Predicting Cereal Root Disease in Western Australia Using Soil DNA and Environmental Parameters

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ABSTRACT

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Root diseases have long been prevalent in Australian grain-growing regions, and most management decisions to reduce the risk of yield loss need to be implemented before the crop is sown. The levels of pathogens that cause the major root diseases can be measured using DNA-based services such as PreDicta B. Although these pathogens are often studied individually, in the field they often occur as mixed populations and their combined effect on crop production is likely to vary across diverse cropping environments. A 3-year survey was conducted covering most cropping regions in Western Australia, utilizing PreDicta B to determine soilborne pathogen levels and visual assessments to score root health and incidence of individual crop root diseases caused by the major root pathogens, including

Root diseases that limit small grain cereal yields (Triticum aestivum L., Hordeum vulgare L., and Avena sativa L.) are prevalent across Western Australia and have varying impact dependent upon agroecological zones, management practices, and seasonal environmental conditions ([2\)](#page-9-0). The advent of minimum-tillage practices increased the occurrence of root diseases and pathogens in Western Australia [\(2,25\)](#page-9-0). The main root disease pathogens are Rhizoctonia $solani$ J. G. Kühn anastomosis group (AG) 8 (common name Rhizoctonia root rot); Gaeumannomyces graminis var. tritici J. Walker (common name take-all); Pratylenchus neglectus(Rensch) Filipjev & Schuurmans Stekhoven, P. teres, and P. thornei Sher & Allen (root-lesion nematodes); Fusarium pseudograminearum O'Donnell & T. Aoki (teleomorph Gibberella coronicola T. Aoki & O'Donnell) and F. culmorum (Wm. G. Sm.) Sacc. (teleomorph unknown) (causal agents of crown rot); Heterodera avenae Wollenweber (cereal cyst nematodes); and Bipolaris sorokiniana (Sacc.) Shoemaker (syns. Helpminthosporium sativum Pammel, C. M. King & Bakke and H. sorokinianum Sacc.) (causal agent of common root rot) ([2](#page-9-0),[10\)](#page-9-0). These species occur across Australia and other parts of the world, including the Pacific Northwest (PNW) of the United States and Turkey, and historically have been studied individually ([2](#page-9-0),[15,17,31,33,35](#page-9-0),[43](#page-10-0),[51](#page-10-0)). However, in reality, these species occur as mixed populations in cereal cropping systems [\(15](#page-9-0),[26,36,37,](#page-9-0)[55\)](#page-10-0). In Western Australia, root diseases cause

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Rhizoctonia solani (anastomosis group [AG]-8), Gaeumannomyces graminis var. tritici (take-all), Fusarium pseudograminearum, and Pratylenchus spp. (root-lesion nematodes) on wheat roots for 115, 50, and 94 fields during 2010, 2011, and 2012, respectively. A predictive model was developed for root health utilizing autumn and summer rainfall and soil temperature parameters. The model showed that pathogen DNA explained 16, 5, and 2% of the variation in root health whereas environmental parameters explained 22, 11, and 1% of the variation in 2010, 2011, and 2012, respectively. Results showed that R. solani AG-8 soil pathogen DNA, environmental soil temperature, and rainfall parameters explained most of the variation in the root health. This research shows that interactions between environment and pathogen levels before seeding can be utilized in predictive models to improve assessment of risk from root diseases to assist growers to plan more profitable cropping programs.

Additional keywords: Bipolaris sorokiniana, common root rot, crown rot.

substantial root damage and subsequent yield loss in cereals, with annual losses estimated by Brennan and Murray ([30](#page-9-0),[31](#page-9-0)) for R. solani, F. pseudograminearum, and Gaeumannomyces graminis var. tritici to be \$27, 7, and 4 million per annum, respectively. Others have suggested substantial yield losses to wheat due to root-lesion nematodes P. neglectus and P. thornei in Australia and throughout the world [\(33,37,](#page-9-0)[53](#page-10-0),[58](#page-10-0)–[60](#page-10-0)).

The focus of surveys and ecological studies have been on these various individual root diseases in replicated plot trials conducted throughout the United States, Australia, Europe, and New Zealand ([3,4,6–9](#page-9-0),[11](#page-9-0),[12,14,20,22](#page-9-0),[23](#page-9-0),[32,34](#page-9-0)[,39,41,43](#page-10-0),[52](#page-10-0),[62](#page-10-0)). Plant pathologists generally accept the concept that disease is a result of the interactions between the pathogen, host, and environment [\(1\)](#page-9-0). In the case of take-all, environmental conditions can play a significant role in disease expression regardless of the inoculum levels ([8,20,35](#page-9-0)). Others have shown that crop rotation and tillage management practices have had significant effects on expression of Rhizoctonia bare patch [\(16](#page-9-0),[18](#page-9-0),[19](#page-9-0),[22](#page-9-0)[,41\)](#page-10-0). Okubara et al. [\(34\)](#page-9-0) reported significant interactions between environmental climatic factors of temperature and rainfall with R. solani AG-8 and R. oryzae pathogen DNA. R. solani AG-8 pathogen DNA concentration was negatively correlated with precipitation; that is, higher concentrations under lower rainfall conditions. Okubara also found lower concentrations of R. solani AG-8 pathogen DNA following a broadleaf crop. Several have reported the significant effect of no-tillage and crop rotation practices on the prevalence of these root diseases ([13](#page-9-0),[15](#page-9-0),[18,23,34,38,](#page-9-0)[40](#page-10-0),[41](#page-10-0),[45,50\)](#page-10-0).

The Western Australia cropping region is characterized by a Mediterranean environment, with strong winter dominant rainfall

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and hot dry summers. Stephens and Lyons [\(57\)](#page-10-0) regressed the rainfall from various weather stations throughout Western Australia with wheat yields to develop 16 agrometeorological zones. They reported that autumn rainfall and finishing rains after July were the most critical for wheat yield ([57\)](#page-10-0). DNA-based methods were developed in 1995 to estimate pathogen levels for G. graminis var. tritici and Rhizoctonia spp. ([27,35](#page-9-0)). Soil-based DNA-based methods have been developed to detect the major root pathogens to aid in surveys and ecological studies ([18,34](#page-9-0),[35,](#page-9-0)[49](#page-10-0)). There have been several attempts to model the relationships between environment and disease to predict severity of take-all, eyespot, and crown rot [\(4,13,21](#page-9-0),[39](#page-10-0)). Lawes and Renton ([24](#page-9-0)) used root diseases in a predictive model known as Land Use Sequence Optimizer (LUSO) that estimates the break crop effect based on variables including price, crop yield, nitrogen fertilizer cost, and weed load thresholds.

DNA-based technology has allowed for potentially more accurate assessment of soilborne pathogen levels that can be utilized in modeling cereal root disease development. Herdina and Roget ([21](#page-9-0)) developed a DNA-based soil assay to predict take-all disease risk. Schoeny et al. ([46](#page-10-0)) modeled take-all risk based on relating disease incidence and severity to winter wheat yield. Bithell et al. ([8\)](#page-9-0) showed that the relationship between presown G. graminis var. *tritici* soil DNA concentration and take-all disease development in crop was a logarithmic function dependent upon the environmental conditions during the growing season, and used this information to develop four disease risk categories.

Through the development of high-throughput DNA-based assays to estimate pathogen levels of a broad range of soilborne pathogens by the South Australian Research and Development Institute (SARDI) root disease testing service (RDTS), large-scale surveys can be conducted to investigate possible interactions between pathogen levels, environment and disease. SARDI's RDTS offers a range of soil-based DNA available commercially for growers and researchers for the estimation of cereal root disease risk assessment. Root disease incidence and severity is usually assessed between seedling stage and anthesis [\(20,33](#page-9-0),[39](#page-10-0),[41,43,44,46](#page-10-0),[51](#page-10-0),[53](#page-10-0),[56\)](#page-10-0). In the survey described here, root health was assessed at approximately anthesis to more closely reflect the final state of the health of the root system and the cumulative effect of root disease at the end of the season during grain fill. Research focused on this time frame because it was postulated that this would show the greatest effects of all the root diseases cumulatively. The objectives of this research were to model the relationship between soilborne pathogens levels assessed presowing using DNA-based assays, environment, and root health at anthesis utilizing data collected over 3 years from cropping fields in Western Australia.

MATERIALS AND METHODS

Sampling of field sites and root health disease assessments. The intention of this study was to focus on cereal root diseases for continuous wheat rotations; therefore, fields in wheat production for each year were selected. History of the previous crop was recorded to assess disease occurrence under different crop rotation conditions. Survey sites included 184 fields located across the major wheat-producing regions of Western Australia, from Albany in the south to Geraldton in the north. Of these locations, some fields that lacked viable data were eliminated and only 115, 50, and 94 were included in the analysis for the years 2010, 2011, and 2012, respectively. There were 21 fields that were consistently sampled across the 3 years covered in the study. Samples were taken at the seedling (June to July) and anthesis (September to October) phases of development (approximately 6 weeks after emergence for the seedling phase). Preliminary data analysis showed that plant-based disease assessments taken at anthesis were more accurate and descriptive of the wide range of cereal root diseases being assessed in this study. Although root diseases such as Rhizoctonia root rot and take-all are often assessed during the seedling stage, those of common root rot and Fusarium crown rot express themselves later in the season during the anthesis phase of development. For this reason, the authors felt that the anthesis timing would reveal a greater expression and damage assessment for the diverse biology involved with these cereal root diseases.

Forty plants were systematically sampled from a 1-ha area in each field. Each 1-ha area was divided into four replication areas with an area of 25 by 100 m for sampling with an Accucore probe (Adelaide, SA, Australia), with samples approximately 10 m apart. GPS coordinates were recorded for each field site using a Garmin Etrex 12 (Olathe, KS) channel device on World Geodetic System 1984 datum (WGS84). Roots and soil surrounding samples were sealed in a labeled plastic bag, foliage extending out of the bag. The entire sample, root, soil and foliage was placed in a paper bag and sent to Department of Agriculture and Food of Western Australia (DAFWA) in Perth, WA for visual assessment of disease. The number of plants (incidence) with symptoms of the major root diseases, including Rhizoctonia root rot, take-all, Fusarium crown rot, common root rot, and root-lesion nematode, were recorded. Rhizoctonia root rot symptoms were scored according to the degree of spear tipping on the roots; take-all symptoms were recorded according to blackened lesions on roots; Fusarium crown rot symptoms were scored according to the degree of browning on the first stem internode; root-lesion nematode symptoms were scored according to the lack of lateral branching and degree of browning on roots; and common root rot symptoms were scored according to the degree of blackening on the subcrown internode. In total, 40 samples (including stem and root mass) per field sample were scored, representing a combined and cumulative "disease severity" or "root health" score for each of the aforementioned root diseases on a 0-to-5 disease severity scale that will be referred to as root health, where $0 =$ no disease, $1 = 1$ to 5% disease, $2 = 6$ to 25% disease, $3 = 26$ to 50% disease, $4 = 51$ to 75% disease, and $5 = 76$ to 100% disease.

DNA pathogen level estimations (PreDicta-B). Soil sampling was conducted with an AccuCore soil probe (Spur soil probes, Adelaide, SA, Australia) coring device with a 10-mm width to a depth of 10 cm, as suggested by the manufacturer for sampling and according to previous research conducted by SARDI [\(36\)](#page-9-0). In total, 40 cores were taken from the 1-ha area per field surveyed to make a total sample that weighed approximately 500 g. Presow samples were taken in March to April and postharvest samples were taken in approximately November to December to estimate DNA levels both prior to and at the end of the season. DNA extraction was conducted and analyzed in a TaqMan-based assay in 384-well plates according to Ophel Keller et al. [\(36\)](#page-9-0). DNA-based primers are proprietary to SARDI RDTS. Results from the PreDicta-B DNA analysis were reported for the major aforementioned root disease pathogens, including R. solani AG-8, G. graminis var. tritici, F. pseudograminearum, F. culmorum, B. sorokiniana, P. neglectus, P. thornei, and P. teres, in picograms per gram of soil.

Climate and weather data. Weather data for each individual year of the study (2010, 2011, and 2012) were retrieved from 15 weather stations (within 30 km of each sample location) from the DAFWA climate data website and assigned to survey fields based on the nearest proximity to a weather station and the relevant weather conditions, with some interpretation of local conditions ([http://www.agric.wa.gov.au\)](http://www.agric.wa.gov.au). Seasonal data for 2010, 2011, and 2012 was summarized according to weather data for mean soil temperature of monthly combinations of February-March and April-May, and a sum of rainfall for the months of January, February, and March and autumn to early winter months of April, May, and June rainfall for each of the 3 years was utilized as well as for 30-year averages from the aforementioned DAFWA climate website. These specific months were selected and combined in their respective groups according to data from a logistic regression testing the effect of each specific month of the year for soil temperature, ambient air temperature, and rainfall on disease incidence and severity or root health. In addition to logistic regression, a procedure known as Step-AIC was used in R, version 2.15.3, to select the greatest effects of most significant environmental and DNA parameters on each of the respective root disease incidence and the overall disease severity or root health. Step-AIC is a stepwise comparison of model quality with each data parameter by Akaike Information Criteria (AIC) ([61](#page-10-0)).

Meteorological data collection and statistical analysis. Long-term averages were retrieved from the Australian Bureau of Meteorology website from 14 weather stations in close proximity (within 50 km) to each sample location ([http://www.bom.gov.](http://www.bom.gov.au/wa/?ref%20=%20hdr) [au/wa/?ref](http://www.bom.gov.au/wa/?ref%20=%20hdr) = hdr) according to the data available from each location. For the long-term data, we used the climate parameters mean monthly ambient air temperature calculated for similar variables as those described above for February-March and April-May soil temperatures (°C) as well as January-February-March and April-May-June rainfall (mm) from the same monthly periods. Additional parameters included elevation and soil type. Soil type textures at each location were generically classified as s and $= 1$, sandy loam $=$ 2, $\text{loam} = 3$, silty $\text{loam} = 4$, and clay or clay $\text{loam} = 5$, based on interpretations made by the lead author from summaries made by DAFWA soil scientists in Geraldton, Western Australia. Pearson's correlations were calculated between elevation and maximum, minimum, and mean annual soil temperature, as well as mean annual precipitation, disease incidence of the complex of root diseases, and disease severity; and PreDicta-B DNA of each specific root disease organism was calculated to determine initial relationships between parameters.

Means, coefficient of variation (CV) and standard deviations representing soil temperature, rainfall, overall frequency of disease incidence, disease severity or root health, as well as soil DNA levels of various fungi associated with the root disease complex were calculated using SAS (v9.3; SAS Institute Inc., Cary, NC.). Species incidence represented an annual mean occurrence across samples of Rhizoctonia root rot, take-all, Fusarium crown rot, common root rot, and root-lesion nematode symptoms from 40 randomly sampled stem, crown, and root sections from each surveyed field. The relative detection of DNA for R. solani AG-8, G. graminis var. tritici, F. pseudograminearum, F. culmorum, B. sorokiniana, and *Pratylenchus* spp. were converted to $log + 1$ values and analyzed accordingly in a logistic regression using SAS v9.3 for distribution.

Regional analysis according to cereal variety testing zones. Surveyed fields were placed within crop variety testing (CVT) zones published by the DAFWA to provide a basis for spatial analysis ([57\)](#page-10-0). Due to congruencies of weather data and geographical location, zones were statistically analyzed and combined for analysis of disease parameters on the basis of those analyses. Zone 1 represented a combination of high rainfall (450 to 750 mm) and low rainfall (350 to 450 mm). Zones 1 and 2 were combined on the basis of their geography to represent the northern region of Western Australia. Zone 2 comprised 53 fields within the mediumrainfall (325 to 450 mm) southern region of Western Australia. Zone 3 comprised 26 fields within the medium-rainfall (325 to 450 mm) north-central region of Western Australia. Zone 4 comprised 80 fields within the medium-rainfall (325 to 450 mm) central region of Western Australia. Zone 5 comprised 33 fields within the mediumrainfall (325 to 450 mm) zone of the south-central region of Western Australia. Zone 6 comprised agronomic zones of high (450 to 750 mm) and medium (325 to 450 mm) rainfall in the south-central and western regions of Western Australia. Due to an imbalance of field numbers in zones 1 through 4 and 9 through 12, as well as statistical similarities in preliminary regional analyses, the fields surveyed within these zones were grouped into their own zones to make six zones in total, to enhance the balance of sampled fields within agronomic zones. Long-term means of rainfall and temperature data from weather stations in close proximity to survey sites

were obtained from the Australian Bureau of Meteorology website (Table 1). Disease incidence numbers (not proportions), DNA, and environmental (temperature and rainfall) parameters outlined previously were analyzed according to these CVT zones or regions in a general linear model procedure in SAS v9.3.

Logistic and tree regression analysis. In order to model the relationship between root disease, presow soil DNA levels, and environmental climate parameters, each disease incidence response was tested in a logistic and regression tree approach. The presence or absence of each respective root disease measured in the root disease complex, including Rhizoctonia root rot, take-all, crown rot, common root rot, and root-lesion nematodes for each of the total diseased plants out of the total 40 stem or root systems per field, were regressed against soil DNA (measured by PreDicta-B) and environmental April-May and February-March soil temperature as well as April-May-June and January-February-March precipitation parameters separately for each year (2010, 2011, and 2012) using logistic regression analysis with the logistic procedure in SAS v9.3. The terms for the Rhizoctonia root rot incidence logistic model were R. solani AG-8 soil DNA per PreDicta B ($\beta_i x_i$), correction factor (CF) -R. solani AG-8 soil DNA (CF) is the polynomial correction term calculated by squaring the DNA values used to fit them into the logistic model) ($\beta_i x_i$), January-February-March rainfall $(\beta_k x_k)$, April-May-June rainfall $(\beta_l x_l)$, April-May soil temperature ($\beta_{m}x_{m}$), February-March soil temperature ($\beta_{n}x_{n}$), and the error term (E_{ijklmn}) , where Rhizoctonia disease incidence $(Y_{ijklmn}) = \beta_i x_i + \beta_j x_j + \beta_k x_k + \beta_l x_l + \beta_m x_m + \beta_n x_n + E_{ijklmn}$ Diagnostics on the logistic regression models did not indicate any extremely influential points.

In summarizing the root disease and soil DNA data, descriptive statistics did not explain the variation in the data, because there was a lack of relationship between the soil DNA and disease incidence data. Environment was postulated as having a potential effect due to disease–environment interactions described by the "disease triangle" postulated by plant pathologists, describing the relationship of disease as an interaction between the host plant, environment, and plant pathogen (inoculum) levels [\(1\)](#page-9-0). Due to the large number of variables, their potential interactions with each other, and the nonlinearity of the disease data response, a regression decision tree approach that determined thresholds was used to model the disease response in addition to the logistic regression.

Regression tree analysis was conducted using the disease incidence, DNA, and environmental data in R version 2.15.3 ([61](#page-10-0)). A disease severity or root health model was developed from the regression tree analysis utilizing survey data for the respective root disease pathogens (all pathogens involved in the root health disease severity rating) and environmental soil temperature and rainfall values. The terms for the regression tree analysis were R . solani AG-8

TABLE 1. Mean root health or disease severity and incidences (%) of the major root diseases involved in the cereal root disease complex for the respective survey years

	Incidence $(\%)$				
Ratings and disease incidence	2010	2011	2012	Mean ^x	
Root health or disease severity ^y	0.5	1.0	1.3	0.9	
Rhizoctonia root rot ²	15	27	43	28	
Root-lesion nematode ^z	10	28		15	
$Take-allz$	6				
Fusarium crown rot ^z	0		24		
Common root rot ²					

^x Mean represents the average for each species isolated across the 2010, 2011, and 2012 survey years.

^y Root health or disease severity was rated on a scale of 0 to 5, where $0 = no$ disease, $1 = 1$ to 6.25% disease, $2 = 6.25\%$ to 25% disease, $3 = 26\%$ to 50% disease. $4 = 51\%$ to 75% disease, and $5 = 76\%$ to 100% disease.

^z Disease incidence proportions per field of Rhizoctonia root rot, root-lesion nematode, take-all, Fusarium crown rot, and common root rot are expressed as the proportion of plants out of 40 with disease per field.

soil DNA per PreDicta B $(\beta_i x_i)$, CF-R. solani (AG-8) soil DNA $(\beta_i x_i)$, G. graminis var. tritici soil DNA $(\beta_k x_k)$, P. neglectus soil DNA $(\beta_l x_l)$, *F. pseudograminearum* soil DNA ($\beta_m x_m$), summer (January-February-March) rainfall $(\beta_n x_n)$, autumn (April-May-June) rainfall $(\beta_0 x_0)$, autumn (April-May) soil temperature $(\beta_0 x_0)$, summer (February-March) soil temperature $(\beta_{q}x_{q})$ and the error term (E_{ijklmn}) , where disease severity or root health $(Y_{ijklmn}) = \beta_i x_i +$ $\beta_j x_j + \beta_k x_k + \beta_l x_l + \beta_m x_m + \beta_n x_n + \beta_o x_o + \beta_p x_p + \beta_q x_q + E_{ijklmn}$ Environmental parameters of February-March and April-May soil temperature and rainfall were placed in the regression tree models along with the respective DNA and disease data obtained for each field. The models returned the predicted number of diseased plants with poor root health expected according to the model, given levels of pathogen DNA, root disease, and environmental parameters (autumn and summer soil temperature and rainfall) for each field.

RESULTS

The mean frequency of incidence from 40 randomly selected stem and root systems over the three survey years were the greatest for Rhizoctonia root rot (28%) followed by root-lesion nematode (15%), take-all (6%), common root rot (3%), and crown rot (9%) ([Table 1\)](#page-2-0). Mean Rhizoctonia root rot incidence was the greatest for 2012 (42%). Take-all and Rhizoctonia root rot incidence varied to the greatest degree across the 3 years. There were significant differences between years for Rhizoctonia root rot, Fusarium crown rot, and common root rot disease incidences ([Table 1\)](#page-2-0). Rhizoctonia root rot disease incidence had the greatest correlation to root health disease severity over the three survey years ($r = 0.68$, $P < 0.0001$) (Table 2).

The mean root health disease severity rating among fields was 0.5 (10%) in 2010, 1.0 (20%) in 2011, and 1.3 (26%) in 2012, with a percent CV of 144, 139, and 66% for each year, respectively ([Table 1](#page-2-0)). The range for root health disease severity ratings (rated 0 to 5) was from 0 to 31 in 2010, 0 to 72 in 2011, and 0 to 43 in 2012. The $R²$ values and standard errors from Step-AIC revealed that soil temperature and rainfall parameters for the respective months selected had the greatest impact on disease severity and root health.

Of the five root diseases reported in this article, Rhizoctonia root rot and root-lesion nematode symptoms were rated at the greatest frequency ([Table 1\)](#page-2-0). Rhizoctonia disease incidence was most strongly correlated with R. solani AG-8 soil DNA but was also correlated with April-May-June rainfall $(r = -0.31, P < 0.0001)$, and April-May soil temperature $(r = -0.31, P < 0.0001)$ (Table 2). Root health disease severity was most strongly associated with R. solani AG-8 soil DNA, weakly related to January-February-March and April-May-June rainfall, and negatively correlated with February-March soil temperature $(r = -0.30, P < 0.0001)$ (Table 2). Rootlesion nematode incidence was not significantly correlated with any soil DNA pathogen tests but was most closely associated with take-all incidence $(r=0.26, P<0.0001)$ and root health disease severity $(r = 0.27, P < 0.0001).$

Detection of B. sorokiniana (45%) and P. neglectus (39%) soil DNA according to PreDicta-B was the greatest in the study across the 3 years. The percentage of fields with detectable levels of R. solani AG-8 soil DNA (23%) was ranked third of all the pathogens across the 3 years and was the greatest in 2012, with substantial variation across years (Table 3). Detection of G. graminis var. tritici soil DNA only occurred in an average of 3% of the fields. Detection of *P. thornei* and *P. teres* was also low (3 and 9%, respectively). Detection of F. pseudograminearum, B. sorokiniana, and P. neglectus was the greatest in 2012 (Table 3). There were weak associations between R. solani AG-8 soil DNA and monthly rainfall parameters as well as between P. neglectus soil DNA and soil temperatures. Presow and postharvest soil DNA level associations were the strongest for P. neglectus, P. thornei, B. sorokiniana, and R. solani AG-8, showing stability in populations from presow to postharvest sampling [\(Table 4\)](#page-4-0).

Regional analysis utilizing CVT zones. When climate, disease, and pathogen DNA levels were analyzed according to CVT zones, there were significant differences between zones for Rhizoctonia root rot incidence and root health disease severity ([Fig. 1](#page-4-0); [Table 5](#page-4-0)). Due to an imbalance of paddocks per zone, CVT zones were pooled according to similarities in locale and weather parameters and analyzed for Rhizoctonia root rot incidence and root health disease severity. The February-March and April-May soil temperatures were significantly lower in the four southern zones—zone 9 (H4), zone 10 (H5W), zone 11 (M5C), and zone 12 (M5W)—and lower for the northern zones—zone 2 (H2), zone 3 (L1), zone 4 (L2), and zone 5 (M1)—therefore, these were grouped into two different regions ([Table 5](#page-4-0)). Zones 5 and 6 were

TABLE 3. Mean proportions (percentage of fields in Western Australia) of detectable soil DNA before sowing for each pathogen involved in the cereal root disease complex

	Fields with detectable soil DNA $(\%)$ ^y			
Disease (common name), pathogen DNA	2010	2011	2012	Mean ^z
Rhizoctonia root rot, Rhizoctonia solani				
$AG-8$	17	16	37	23
Take-all, Gaeumannomyces graminis				
var <i>tritici</i>	2	4	\overline{c}	3
Crown rot, Fusarium pseudograminearum	12.	12	22	16
Common root rot, Bipolaris sorokiniana	37	48	51	45
Root-lesion nematodes				
Pratylenchus neglectus	38	30	50	39
P. thornei	6	θ	4	3
P. teres	11	8	9	9

^y Numbers represent proportions of fields with detectable soil DNA levels according to PreDicta-B within each year.

^z Mean represents the average for each species isolated across the 2010, 2011, and 2012 survey years.

TABLE 2. Pearson's correlation coefficients for disease incidence and severity compared with soil DNA levels for the most important soilborne pathogens involved in the cereal root disease complex

			Soil DNA			
			Presow	Presow	Postharvest	
Disease, pathogen soil DNA	Rhizoctonia root rot	Disease severity	Rhizoctonia solani	Pratylenchus neglectus	P. neglectus	
Rhizoctonia root rot incidence	1.00	0.68	0.22	0.08	0.04	
P value	\cdots	< 0.0001	0.0005	0.21	0.56	
Disease severity	\cdots	1.00	0.22	0.04	0.04	
P value	\cdots	\cdots	0.0004	0.50	0.56	
Presow R. solani AG-8 soil DNA	\cdots	\cdots	1.00	0.17	0.05	
P value	\cdots	\cdots	\cdots	0.005	0.43	
Presow P. neglectus soil DNA	\cdots	\cdots	\cdots	1.00	0.52	
P value	\cdots	\cdots	.	\cdots	< 0.0001	
Postharvest P. neglectus soil DNA	\cdots	.	\cdots	\cdots	1.00	

characterized by cooler February-March and April-May soil temperatures with higher April-May-June rainfall soil temperatures (Fig. 1; Table 5). The 2010, 2011, and 2012 growing seasons represented distinct diversity in weather patterns (Table 6). The first

TABLE 4. Pearson's correlation coefficients for presow versus postharvest soil DNA levels according to PreDicta-B for the important soilborne pathogens involved in the cereal root disease complex for the 259 fields analyzed over the 3 years of the study (2010, 2011, and 2012)

Disease (common name), pathogen DNA detection $x \rightarrow r$ Valuey		P value ^{z}
Rhizoctonia root rot, Rhizoctonia solani AG-8	0.32	0.0001
Take-all, Gaeumannomyces graminis var. tritici	0.16	0.009
Crown rot, Fusarium pseudograminearum	0.26	0.0001
Common root rot, Bipolaris sorokiniana Root-lesion nematodes	0.55	0.0001
Pratylenchus neglectus	0.52	0.0001
P. thornei	0.69	0.0001
P. teres	0.31	0.0001

^x Pathogen DNA test utilized PreDicta-B for each pathogen and root disease represented.

The r value was calculated according to Pearson's correlation coefficients in SAS v9.3.

 Z P values from Pearson's correlation analysis are considered significant at $P = 0.05$ and below.

Fig. 1. Summary of survey locations, Rhizoctonia solani AG-8 soil DNA, and the cereal variety testing zones outlined by Stephens and Lyons ([57\)](#page-10-0) according to Western Australia rainfall and yield for broad-acre cereal grain crops.

year of the survey, 2010, was characterized by the lowest January-February-March and April-May-June rainfall, which was lower than the other 2 years as well as the 30-year average. The 2011 growing season was characterized by a very wet summer (January-February) (in comparison with the other 2 years) and warmer autumn (April-May) temperatures. Significance and conclusions were interpreted with care from 2011 data, because the number of fields surveyed in that year was very low (50 fields). The 2012 growing season was characterized by a wet April-May (Table 6).

Rhizoctonia disease incidence and root health disease severity were significantly higher in zone 6, characterized by high autumn rainfall and cooler autumn soil temperatures (Tables 5 and [7](#page-5-0)), compared with all other CVT zones in Western Australia. R. solani AG-8 pathogen DNAwas numerically the greatest in zone 6 but not significantly different from zones 2 and 5 [\(Table 8](#page-5-0)). Root health disease severity was predicted to be 26% ([Table 7](#page-5-0)). R. solani AG-8 pathogen soil DNA was significantly higher in zones 2, 5, and 6 ([Table 8](#page-5-0)). Zone 6, representing 21 fields, also had the greatest levels of G. graminis var. tritici and F. pseudograminearum pathogen DNA. Zone 6, with the greatest levels of soilborne pathogen DNA for R. solani AG-8, G. graminis var. tritici, and F. pseudograminearum, also had the greatest level of disease severity [\(Tables 7](#page-5-0) and [8](#page-5-0)).

R. solani AG-8 incidence logistic regression model. A logistic regression model used Rhizoctonia root rot incidence as the measured y variable effect against parameters of R. solani AG-8 pathogen DNA, a polynomial R. solani AG-8 pathogen DNA correction factor (CF-R. solani [AG-8] pathogen DNA), and environmental factors (summer and autumn rainfall and soil temperature). When predicted levels of Rhizoctonia root rot incidence according to the logistic regression model were compared with actual levels measured for the 259 individual fields across the 3 years, the relationship was relatively strong ($R^2 = 0.29$, $r = 0.54$, $P < 0.0001$). When surveyed fields were placed into the respective

TABLE 6. Mean rainfall and soil temperature for the years included in the survey of Western Australia

Monthly weather parameter ^y	2010	2011	2012	Meanz
January-February-March rainfall	34	63	37	50
February-March soil temperature	33	31	28	24
April-May-June rainfall	61	99	103	142
April-May soil temperature			20	

^y Rainfall represent a sum of the rainfall for the months outlined in centimeters and temperature represents the mean of the soil temperature for the months outlined in degrees Celsius.

^z Mean represents the 30-year average for each weather parameter across the 2010, 2011, and 2012 survey years.

TABLE 5. Mean long-term (more than 30-year) values for January-February-March rainfall, April-May-June rainfall, February-March soil temperatures, and April-May soil temperatures for the modified cereal variety testing (CVT) zones^w

		January-February-March	February-March	April-May-June	April-May
CVT zone x	п ^у	Rainfall (cm)	Temp $(^{\circ}C)$	Rainfall (cm)	Temp $(^{\circ}C)$
Zone 1 (H1, H2, L1, L2)	45	38 ns	33a	113 a	23a
Zone $2 \ (M1)^z$	53	40	31 _b	59 c	21 _b
Zone $3 \ (M2)$	26	39	31 _b	62 bc	21 _b
Zone $4 \,(M3)$	80	37	31 _b	73 _b	21 _b
Zone $5 \ (M4)$	33	55	28c	74 h	19c
Zone 6 (H4, H5W, M5C, M5W)	21	47	26d	105a	18 d
Mean	43	43	30	81	21

^w Zones were pooled to improve balance of paddocks within zones. Rainfall amounts are expressed in centimeters per summation of January, February, and March based on long-term averages in centimeters. Soil temperature (Temp) is based on a mean of degrees Celsius per month delineated for months, including the months of February-March and April-May. Numbers followed by the same letter within a column were not significantly different across CVT zones at $P = 0.05$

according to a least significant difference multiple range test; ns = not significant at $P = 0.05$.
CVT zones for Western Australia developed by Stevens and Lyons [\(57](#page-10-0)). Zone 1 = high rainfall, north; zone 2 = medium rainf north-central; zone 4 = medium rainfall, central; zone 5 = medium rainfall, south-central; and zone 6 = high and medium rainfall, south-central and west. *y* Number of fields within each CVT zone.

^z Although zone 2 (H2) is meant to be a high rainfall zone, it appears that rainfall in the autumn period for the weather stations in the vicinity of selected fields surveyed in this region reflected very low rainfall compared with the rest of the year.

CVT zones and predicted and actual means of Rhizoctonia root rot incidence were calculated and compared according to the zones, the relationship was stronger $(R^2 = 0.69, P < 0.0001)$, showing evidence of regional disease risk calculated by the model and levels actually measured. The highest predicted Rhizoctonia root rot incidence was in southern zones 9 (H4), 10 (H5W), 11 (M5C), and 12 (M5W) (Table 7).

Results showed that R. solani AG-8 pathogen DNA had a significant impact on disease severity in 2010 and 2012 but not 2011, which was characterized by higher January-February-March late-summer rainfall [\(Fig. 2](#page-6-0); [Table 9\)](#page-6-0). R. solani AG-8 pathogen DNA explained the greatest amount of variation $(9 \pm 0.22\%$ in 2010 and $2 \pm 0.15\%$ in 2012) while environmental rainfall and soil temperature parameters explained approximately $4 \pm$ 0.005 to $5 \pm 0.13\%$ cumulatively. January-February-March rainfall and February-March soil temperature had a significant impact on Rhizoctonia disease incidence, mainly in 2010 and 2012, and explained 4 ± 0.05 and $2 \pm 0.004\%$, respectively, of the variation in Rhizoctonia root rot incidence. February-March and April-May soil temperatures were significant across all 3 years, with varying degrees of impact in each season. April-May soil temperatures had a positive effect on Rhizoctonia root rot incidence in 2010 and a negative effect in 2011 and 2012, while February-March soil temperatures had a negative impact on Rhizoctonia root rot incidence in 2010 and a positive effect in 2011 and 2012. All rainfall and soil temperature parameters explained $11 \pm 0.005\%$ of the variation in 2010 and $4 \pm 0.13\%$ of the variation in 2012 in the Rhizoctonia disease incidence model. R. solani AG-8 pathogen DNA seemed to have a more definitive impact on disease expression in 2010.The probability of Rhizoctonia root rot incidence was the greatest in 2010 under high April-May-June rainfall and in 2012 with higher February-March soil temperatures and lower January-February-March rainfall [\(Fig. 2](#page-6-0)).

The impact of R. solani AG-8 pathogen DNA was less in 2012 (explaining 2% of the variation), with greater effects of environment and greater probability of Rhizoctonia root rot incidence (ranging from approximately 60 to 90%, with an exponential curve shape as AG-8 pathogen DNA increased from log DNA = 1 to 3) under higher February-March soil temperatures and lower January-February-March rainfall. There was less disease incidence under higher January-February-March rainfall and cooler February-March soil temperatures (ranging from approximately 25 to 75% as AG-8 pathogen DNA increased from log DNA = 1 to 3) [\(Fig. 2C and D](#page-6-0)). There were not enough fields surveyed in 2011 to report results from that year in this part of the data analysis.

Root health disease severity regression tree model. When predicted values from the regression tree model for root health disease severity were compared with the actual measured values, there was a significant relationship ($R^2 = 0.38$, $r = 0.71$, $P < 0.0001$) ([Fig. 3](#page-7-0)). The regression tree analysis determined that environmental parameters of April-May-June rainfall, January-February-March rainfall, and February-March soil temperatures, as well as R. solani AG-8 pathogen DNA, were the most influential in estimating root health disease severity on a 0-to-5 scale. Root health disease severity predicted by the regression tree model was significantly correlated to actual disease severity ([Fig. 3\)](#page-7-0). When fields were

TABLE 7. Mean Rhizoctonia root rot incidence and disease severity estimates predicted by the regression tree model compared with actual measurements according to regions within cereal variety testing (CVT) zonesx

		Rhizoctonia incidence		Disease severity	
CVT zone ^y	n	Predicted ^z	Actual	Predicted ^z	Actual
Zone 1 (H1, H2, L1, L2)	45	12 b	9c	0.90 _b	0.98 _b
Zone $2 \ (M1)$	53	9 d	12 b	0.80 bc	0.80 _b
Zone $3 \ (M2)$	26	8 d	10 _{bc}	0.75c	0.80 _b
Zone $4 \,(M3)$	80	10c	10 _{bc}	0.80 bc	0.80 _b
Zone 5 (M4)	33	11 c	8 c	0.80 bc	0.75c
Zone 6 (H4, H5W, M5C, M5W)	21	18 a	19 a	1.30a	1.30a
Mean	43	16		0.89	0.90

^x CVT zones published by Department of Food and Agriculture of Western Australia (2006). Numbers followed by the same letter within a column are not significantly different at $P = 0.05$ according to Duncan's least significant difference multiple range tests that were performed only on actual measurements.
^y CVT zones were combined due to similarities in climate. Zon

zone 4 = medium rainfall, central; zone 5 = medium rainfall, south-central; and zone 6 = high and medium rainfall, south-central and west.
Predicted values for Rhizoctonia root rot incidence and disease severity are parameters on the basis of pathogen DNA and seasonal soil temperature or rainfall parameters. Numbers represent the number of plants out of 40 per field with the potential to be diseased by Rhizoctonia root rot. Values were standardized according to the number out of 40 plants.

TABLE 8. Mean soil DNA levels measured by PreDicta-B for four of the most prevalent and wide-spread soilborne pathogens involved in the root disease complex according to regions within cereal variety testing (CVT)^u

CVT zone v	n	Rhizoctonia ^w	Gaeumannomyces ^x	Fusarium ^y	Pratylenchus ^z
Zone 1 (H1, H2, L1, L2)	45	9 b		29 _b	0.3c
Zone $2 \,(M1)$	53	18ab	0.8 _b	26 _b	10a
Zone $3 \ (M2)$	26	!b	0.5 _b	4 h	4 b
Zone $4 \ (M3)$	80	6 b	2 _b	23 _b	1 c
Zone $5 \ (M4)$	33	12ab	2 _b	15 b	2 c
Zone 6 (H4, H5W, M5C, M5W)	21	31 a	6 a	346 a	5 b
Mean	43	15.0	2.0	81.7	2.8

u Pathogen numbers (picograms of DNA) followed by the same letters are not significantly different (columns) among regions at $P = 0.05$ according to Fisher's least significant difference multiple range test.

CVT zones published by Department of Food and Agriculture of Western Australia (2006). Zone $1 =$ high rainfall, north; zone $2 =$ medium rainfall, south; zone $3 =$ medium rainfall, north-central; zone 4 = medium rainfall, central; zone 5 = medium rainfall, south-central; and zone 6 = high and medium rainfall, south-central

and west.
 W Rhizoctonia solani (anastomosis group 8) is responsible for causing Rhizoctonia root rot.

x Gaeumannomyces graminis var. tritici is responsible for causing take-all.

y Fusarium pseudograminearum is responsible for causing crown rot.

^z Pratylenchus neglectus is the main causal pathogen species of root-lesion

placed intoCVT zones, model results from the regression tree analysis showed stronger relationships between actual and predicted values $(R² = 0.94, P < 0.0001)$ [\(Fig. 4\)](#page-7-0). Results of the regression tree analysis showed disease levels of 6 to 54%, with the most influential environmental and pathogen parameters being April-May-June rainfall, January-February-March rainfall, F. pseudograminearum pathogen DNA, R. solani AG-8 pathogen DNA, P. neglectus pathogen DNA, B. sorokiniana pathogen DNA, and G. graminis var. tritici pathogen DNA on root health and disease severity.

To interpret the regression tree, follow the tree to the left for results less than the critical values represented in the parentheses and to the right for those greater than the critical values in parentheses. Splits near the top of the tree were the most influential on root health, while those at the bottom were less. The regression tree model showed the greatest risk of decline in root health and increased disease severity (54%) under conditions of April-May-June rainfall greater than 66 mm, low February-March soil temperatures (less than 34° C), high R. solani AG-8 pathogen DNA (greater than log DNA = 0.5), with lower levels of both F. pseudograminearum (less than log DNA = 1.1) and B. sorokiniana pathogen DNA (less than log DNA = 1.7) and low January-February-March rainfall (less than 22 mm) [\(Fig. 5](#page-7-0)). It also showed that, under conditions of high April-May-June rainfall (greater than 66 mm), January-February-March summer rainfall greater than 22 mm but less than 42 mm resulted in 26% disease severity (rating $= 1.3$ on a 0-to-5 scale). These conditions and high F. pseudograminearum pathogen DNA levels (greater than 0.5 log DNA) resulted in 21% disease severity.

P. neglectus pathogen DNA showed a significant effect on disease severity (32%) under conditions of higher April-May-June rainfall

Fig. 2. Probability of Rhizoctonia root rot incidence in Western Australia in 2010 under A, low April-May-June rainfall and B, high April-May-June rainfall and in 2012 under C, low January-February-March rainfall and high February-March soil temperatures and D, high January-February-March rainfall and cool February-March soil temperatures as *Rhizoctonia solani* pathogen DNA increased from $log = 0$ to $log = 3$.

TABLE 9. Model of Rhizoctonia root rot incidence using logistic regression with environmental (rainfall and temperature) and soil DNA parameters over 3 years of a survey study (2010, 2011, and 2012)y

	2010		2011		2012	
Parameter	Estimate	P value	Estimate	P value	Estimate	P value
Intercept	0.44	0.71	2.25	0.02	1.16	0.09
R. solani AG-8 DNA	2.26	< 0.0001	0.37	0.22	0.85	< 0.0001
$CF-R$, solani DNA^z	-1.04	< 0.0001	0.008	0.96	-0.37	< 0.0001
Rainfall						
January-February-March	-0.002	0.46	0.002	0.38	-0.02	< 0.0001
April-May-June	0.02	< 0.0001	0.005	0.005	-0.0004	0.71
Soil temperature						
April-May	0.23	0.02	-0.53	0.0002	-0.21	< 0.0001
February-March	-0.25	< 0.0001	0.24	0.003	0.12	0.002

^y Estimate was calculated as the probability of Rhizoctonia root rot incidence = Rhizoctonia solani AG-8 DNA + I R. solani DNA (squared) + January-February-March rainfall + April-May-June rainfall + April-May soil temperature + February-March soil temperature. Estimates were used to calculate predicted values.

P value less than $P = 0.05$ was considered a significant effect.
^z CF-Rhizoctonia solani soil DNA was squared and as a correction factor for soil DNA in the model to enable it to fit a polynomial curve.

Fig. 3. Regression ($R^2 = 0.38$, $P < 0.0001$) of predicted values of disease severity (rated 0 to 5) according to the logistic regression model plotted against actual measured disease severity values across the 3 years of the Western Australia focus field survey.

Fig. 4. Regression analysis ($R^2 = 0.94$, $P < 0.0001$) plot of predicted values of disease severity (rated 0 to 5) according to the logistic regression model plotted against actual measured disease severity values for cereal variety testing (CVT) zone means across the 3 years of the Western Australia focus field survey.

(greater than 66 mm), cooler February-march soil temperatures (less than 34° C), lower R. solani AG-8 pathogen DNA levels (less than log 0.5 DNA), and P. neglectus DNA levels greater than log $DNA = 0.28$ but less than $log DNA = 3.5$. G. graminis var. tritici pathogen DNA significantly affected disease severity (26%) under the aforementioned conditions, with April-May-June rainfall between 102 and 119 mm and G. graminis var. tritici pathogen DNA greater than log DNA = 0.52 (Fig. 5). All of these pathogens seemed to interact together under conditions of high F. pseudograminearum (greater than $log DNA = 1.1$) and G. graminis var. *tritici* pathogen DNA (greater than log DNA = 0.52) and higher R. solani AG-8 pathogen DNA levels (greater than $log DNA = 0.5$) all interacting with the aforementioned environmental conditions of cool February-March soil temperatures and higher April-May-June rainfall to result in 30% disease severity and decline in root health (Fig. 5).

DISCUSSION

This study documents the interactions between environmental conditions, pathogen DNA concentration in soil preseeding, and root disease expression in the crop at anthesis. Our research agrees with that of others and has shown that root health in cereals can be predicted with a certain level of confidence using both a logistic regression and regression tree approach, with significant $R²$ values of 0.94 similar to that of Okubara et al. [\(34\)](#page-9-0), Poole et al. ([39](#page-10-0)), and Lawes et al. [\(24\)](#page-9-0). The regression tree analysis showed that root health disease severity was greatest under higher April-May-June rainfall, cooler February-March soil temperatures, higher R. solani AG-8 soil DNA concentrations (DNA at greater than 0.5 log pg/g of soil), lower levels of other soilborne pathogens (namely F. pseudograminearum, P. neglectus, and B. sorokiniana), and lower January-February-March summer rainfall. These results really showed that all of the pathogens interacted together in a competitive relationship, where R. solani AG-8 seemed to dominate under lower summer rainfall and disease levels were

Fig. 5. Regression tree of predicted disease severity given Rhizoctonia solani AG-8, Fusarium pseudograminearum (Fp) Pratylenchus neglectus (Pn), and G. graminis var. tritici (Ggt) pathogen DNA levels and different levels of autumn (April-May-June) rainfall, summer (February-March) soil temperature (temp), and summer (January-February-March) rainfall. Critical levels of variables are within parenthesis with units for rainfall (millimeters), pathogen DNA (log DNA + 1 pg/g of soil), and soil temperature in degrees Celsius. Follow the diagram to the right for levels exceeding critical levels and to the left for levels less than critical levels. Disease severity predicted by the model is represented by the proportional numbers at the bottom.

greater in many instances when other pathogen DNA concentrations (i.e., P. neglectus, F. pseudograminearum, and B. sorokiniana) were lower.

Many pathologists have reported the association of less R. solani inoculum and disease pressure under conditions following higher summer rainfall in Australia and the United States ([15–17](#page-9-0),[19,28,29](#page-9-0),[34](#page-9-0),[42\)](#page-10-0). Mazzola and Gu ([28\)](#page-9-0) reported that R. solani AG-8 interacted with the microbial community in a diverse way and, with time, would be outcompeted from increased microbial activity under higher moisture conditions. Roget et al. $(41, 42)$ $(41, 42)$ $(41, 42)$ $(41, 42)$ $(41, 42)$ also reported a "suppressiveness" of R. solani in soils where microbial activity was accelerated from increased moisture and organic matter following summer rainfall events. Gill et al. [\(19\)](#page-9-0) reported that disease severity caused by R. solani AG-8 decreased by 69% under wetter soil moisture levels of -3 to -10 kPa and concluded that soil microbial activity increased under increasing soil moisture, postulating that the activity was responsible for suppressive effects on the Rhizoctonia fungus.

Okubara et al. [\(34\)](#page-9-0) also reported lower R. solani AG-8 pathogen DNA levels under environments with higher precipitation in a survey conducted in the PNW of the United States. There were also lower concentrations of R. solani AG-8 pathogen DNA following broadleaf crops. R. solani AG-8 pathogen DNAwas most strongly negatively correlated with summer rainfall in the PNW of the United States (July to September, $r = -0.35$, $P < 0.0001$). Our results also showed that Rhizoctonia disease incidence and overall disease severity were significantly reduced under conditions of higher (January-February-March) summer rainfall, when modeled with R. solani AG-8 pathogen DNA. Thus, our results agree with those of several others reporting lower Rhizoctonia disease and inoculum following higher moisture patterns ([5](#page-9-0),[16,19](#page-9-0),[28,34](#page-9-0),[41,42](#page-10-0)). Our study also showed significant differences in disease incidence across years as well as within seasons.

In our study, interactions between R. solani AG-8 and environment differed between 2010 and 2012 according to the logistic regression model. Although Rhizoctonia root rot incidence was only weakly associated with presow DNA levels in preliminary correlations, there were large differences in environmental conditions between both seasons. In 2010, which was generally characterized as a warm, dry year, the impact of R. solani AG-8 was greater in CVT zones with higher autumn rainfall and cool autumn or summer temperatures. In 2012, which was characterized as having cooler than average autumn temperatures, there was a greater risk of Rhizoctonia root rot in CVT zones with relatively warm, dry summers and cool autumn temperatures. The reduced slope of the logistic curve for 2012 compared with 2010 indicates that R. solani AG-8 had greater impact on root health in 2010. This indicates that, in the case of this study, $R.$ solani AG-8 was driving the root disease complex interaction between pathogen levels presowing and environment and cereal root health across different regions and seasons.

Aside from showing the magnitude of root disease incidence variation across 3 years, this study also shows the diverse interactions between seasonal environmental conditions as well as the pathogenic microbial community measured by PreDicta-B and the environments of crop production similar to those reported by others [\(4,5,8,9](#page-9-0),[18,34](#page-9-0)[,55\)](#page-10-0). Rhizoctonia and root-lesion nematode disease incidence measured by visually rating plants were most closely correlated with overall disease severity and root health. However, relationships between R. solani AG-8 pathogen DNA, Rhizoctonia disease incidence, and April-May-June and January-February-March rainfall parameters were most closely associated with overall disease severity and declining root health. The regression tree analysis showed that pathogen DNA levels were interactive in somewhat of a complex where, under certain conditions, disease severity was greater when certain pathogen levels (such as F. pseudograminearum and B. sorokiniana) were lower under certain environmental conditions. P. neglectus and the take-all pathogen DNA (G. graminis var. tritici) had more significant effects on disease severity at lower R. solani AG-8 pathogen DNA levels, while F. pseudograminearum dominated to the left of the disease severity regression tree under lower April-May-June rainfall conditions and higher January-February-March summer rainfall conditions where R. solani AG-8 pathogen DNA was not interactive. Thus, this study suggests that certain combinations of both pathogen levels and environments lead to greater disease severity and decline in root health. There is room in future reports of this study to include crop rotational strategies as an additional parameter to include in the disease severity model.

Several studies have reported significant effects of cropping rotational systems and tillage practices on the incidence of Rhizoctonia root rot and other root diseases [\(16,18](#page-9-0),[20](#page-9-0),[22,23,34,38,](#page-9-0) [40–42](#page-10-0),[44](#page-10-0),[45,47,48,54,55](#page-10-0)). Trends have shown decreased Rhizoctonia spp. incidence under conditions following a broadleaf crop and barley crops [\(34,](#page-9-0)[42,47](#page-10-0)). Smiley et al. [\(55\)](#page-10-0) reported increased Rhizoctonia root rot following continuous winter wheat compared with following no-till winter pea. However, they reported that inoculum density (picograms of DNA per gram of soil) of R. solani AG-8 did not change across crop rotations and cropping systems. Schroeder and Paulitz ([47](#page-10-0)) reported higher Rhizoctonia root rot and R. solani AG-8 hyphal activity after 3 to 4 years of transitioning to direct seeding. Although these trends are interesting to report, crop rotational history was not taken into account in this portion of this modeling study because plants were sampled from cereal crops from year to year and likely resembled pressure found in continuous cropping systems lacking rotation.

Environmental conditions and expression of cereal root disease varied significantly both in regional locations outlined by the CVT zones and between the 3 years in this study. The analysis from this study shows that the effects of temperature and rainfall were significant across regions and all 3 years and the incidence and severity cereal root disease varied between the CVT zones ([57](#page-10-0)). Also, pathogen levels varied between the regional zones; zone 6, for example, had greater amounts of R. solani AG-8, G. graminis var. tritici, and F. pseudograminearum and resulted in the greatest disease severity during the course of this study.

Rhizoctonia root rot $(R. \, \text{solani}$ AG-8) had the greatest impact on root health across years. Fusarium spp., Pratylenchus spp., and B. sorokiniana were often detected in the fields. However, their contribution to explaining variation in root health was relatively low. Of the Pratylenchus spp. assessed, P. neglectus was the most widespread and in greatest abundance, occurring in over 39% of the fields [\(53\)](#page-10-0). P. teres was detected in 9% of fields. Presow and postharvest pathogen DNA levels were the most correlated for P. thornei, P. neglectus, and B. sorokiniana, followed by R. solani AG-8.

Root diseases in the Northern region of Australia encompassing parts of Queensland and New South Wales are dominated by a different pathogen complex (namely, P. thornei, P. neglectus, and Fusarium spp. causing crown rot) and are not as affected by environment, where pathogen DNA levels are better predictors of in-crop disease [\(49,59](#page-10-0)). Even though the combination of environmental and pathogen DNA parameters in this survey were significant and useful in interpreting disease interactions and relationships, pathogen and environmental parameters explained a relatively low proportion of variation in root health ranging from 2 to 11%, with environmental parameters explaining most of the relationships. Similar to the findings of Okubara et al. ([34](#page-9-0)), these results show that, regardless of the small amount of variation explained by the statistical analyses, substantial biological inferences can be drawn that agree with that of others ([17](#page-9-0),[28](#page-9-0),[41,42\)](#page-10-0).

This study shows that data from different fields across a broad range of districts and seasons can be used to model variation in root health, similarly to data of the take-all model [\(8\)](#page-9-0). These results provided the framework for an interpretive logistic algorithmic model that more accurately estimates the risk of cereal root diseases based on presow pathogen DNA levels and actual field survey root disease measurements taken across years. When samples come into the lab and DNA levels are reported, the model could potentially be used to provide an associated risk of disease severity based on the sample location, survey results, and potential growing season scenarios. This predictive model will be expanded and confirmed with data from future seasons across years and provided with the pathogen DNA test results to provide more accurate root disease predictions. With further development, this root health model will allow growers and industry to more accurately predict yield loss based on PreDicta-B soil DNA analyses and more accurately adjust their cropping programs to minimize risk of yield loss due to soilborne cereal root diseases.

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