significance of windowpane correlation coefficients. We performed a global test of significance for a given windowpane width using the Simes method (Simes 1986), testing for the global null hypothesis that none of the individual correlations is significant. Because the windowpane analysis was only used as a first step to select a time period for further regression analysis, we used two critical values (α = 0.005 and 0.001) purely for visual examination of the correlation strength.

**Regression analysis.** Because POw was in the range of 0 to 1, we used the β regression model to establish the relationship of POw to temperature-derived variables via a logit link function. The exact temperature variables (windowpane width and starting date) were determined by the initial windowpane analysis. For a given windowpane width and starting date, there were only nine temperature-derived variables. Thus, we used a backward selection method to

### Table 1. Study sites and winter wheat cultivars for wheat stripe rust assessment in autumn and the following spring in the northwestern region of China in three growing seasons (2011–12, 2012–13, and 2013–14)

<table>
<thead>
<tr>
<th>Site</th>
<th>Longitude</th>
<th>Latitude</th>
<th>Elev (m)</th>
<th>2011</th>
<th>2012</th>
<th>2013</th>
<th>Data</th>
<th>Plants assessed in the spring (2011–12)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Qinzhou A</td>
<td>105°42’44’’</td>
<td>34°23’77’’</td>
<td>1,725</td>
<td>MX169; XY22;</td>
<td>XX979</td>
<td>XX22; XX979</td>
<td>1–314; LT91474</td>
<td>2011 10, 20, 20 60</td>
</tr>
<tr>
<td>Qinzhou B</td>
<td>105°42’41’’</td>
<td>34°24’29’’</td>
<td>1,627</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2013 10, 20, 20 60</td>
</tr>
<tr>
<td>Qinzhou C</td>
<td>105°42’52’’</td>
<td>34°24’34’’</td>
<td>1,587</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2011 10, 20, 20 60</td>
</tr>
<tr>
<td>Lixian A</td>
<td>105°16’55’’</td>
<td>34°12’79’’</td>
<td>1,532</td>
<td>FS185; two other local cultivars</td>
<td></td>
<td></td>
<td></td>
<td>2011 20, 20, 60</td>
</tr>
<tr>
<td>Lixian B</td>
<td>105°16’60’’</td>
<td>34°12’83’’</td>
<td>1,538</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2013 20, 20 60</td>
</tr>
<tr>
<td>Zhuanglang A</td>
<td>106°06’84’’</td>
<td>35°12’57’’</td>
<td>1,666</td>
<td>TX46; two other local cultivars</td>
<td></td>
<td></td>
<td></td>
<td>2013 50 60</td>
</tr>
<tr>
<td>Zhuanglang B</td>
<td>106°00’47’’</td>
<td>35°09’73’’</td>
<td>1,543</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2012 50 60</td>
</tr>
<tr>
<td>Guide</td>
<td>101°25’14’’</td>
<td>36°01’83’’</td>
<td>2,226</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2013 50</td>
</tr>
<tr>
<td>Jianzha</td>
<td>101°55’77’’</td>
<td>36°03’16’’</td>
<td>2,099</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2013 60</td>
</tr>
<tr>
<td>Xunhua</td>
<td>102°22’31’’</td>
<td>35°52’56’’</td>
<td>1,877</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>2013 60</td>
</tr>
</tbody>
</table>

*a Cultivars FS185, LT91474, TX46, LT26, LT15, and JN411 = strong winter hardiness; MX169 and XY22 = weak winter hardiness; and XX979, 01-314, and 06-11 = moderate winter hardiness. At the Lixian and Zhuanglang sites in 2011, one of the two local cultivars assessed was of weak winter hardiness and the other of moderate winter hardiness.

*b Only 10, 20, and 20 plants were assessed in the spring for MX169, XY22, and XX979, respectively.

*elevation in meters above sea level.

*US temperature data.
select a “best” model; initially, by fitting a model with all nine variables (or maximum number of possible variables permitted with model fitting due to collinearity), then gradually reducing the number of variables until the best model was selected based on the percentage of variance explained, the pattern of residual errors, and collinearity among model variables. Because of large differences in the association of Pow with temperature (revealed by the windowpane analysis) among cultivars of different cold hardness, β regression was performed separately for each category of winter hardiness. The growing season (year) was used as a factor in the β regression. All data analyses were carried out in R (version 3.5.0) and Zelig ies (2010).

Results

General observations. Analysis suggested that daily average temperatures recorded by the USB loggers are highly correlated with those recorded by the local stations: correlation coefficient is between 0.75 and 0.80. Climatic conditions differed considerably between years. Overall, the 2013–14 winter was the coldest among the three seasons (Fig. 1). There were many days with daily average temperature <−5°C for many sites in the 2013–14 winter; in contrast, there were only a few days with daily average temperature <−5°C for a few sites in the 2011–12 and 2012–13 seasons. For all sites, the coldest period was from the late December to late January (Fig. 1). The extent of daily temperature fluctuation differed considerably among sites. For instance, at Qinzhou A in 2011–12, the minimum and maximum of the daily amplitude was 0.5 and 5.5°C, respectively, with an average of 2.2°C. In contrast, at Qinzhou B in the same season, the corresponding values were 1 and 14.5°C (average 7.9°C).

Average Pow was similar between cultivars with different winter hardness: 20, 21, and 18% for cultivars with weak, moderate, and strong winter hardness, respectively (Fig. 2). However, the variability in Pow was the greatest and least for the cultivars with weak and strong winter hardness, respectively—the corresponding variances were 0.032, 0.020, and 0.012. Pow was higher in the 2011–12 (33%) than in the 2012–13 (13%) and 2013–14 (15%).

Windowpane analysis. Pow was highly correlated with all the nine temperature-derived variables, particularly for cultivars of weak winter hardness, but not for cultivars of strong winter hardness (Fig. 3). Overall, the correlation was stronger for the period from mid-December to early February. Of the nine variables, the two related to temperature of approximately 0°C were least correlated with Pow. The Simes test of windowpane correlation coefficients confirmed these findings (Fig. 4); none of the nine variables was significant for cultivars with strong winter hardness.

β Regression analysis. β Regression was performed separately for each category of winter hardness. Based on the windowpane results, temperature was summarized for the period from 20 December to 18 January for cultivars with weak winter hardness, and from 30 December to 28 January for cultivars with moderate and strong winter hardness. Regardless of winter hardness, the relationship of Pow with temperature-derived variables varied significantly with the season. Specifically, this relationship was similar between the 2012–13 and 2013–14 seasons but differed significantly from the 2011–12 season.

The summary of β regression results is provided in Table 2. For cultivars with both moderate and strong winter hardness, Pow was not significantly related to any temperature-derived variable in the 2011–12 season but was negatively related to the duration when daily average temperature was <−4°C for the other two seasons. For the cultivars with weak winter hardness, Pow was significantly reduced by the increased duration when daily average temperature was <−2°C for all three seasons; however, the effect was much greater for the 2012–13 and 2013–14 seasons than for 2011–12 (Table 2). The model performance was the best and worst for the cultivars with weak and strong winter hardness, respectively, based on the correlation between fitted and observed values (Table 2).

Interpretation of these models can be made on the basis of the odds-ratio because the logit was used as the link function in the β regression. For instance, for the cultivars with weak winter hardness, the odds ratio of successful P. striiformis f. sp. tritici overwintering is reduced by approximately 9% with one additional day with daily average temperature <−2°C in the 2012–13 or 2013–14 season, and by 4% in the 2011–12 season. The relative decline of Pow is shown in Figure 5 in relation to the number days with average temperature <−2°C in the period from 20 December to 18 January for cultivars with weak winter hardness, and <−4°C in the period from 30 December to 28 January for cultivars with both moderate and strong winter hardness. P. striiformis f. sp. tritici survived the most and the least on the cultivars with strong and weak winter hardness, respectively (Fig. 5).

Discussion

Understanding bottlenecks in pathogen survival is important for predicting disease onset and subsequent spread to other regions, and for developing and implementing integrated management strategies. Managing wheat stripe rust in oversummering and overwintering regions is a key component of sustainable stripe rust management strategies in China (Li and Zeng 2002). Based on field data, P. striiformis f. sp. tritici overwintering models were developed and these models can predict quantitative risks of P. striiformis f. sp. tritici overwintering at specific sites.

The present study agrees with previous controlled studies showing that increasing cultivar winter hardness leads to increasing survival of P. striiformis f. sp. tritici under low temperatures (Ma et al. 2015, 2017). This could be related to the reduced survival of leaves of cultivars with weak winter hardness under low temperatures. However, P. striiformis f. sp. tritici mycelia inside leaves could be killed before leaves were killed under constant temperatures of −10 or −20°C (Ma et al. 2015). Previous controlled studies showed that differences in P. striiformis f. sp. tritici overwintering between cultivars of different winter hardness occurred mostly in the temperature range of 0 to −10°C (Ma et al. 2015); most recorded temperatures during the winter in the present study sites were within this range.

P. striiformis f. sp. tritici survival under field conditions was much greater than that obtained under constant-temperature experiments. For instance, after 48 h at constant temperature of −5°C, only approximately 10, 15, and 20% of P. striiformis f. sp. tritici survived on
cultivars with weak, moderate, and strong winter hardiness, respectively (Ma et al. 2015). In the present study, the fitted models predicted that approximately 10 and 40% of *P. striiformis* f. sp. *tritici* may survive for cultivars with moderate and strong winter hardiness, respectively, after 30 days with daily average temperature < −4°C. Most of this difference cannot be explained by the nonlinearity due to using daily average temperatures to predict *P. striiformis* f. sp. *tritici* survival. After 48 h under a fluctuating constant temperature of average −5°C with a range of 8°C, *P. striiformis* f. sp. *tritici* survival ranged from 25 to 30% for cultivars with both weak and strong winter hardiness (Ma et al. 2017). Furthermore, the nonlinear effect is expected to be an issue only mostly for sites in the 2013–14 season, where daily average temperature was in the range of −4 to −10°C on many occasions. Several other factors may partially account for this difference. First, results from controlled constant and fluctuating temperature studies may considerably underestimate *P. striiformis* f. sp. *tritici* survival in field conditions. Under the controlled studies (Ma et al. 2015, 2017), plants were subjected to sudden changes in temperature compared with field conditions. Thus, under field conditions, *P. striiformis* f. sp. *tritici* may be able to tolerate cold temperatures better than predicted by models developed from controlled studies. *P. striiformis* f. sp. *tritici* may acclimatize to low temperatures when the gradient in temperature changes is small. Second, at some sites, winter snow cover may insulate the winter wheat and the pathogen from low temperatures (Sharma-Poudyal et al. 2014). Snow cover is generally not known to last for a long period of time.

![Figure 3](image-url)

**Fig. 3.** Pearson correlation coefficients at four window widths to assess the association between nine temperature-derived variables and the proportion of stripe rust overwintered in northwestern China in three seasons (2011–12, 2012–13, and 2013–14). Black line = cultivars of weak winter hardiness, red line = cultivars of moderate winter hardiness, and blue line = cultivars of strong winter hardiness; blue and red points = *P* value of correlation < 0.005 and 0.001, respectively.

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in the sampling regions, particularly in Gansu Province. Finally, *P. striiformis* f. sp. *tritici* may infect leaves even \(< -4^\circ C\) but remain symptomless in the winter (Burleigh and Hendrix 1970). Visual disease assessment in the autumn would not include these latent infections and, thus, may considerably underestimate the incidence of *P. striiformis* f. sp. *tritici* infections, leading to the overestimation of *P. striiformis* f. sp. *tritici* winter survival under field conditions.

The relationship between *P. striiformis* f. sp. *tritici* overwintering potential and temperature also differed between 2011–12 and the other two seasons: the survival in 2011–12 winter was either not dependent on temperature (for cultivars with moderate and strong winter hardiness) or much higher than in the other two seasons (for cultivars with weak winter hardiness). This is probably due to higher winter temperature at those sites in winter 2011–12 than in the other two seasons, particularly 2013–14. Only a few days were recorded with daily average temperature \(< -4^\circ C\) at a few sites in winter 2011–12. It is thus not surprising that we failed to define a clear relationship between *P. striiformis* f. sp. *tritici* winter survival and temperature for cultivars with moderate and strong winter hardiness. Therefore, the overall *P. striiformis* f. sp. *tritici* winter survival was much greater in the 2011–12 season than in the other two seasons. Differences in sampling methodologies should not have resulted in the observed differences between seasons. Using tagged plants to estimate overwinter potential is expected to give more accurate results; on average, the estimated *P. striiformis* f. sp. *tritici* overwintering potential based on the tagged plants was slightly greater than that based the disease assessment in the autumn and spring (16 versus 11%). Thus, if we had used the tagged methods in the 2011–12 season as well, it might have led to even greater *P. striiformis* f. sp. *tritici* survival estimates. Fewer plants were sampled in 2011–12, which is expected to increase sampling errors but should not affect the magnitude of model parameter estimates. Considering all these points together, we would suggest that only the relationships derived from the 2012–13 and 2013–14 seasons should be taken forward.

The present models differ considerably from a previous model developed in the United States to predict *P. striiformis* f. sp. *tritici* overwintering potential in winter-wheat-growing regions (Sharma-Poudyal et al. 2014). First, the U.S. model was developed based on the synthesis of many previous findings on several aspects of *P. striiformis* f. sp. *tritici* disease cycles. In contrast, we only used field observations in the northwest region of China to construct empirical models without explicitly considering specific disease cycles. The U.S. model included more explanatory variables. Most importantly, the U.S. model only predicts overwintering potential as categorical scores. In contrast, the present models predict the overwintering potential quantitatively, on a continuous scale of 0 to 1. The models are easy to interpret based on the odds ratio. In addition, our models also consider the effects of cultivar winter hardiness on *P. striiformis* f. sp. *tritici* overwintering potential.

In summary, *P. striiformis* f. sp. *tritici* overwintering is primarily affected by the duration of low temperatures in the coldest period from late December to late January. The derived logistic models

![Fig. 4. Simes test for the overall significance of multiple nonindependent correlation coefficients from windowpane analysis. The horizontal line indicates P = 0.05.](image)

**Table 2.** Regression results for proportion of overwintered wheat stripe rust in the northwestern China in three seasons (2011–12, 2012–13, and 2013–14)*

<table>
<thead>
<tr>
<th>Cultivar winter hardiness</th>
<th>2011–12</th>
<th>2012–13 and 2013–14</th>
<th>Correlation (fitted versus observed)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Weak</td>
<td>-0.044  (±0.016) • DLT2</td>
<td>-0.095  (±0.011) • DLT2</td>
<td>0.814</td>
</tr>
<tr>
<td>Moderate</td>
<td>-0.750  (±0.246) + 0.018 (±0.061) • DLT4</td>
<td>-0.750  (±0.246) – 0.087 (±0.0198) • DLT4</td>
<td>0.775</td>
</tr>
<tr>
<td>Strong</td>
<td>-1.411  (±0.267) + 0.096 (±0.065) • DLT4</td>
<td>-1.411  (±0.267) – 0.039 (±0.020) • DLT4</td>
<td>0.647</td>
</tr>
</tbody>
</table>

*In β regression, logit was used as the link function. DLT2 and DLT4 = average number of days with average temperature less than –2 and –4°C, respectively, in the window width. For cultivars with weak winter hardiness, DLT2 was calculated for the period of 20 December to 18 January whereas, for cultivars with moderate and strong winter hardiness, DLT4 was calculated for the period of 30 December to 28 January. Numbers in parentheses indicate the standard error of the corresponding parameter estimate.*
fitted the observed data satisfactorily. *P. striiformis* f. sp. *tritici* survival in low temperatures under field conditions is much greater than predicted from controlled studies. *P. striiformis* f. sp. *tritici* overwintering may be affected by gradient changes in temperatures, which needs further research. The present models can be further improved by incorporating snow cover and possibly the wind chill effect (Ma et al. 2016). Further independent validation is also necessary before they can be used in practice with confidence.

**Literature Cited**


