Winery On-site Assessment of Grapevine Bunch Rot: In Pursuit of Sustainable Practices

Cornelissen, R.J.^{1,3}, Aleixandre Tudo, J.L.^{2,4}, Nieuwoudt, H.H.^{1*}

- (1) South African Grape and Wine Research Institute, Stellenbosch University, Private Bag X1, Matieland, 7602, South Africa
- (2) Department of Viticulture and Oenology, Stellenbosch University, Private Bag X1, Matieland, 7602, South Africa
- (3) Namaqua Wines, P.O. Box 75, Vredendal, 8160, South Africa
- (4) Instituto de Ingeniería de Alimentos para el Desarrollo (IIAD), Departamento de Tecnología de Alimentos (DTA), Unversitat Politecnica de Valencia (UPV), Valencia, Spain

R.J. Cornelissen: Current E-mail address: marina@namaquawines.com;

Submitted for publication: December 2022 Accepted for publication: May 2023

Keywords: Producer winery, botrytis rot, sour rot, wine quality, rot assessment, severity, winery intake, sustainability, infrared spectroscopy

Producer wineries are responsible for processing 75 % of South Africa's annual wine grape production. A characteristic of producer wineries is the processing of large volumes of wine grapes, thus incorporating enormous variability in grape quality, which includes the presence of grapevine bunch rots. Grapevine bunch rots are detrimental to grape and wine quality. The "Introduction" provides background information on the industrial working conditions at producer wineries and the economic effect of rot on wine production. The next two sections discuss the principal reasons for bunch rot being an inevitable part of grape production, namely cultivar susceptibility and climatic conditions. The challenges regarding grapevine bunch rot assessment, specifically concerning producer wineries or wineries processing large amounts of grapes, are set out in "Assessing rot intensity". Representative sampling, mechanical harvesting of grapes and visual evaluation are discussed from an industrial perspective. The last section of this review focuses on the quest for sustainable appraisal of grapevine bunch rot. Infrared spectroscopy (IR) could provide a sustainable option for objectively assessing rot on-site at producer wineries. However, even with the availability of plenty of spectroscopic methods and the demonstrated potential of IR spectroscopy for rapid assessment of grapevine bunch rot, these methods are yet to be applied routinely under industrial working conditions.

INTRODUCTION

Industrial scenarios at large producer wineries

Currently, South Africa has 43 producer wineries responsible for processing 75 % or 1 094 000 tonnes of South Africa's annual wine grape production (SAWIS, 2021). A producer winery, also referred to as a producer cellar, is defined as "a role player or entity where grapes are received and processed on behalf of a group of wine grape producers, its members, into wine grape products and the marketing thereof as packaged or bulk" (SAWIS, 2021). Globally, producer wineries account for one-quarter to one-third of global wine production, with well-known winery examples in France, Italy, Spain, Germany, and South Australia (Bezuidenhout, 2014).

A unique characteristic of producer wineries is the scale of production associated with processing large volumes of wine grapes. Thirty-two of South Africa's producer wineries process more than 10 000 tonnes per annum (SAWIS, 2021), but processing could amount to as much as 60 000 tonnes per annum at a single premise. Producer wineries arguably have advantages in the cost-effectiveness of operational costs due to economy of scale and market competitiveness by guaranteed volume supply to retail. However, the production of large volumes of grapes by many grape growers from commercial vineyards spread out over large surface areas creates significant logistical challenges, primarily related to grape quality control and resulting wine quality.

The producer winery's operational structure usually includes the advisory service of viticulturists, aiming to provide input on vineyard management practices at the farm level and thus manage grape quality according to desired wine style. Reviewed by Steel *et al.* (2013), the detrimental effect of grapevine bunch rots on grape berry composition and eventual wine quality is well-researched.

*Corresponding author: E-mail address: hhn@sun.ac.za

Acknowledgements: Funding of the research by South African Grape and Wine Research Institute (SAGWRI), Stellenbosch University, South Africa, the National Research Foundation (NRF) of South Africa, and Namaqua Wines, South Africa

Thus, rot assessments are imperative to grape quality control at producer wineries. Constraints in terms of time, cost, and trained assessors required for in-field rot assessments justify consideration of alternatively assessing grape quality at the point at which grapes are delivered on-site at the winery, also referred to as grape or winery intake. Assessing individual grape loads at winery intake would also encompass in-vineyard variability, often overlooked by in-field assessments. It is long known that rot can exist as "pockets" in vineyard blocks (Seem, 1984). Grape (load) assessment at winery intake would also be advantageous if grape transportation by road from farm to winery is needed. Extended travel time and exposure to high temperatures during transport could significantly increase the microbial activity in grape loads (Swanepoel, 2006). In the instance where microbial rot pathogens are already present on the grapes in the vineyard, these travelling conditions could further deteriorate grape quality.

Currently, the parameters assessed at winery intake include sugar concentration (°Brix), pH, total titratable acidity (TA, measured in g/L), temperature (°C) and visual assessment of grape loads for rot and matter other than grapes (MOG). These parameters are limited since they do not account for the detrimental chemical compositional changes in the grape berries brought about by rot. Furthermore, the reported subjectivity and biased nature of visual rot assessments (Hill et al., 2014) often have led to disputes between grape growers and wineries (Longbottom et al., 2013). Logistical challenges during harvest require that producer wineries often employ seasonal workers with minimal knowledge of/or training in the visual assessment of rot in grape loads delivered to the winery. Furthermore, assessments must be done fast and not slow down harvestrelated operations, which are hectic at large producer wineries.

Considering that this final grape quality evaluation at winery intake will ascertain grape growers' income, wineries are responsible for ensuring that quality assessment is objective and accurate. However, the presence of rot also has an impact on winery profits.

Economic implications of rot on wine quality

The wide variety of grapevine bunch rots was discussed in a large body of recently published literature (Jackson, 2014; Ioriatti *et al.*, 2015; Steel *et al.*, 2016; Hall *et al.*, 2018; Steel *et al.*, 2018; Gao *et al.*, 2020; Crandall *et al.*, 2022; Kellner *et al.*, 2022). Regarding the economic impact of rot on wine production, it is especially botrytis rot and sour rot that are of concern (Ky *et al.*, 2012; Hall *et al.*, 2018; Molitor *et al.*, 2018). Thus, this review focuses on the deleterious botrytis rot, also referred to as grey rot and sour rot.

The economic impact of rot can already be observed during the early stages of wine production by the significant decrease in juice yield obtained from rot-affected compared to healthy grapes. Sour rot-affected Cabernet Sauvignon grapes with 30 % severity could lead to as much as a 25 % decrease in the volume of grape juice recovered for fermentation (Barata *et al.*, 2011b). Calculating a theoretical scenario: One tonne of healthy grapes provides an average juice yield of 770 L (Price Waterhouse Cooper, 2013), compared to rotaffected grapes, which could decrease the volume of juice recovered per tonne of grapes by as much as 25 %. This would mean that a winery could incur an estimated loss of as much as 40 % in South African rand (ZAR) value for each tonne of rot-affected grapes received compared to healthy grapes (Table 1).

In addition, rot increases wine production costs with specific interventions needed to mitigate the effect of rot on the grape must and wine matrix. For example, the presence of laccase in grape must and wine poses a serious threat to the quality of wine produced. Laccase is an oxidising enzyme produced by *Botrytis cinerea*, the causative fungi responsible for botrytis rot. Uncontrolled oxidative reactions of phenolic compounds, catalysed by laccase, cause browning and premature ageing of wine (Claus *et al.*, 2014). Processing strategies to reduce laccase activity include short heat treatment of grape must and application of oenological tannins as early as crushing (AWRI, 2023; Vignault *et al.*, 2019, Claus, 2020).

Another processing difficulty with a consequent increase in production costs caused by rot, is the early clogging of filter membranes. Wine filtration removes unwanted particles from the wine and enhances its stability. Decreased filterability is associated with the increased presence of β -glucans in wines (Francioli *et al.*, 1999). *Botrytis cinerea* produces β -glucans responsible for botrytis rot in wine grapes (Jackson, 2014). Remedial actions include the addition of enzymes responsible for the hydrolysis of glucans to improve the filterability of wines (Jadhav & Gupta, 2016; Espejo, 2021).

However, the detrimental effect of rot on wine quality is primarily associated with the negative impact on wines' sensory properties. Rot could have a detrimental impact on the aroma of grape must and wine due to the following reactions: The laccase-induced oxidation of phenolic aroma compounds, as earlier discussed, the production of detrimental flavours, and the hydrolysis of aromatic grape components. The latter two complications are discussed in the following paragraphs.

Meneguzzo et al. (2008) reported that Gewürztraminer wines made from rot-affected grapes differed significantly from healthy grapes due to the sensory detection of mould flavour and a dominating acetic character in rot-affected wines. Off-flavours regularly associated with rot-affected wines include earthy, mouldy, and fresh mushroom (Steel et al., 2013). Sweet-like honey off-flavours of ethyl phenylacetate and phenylacetic acid are associated with wine produced from sour rot-affected grapes (Barata et al., 2011a). Volatile markers reported for fungal rots include sesquiterpenes, 1,5-dimethylnaphthalene, 2-(4-hexyl-2,5-dioxo-2,5dihydrofuran-3-yl) acetic acid, m-cresol, γ-nonalactone (Schueuermann et al., 2019) and 2-transhexenal (Santos et al., 2022).

Besides producing unwanted sensory volatiles, rot can also induce wine aroma losses. Volatile organic compounds (VOCs) are the main contributor to wine aroma (Dudareva *et al.*, 2013). Botrytis rot berry modifications could include the transformation of monoterpenes, the aromatic compounds responsible for the varietal aromas of Muscat cultivars, into less odorous compounds (Rienth *et al.*, 2021). Furthermore, the disappearance of fruity aromas in young wines could be related to the hydrolysis of fatty acids' ethyl esters, which contribute to the fermentation aromas of wine by the esterase activity of *Botrytis cinerea*. Barata *et al.* (2011b) suggested that sour rot significantly affects the wine yeast's secondary metabolism, decreasing the levels of volatiles related to fatty acids and amino acid synthesis. Unwanted sensory properties in wine or the loss of varietal character could warrant wine quality downgrades with consequent economic implications for wineries (Table 1).

Furthermore, rot could also influence the food safety aspects of wine via the production of the mycotoxin, ochratoxin A (OTA) (Rousseaux *et al.*, 2014; Welke, 2019; Ubeda *et al.*, 2020). Classified by the International Agency for Cancer Research's (IARC) monographs (1993) in Group 2B as "possibly carcinogenic to humans", OTA levels in wine are regulated (European Commission, 2006). Selected winemaking practices could decrease OTA levels in wine, but these practices could be detrimental to the impact of wine's positive volatile compounds (Gambuti *et al.*, 2005).

GRAPE CULTIVAR SUSCEPTIBILITY TO ROT

Over time, an extensive body of literature has developed on wine grape cultivars' natural differences in susceptibility to

TABLE 1

Estimated loss of income (%) per tonne of rot-affected grape	s.
--	----

rot. As an example of ranking cultivars' susceptibility to rot, research by Paňitrur-De La Fuente *et al.* (2018) is discussed. Based on visual assessment, the in-field susceptibility of 13 wine grape cultivars naturally infected with botrytis rot was compared and classified in relation to fruit maturity under Chile and France's contrasting climate and vineyard practices. Despite differing climatic conditions and vineyard practices, cultivar susceptibility ranking on both sites was similar. High susceptibility to botrytis rot was associated with the cultivars Sauvignon blanc and Gewürztraminer, followed by Chardonnay and Pinot noir. Petit Verdot, Cabernet Sauvignon, Mourvèdre and Syrah were highly resistant to botrytis rot (Paňitrur-De La Fuente *et al.*, 2018).

Cultivar characteristics regularly reported as influencing susceptibility to rot include bunch compactness (Hed *et al.*, 2009; Lisek & Lisek, 2021), berry skin traits such as physical berry resistance and cuticle thickness (Herzog *et al.*, 2015). Grapevine breeders consider loose bunch architecture as the most important selection criterium to improve cultivars' resistance to botrytis rot (Herzog *et al.*, 2022). Bunch compactness is defined as the tightness or the spatial arrangement of berries within a bunch (Tello & Ibáñez, 2018). Figure 1 illustrates an example of the difference in

Estimated for	(70)	for tonne of for uncered g	Jupes.		
Cultivar	Juice yield of healthy grapes (L/tonne)	Healthy grapes traded as cultivar wine (ZAR/tonne of grapes)	Juice yield of rot- affected grapes (L/tonne)	Rot-affected grapes traded as dry red/white wine (ZAR/tonne of grapes)	Loss/tonne of rot- affected grapes ⁽⁴⁾ (%)
Cabernet Sauvignon	770L ⁽¹⁾	ZAR7 253 ⁽³⁾	558L ⁽²⁾	ZAR4 380 ⁽³⁾	-40%
Chenin blanc	770L ⁽¹⁾	ZAR4 543 ⁽³⁾	558L ⁽²⁾	ZAR2 784 ⁽³⁾	-39%

⁽¹⁾Average juice recovery (L)/tonne of healthy grapes at approximately 770L/tonne (Price Waterhouse Cooper, 2013). ⁽²⁾Average juice recovery (L)/tonne of rot-affected grapes estimated at 558L/tonne (Barata *et al.*, 2011b). ⁽³⁾Wine price (ZAR/L): Cabernet Sauvignon – ZAR9.42/L; Chenin blanc – ZAR5.90/L; Dry Red – ZAR7.85/L; Dry White – ZAR4.99/L (SAWIS, 2022). ⁽⁴⁾Decrease in juice yield + wine quality downgrade.



FIGURE 1

Bunch compactness of white wine grape cultivars.

Left: Healthy Colombar bunches with berries in loose contact with each other. Right: Botrytis rot susceptible cultivar Sauvignon blanc with berries not readily movable. Regions where berries are dense are indicated with white circles, and often the growth of *Botrytis cinerea* begins in these dense bunch regions because of berry injury.

bunch compactness amongst cultivars.

Several factors contribute to dense bunches being more susceptible to rot. After flower dehiscent, compact bunches retain more flower debris between berries than more loose bunches leading to rot, for example, by harbouring Botrytis inocula (Hed et al., 2009). Insufficient development of berries' epicuticular wax between compact berries reduces resistance to pathogen attack (Marois et al., 1986). Berries in dense bunches press against each other as berry size increases towards maturity, which could lead to berry rupturing. This injury to the berry skin provides entry for secondary microorganisms while leaking grape juice supplies nutrients that stimulate microbial growth. Berry injuries are considered the primary source of sour rot (Hall et al., 2017; Ioriatti et al., 2018). The dense arrangement of berries could prevent fungicide spray from penetrating between berries and faster drying of bunches after rainfall, leaving a conducive environment for developing rot (Hed et al., 2009). However, Molitor et al. (2018) found that differences in clonal bunch compactness do not primarily explain the natural variation in botrytis rot susceptibility of nine White Riesling clones. The authors suggested that the general link between bunch compactness and rot susceptibility may not hold true withincultivar, for example, for a group of clones from the same cultivar.

The berry skin is an important physical and mechanical barrier to rot. Thin-skinned cultivars are more prone to injury, providing entry for microbial colonisation and rot development. A recent study by Lisek & Lisek (2021) concluded that the resistance of cultivars to sour rot strongly correlated with increased berry skin thickness. For example, research has provided evidence that the berry skin of Ugni blanc (115 μ m) is significantly thinner compared to the skins of Chardonnay (203 μ m) and Viognier (243 μ m) (Jin *et al.*, 2017). Thin-skinned berries are also more predisposed to rot in high temperatures and high visible light vineyard environments. These conditions more easily cause skin injury to thin-skinned berries compared to thicker-skinned cultivars (Steel *et al.*, 2016; Lisek & Lisek, 2021).

Berry cuticle thickness and hydrophobic epicuticular waxes on the berry outer skin also influence cultivars' rot susceptibility (Herzog *et al.*, 2015). The cuticle covers the outer surface of the berry skin. In-field evaluation of naturally infected grapes showed that a thicker cuticle associated with epicuticular waxes negatively correlates with susceptibility to botrytis rot, especially for compact-bunch cultivars (Herzog *et al.*, 2015).

Considering reported differences in cultivar susceptibility to rot, it is conceivable that grape growers should be advised to plant rot-resistant cultivars and clones. However, cultivar choice is a long-term and strategic investment. Over time wineries build a relationship with the retail and clients, which could include the guaranteed volume supply of specific varietal wines. Furthermore, prominent wine regions in the world are characterised by cultivar-specific production; these include Merlot from Bordeaux, (Ky *et al.*, 2012), Sauvignon blanc and Sémillon from Sauternes (Kallitsounakis & Catarino, 2020), Chenin blanc from the Loire Valley, France (Carbajal-Ida *et al.*, 2016), and the sought-after, unique fruity Sauvignon blanc wines from Marlborough, New Zealand (Benkwitz *et al.*, 2012). These cultivars are known to be susceptible to rot (Paňitrur-De La Fuente *et al.*, 2018), but due to being an economic driving force of these wine regions, they will continue to be planted for many years. Furthermore, even targeting cultivar and clone selection strategies associated with resilience to rot, climatic conditions still significantly influence the intensity of rot.

CLIMATIC INFLUENCES ON BOTRYTIS AND SOUR ROT OF GRAPEVINE BUNCHES

Botrytis rot can either manifest as unwanted grey rot, also called gray mold or just bunch rot, or as the special form of Botrytis cinerea infection, namely noble rot (Jackson, 2014). Noble rot is associated with the highly valued sweet white wine styles of Sauternes (Landrault et al., 2002) and Tokaj (Furdiková et al., 2019). Under controlled conditions, optimal botrytis rot development is observed at high relative humidity (\geq 90 %) and cool temperatures of 20°C to 25°C (Ciliberti et al., 2015). Very specific climatic conditions of humid nights, cool and foggy mornings, followed by dry, sunny, and windy afternoons are required for the development of noble rot (Furdíková et al., 2019). However, the occurrence of this rare botrytised noble rot grapes is geographically restricted (Kallitsounakis & Catarino, 2020), and under most conditions, botrytis rot is undesirable. Optimum conditions for the development of grey rot include temperatures of approximately 18°C to 22°C, high humidity (> 90 %), and prolonged wet conditions (Würz et al., 2020; Rienth et al., 2021). In contrast to the cool temperatures associated with botrytis rot, sour rot is favoured by hot summers (VanderWeide et al., 2020; Lisek & Lisek, 2021). Average monthly temperatures of approximately 18°C during berry ripening were reported by Lisek & Lisek (2021) for the vintages having severe sour rot. Significant to inducing sour rot is rainfall from veraison to harvest (VanderWeide et al., 2020; Lisek & Lisek, 2021). Rainfall during berry ripening could lead to grape berry splitting. Berry injuries are conducive to the development of sour rot by providing grape juice as a suitable substrate for yeast and acetic bacteria proliferation (Ioriatti et al., 2018).

Different amounts and the distribution of rainfall, which lead to the increase in relative humidity and berry wetness duration, influence the intensity of rot (Becker & Knoche, 2012; Paňitrur-De La Fuente et al., 2018). Botrytis rot's potential infection occurrences under future European climatic conditions (2030 and 2050) were modelled using 15 years of temperature and leaf wetness duration data (Bregaglio et al., 2013). Mixed results emerged from the study, ranging from more than a 100 % increase in the occurrence of botrytis rot, maintaining similar levels of pressure compared to the current situation, or decreasing disease occurrences. The rise in botrytis rot relates to more humid conditions, favouring pathogen development. Bregagli et al. (2013) concluded that botrytis rot will remain a relevant problem to European wine production in the next 10 to 30 years. Using South African weather data from 1984 to 2015 and modelling plausible future conditions, similar heterogeneous climatic changes and conditions for the future were reported (Southey, 2022).

The published literature mainly focuses on the economic effect of a change in temperature on wine production (Ashenfelter & Storchmann, 2014). However, the effect of climate change relating to increased rainfall events during the summer months, i.e. the berry ripening period from veraison to harvest, is also of economic concern to the wine industry. These rainfall events would significantly impact the intensity of sour rot. An increase in the intensity of sour rot has already been observed in Europe (Hausinger *et al.*, 2015; Lisek & Lisek, 2021). In-field observations of an increase in sour rot intensity could explain the series of recent studies on the etiology of sour rot (Ioriatti *et al.*, 2015; Madden *et al.*, 2017; Hall *et al.*, 2018; Hall *et al.*, 2019; Entling *et al.*, 2019; Pinto *et al.*, 2019; Gao *et al.*, 2020).

Long-term strategies to lower the intensity of rot related to climatic conditions include site-specific plantings to avoid climatic conditions favourable to the development of rot. However, there is no strong indication in the literature that this strategy has been followed. For example, recent research suggests that new plantings are established under climatic conditions, knowing that botrytis rot will be problematic (Würz *et al.*, 2020). Due to the dominant effect of climatic conditions on rot, it can be concluded that rot will inevitably be part of grape production.

Earlier literature reviews focused on the predisposing factors contributing to the development of rot, microorganisms associated with grapevine bunch rots, disease expression such as visual symptoms, the effect of rot on grape composition and wine chemistry, as well as vineyard management practices aimed at the control of rot (Mundy, 2008; Steel *et al.*, 2013; Rousseaux *et al.*, 2014; Latorre *et al.*, 2015; Kallitsounakis & Catarino, 2020; Rienth *et al.*, 2021; Crandall *et al.*, 2022). However, grapevine bunch rot assessment from an industrial perspective has not been reviewed before. In the following sections, this review provides an industrial perspective on challenges specific to large or producer wineries concerning the assessment of grapevine bunch rots.

ASSESSING ROT INTENSITY

Before discussing the challenges relating to the accurate assessment of rot under industrial working conditions, a summary of relevant terms in the field of plant disease assessment is given. For a comprehensive glossary, the reader is referred to Bock *et al.* (2022a).

Plant disease assessment or phytopathometry concerns the visual estimation or instrument or sensor-based measurement of the amount of plant disease that usually is symptom-based (Bock *et al.*, 2022a). Disease assessment focuses on detection, identification, quantification, or a combination of these objectives (Bock *et al.*, 2022a). Disease intensity is a general term used to describe the amount of disease present in a population (Nutter, 1991). It can be defined by the terms prevalence, incidence, or severity (Bock *et al.*, 2022a). In earlier research, prevalence, incidence, and frequency were often used as synonyms (Seem, 1984). However, incidence is the proportion or number of the whole of disease entities within a sampling unit, and prevalence is the proportion of diseased plots or fields in a defined area (Bock *et al.*, 2022b). On the other hand, severity measures the quantity of disease within the sampling unit (Bock *et al.*, 2022b).

Compared to prevalence or incidence, severity is more evident in research describing rot intensity on wine grape cultivars. The reason could be that as severity increases, the detrimental effect on wine quality also increases (Zoecklein, et al., 2000; Barata et al., 2011b; Steel et al., 2018). Globally, visual assessment is still the reference method to determine grapevine bunch rot intensity in-field, which could include visually estimating the percentage of berries infected per bunch, the percentage of tissue infected per bunch, or severity estimations according to scales (Hed et al., 2009; Hill et al., 2013; Molitor et al., 2020; Steel et al., 2020). Rot intensity assessment of grape loads at winery intake includes visual detection of rot-affected bunches, and/or calculating the severity by percentage of weight of rot in the sample. (Durgun, 2010). Bock et al. (2022a) strongly advised clearly defining how a disease assessment term is applied under specific industrial conditions. Wineries often impose a price penalty if rot intensity exceeds a certain threshold (Hill et al., 2014). Imposing price penalties could severely reduce grape growers' income; accordingly, rot intensity assessments must follow a standard protocol and ensure repeatability, objectivity, and accuracy. The next section will discuss three challenges concerning rot intensity assessment from an industrial perspective. These challenges include representative sampling, mechanical harvesting of grapes and visual assessment of rot.

Challenges relating to rot intensity assessment under industrial working conditions

Trustworthy assessment stems from representative samples (Wagner & Esbensen, 2015). The reality of variability within a vineyard is well-known (Bramley *et al.*, 2011), which also applies to the variability of disease intensity within a single vineyard block. Clumping or pockets of rot within a single vineyard block is not uncommon, and rot intensity could differ between parts in the same vineyard block (Evans *et al.*, 2010). Under industrial conditions, the sample size, and the spatial distribution from which the sample is collected, should be representative of the vineyard block if the assessment is executed at vineyard level; or the grape load which is assessed at winery intake. Non-representative sampling procedures may lead to invalid results.

Accuracy, or the lack thereof, in vineyard sampling comes with an increasing cost for both grape growers and wineries. Theoretically calculated, Hill et al. (2019) estimated that annual crop losses/ha could be as high as NZ\$2 578 (approximately ZAR27 000/ha) due to incorrect estimates of botrytis rot severity in vineyards. The cost concerning the time spent and human resources needed to accurately assess many vineyard blocks in a large producer winery's setup could render in-vineyard assessment unsustainable. Interpolating vineyard blocks from small samples taken within the specific vineyard block was proposed by Hill et al. (2019) to reduce the amount of sampling. Assessing rot intensity at winery intake could provide an economical substitute for in-vineyard checks. However, mechanisation of the grape harvesting process confronts winery intake visual assessment with structural integrity loss of bunches

(Durgun, 2010), which obscures the presence of rot.

When grapes are machine-harvested, mechanical actions are exerted on grapevines to remove the berries from the stalks. Grape bunches and berries lose their structural integrity due to the mechanical forces exerted on them. Grape loads delivered to the winery consist of a shapeless mass of juice, berry skins and pulps called grape mash; see Figure 2 for a visual demonstration of the effect of mechanical harvesting on grape bunches. However, regarding the sampling strategy of grape loads at winery intake, the sample size of machineharvested grape loads could be smaller compared to handharvested loads. Marois *et al.* (1993) found that hand-harvest grape loads had a more aggregated distribution of *Botrytis cinerea* than machine-harvested grape loads, which had a more uniform distribution of *Botrytis* antigen.

In addition to the above two challenges, objective visual assessment proves to be difficult. For a review of the sources of error related to the visual assessment of plant diseases Bock *et al.* (2010) can be consulted. Under industrial working conditions, subjectivity can intensify by using seasonal employees without prior training or experience with visual rot assessments, as is often employed in large wineries. Several studies have demonstrated that experienced assessors estimate severity more accurately (Bock *et al.*, 2022b).

Aside from the subjectivity and bias associated with visual assessment, this method does not fully capture the unseen chemical berry composition changes induced by rot. In the event of latent rot, visual symptoms would not yet have manifested, but chemical changes to the berry have already started to occur (Versari *et al.*, 2008). It is also possible that an extended rot period increased the concentration of rot-associated grape berry metabolites compared to a shorter infection period. Still, visually berries are assessed with the same intensity (Fischer & Berger, 2007).

Since early investigations, such as Berg's *et al.* (1958) study into a practical grading system for defect detection at winery intake, efforts are still ongoing in the quest to improve the objectivity and speed of rot intensity assessments. The following section critically reviews the industrial application of sustainable rot assessment in the last 15 years, from 2007 to 2022.

THE QUEST FOR SUSTAINABLE ASSESSMENT OF GRAPEVINE ROT IN THE FOURTH INDUSTRIAL REVOLUTION

Sustainability has three core concepts, namely economic, environmental, and social sustainability (Contini & Peruzzini, 2022). Characteristic of the current fourth industrial revolution (Industry 4.0) is digitisation and automation in the manufacturing environment aiming to increase sustainability (Ortt *et al.*, 2020; Hassoun *et al.*, 2023). The vine and wine sectors also focus on digitisation to increase efficiency, productivity, transparency, value proposition and new business models, and sustainability (OIV, 2021).

Initiatives for sustainable development in the vine sector were launched in recent years. Examples of this inventiveness applied under industrial working conditions include the use of the normalised difference vegetation index (NVDI) that characterises in-vineyard variability (Best *et al.*, 2015; Pañitrur-De la Fuente *et al.*, 2020), and crop protection

modelling (Molitor *et al.*, 2020) which use meteorological conditions to predict the risk of rot.

Sustainable development in wine grape production also embraces replacing wet chemistry methods with nondestructive technologies such as infrared (IR) spectroscopy to quantify grape compounds rapidly (Swanepoel et al., 2007; Petrovic et al., 2020; Ferrer-Gallego et al., 2022). Infrared spectroscopy methods contribute to all three of the core sustainable concepts. Concerning the economic concept, accuracy, speed, and simplification of IR analysis methods can provide rapid results for actionable operational insights. Instrumentation and so-called global calibration models for the prediction of grape compounds are also available at a reasonable cost. Environmentally, IR methods reduce the use of harsh chemicals and thus also reduce chemical waste. Socially, since less or no chemicals are used during analysis, operators' health and safety are also considered. It is also possible for non-specialist operators to use spectroscopy instrumentation (Marcinkowska et al., 2019) as part of routine applications.

The above sustainable development initiatives indicate that the vine sector may be receptive to decision support using technological interventions. Infrared spectroscopy methods could provide options for sustainable rot assessment, and grape production can notably benefit from rot assessment decision support (Porep *et al.*, 2014; Schmidtke *et al.*, 2019). However, hurdles need to be cleared to achieve automated rot assessment including ensuring representative sampling, data analytics for actionable insights and winery decisions on sorting rot-affected grapes into different production streams.

Evaluating the progress towards this stance, published studies in the last 15 years, from 2007 to 2022, on the use of IR spectroscopy for grapevine bunch rot assessments in wine grapes are critically reviewed from an industrial perspective (Table 2).

Overview of published studies (2007 to 2022) using infrared (IR) spectroscopy for grapevine bunch rot assessment

Quantifying rot intensity was the focus of earlier IR studies (Versari et al., 2008; Durgun, 2010; Hill et al., 2014). In these studies, Fourier-transform mid-infrared (FTIR-MIR) showed an ability to predict severity (%) with $R^2 = 0.8$ (Versari *et al.*, 2008) and $R^2 \ge 0.9$ (Durgun, 2010; Hill *et al.*, 2014). Near infrared (NIR) spectroscopy also demonstrated potential for quantifying rot intensity ($R^2 = 0.9$). However, for the ease of in-field industrial application, Hill et al. (2014) recommended the use of digital image analysis instead of IR spectroscopy techniques. Digital image analysis does not require additional equipment for in-vineyard assessments and could save time. Still, emphasising the previously discussed challenges of rot assessment for producer wineries, the time implications of in-vineyard assessments are not sustainable. However, with an RPD of 2.5 and 2.0 for NIR and FTIR-MIR severity (%) predictions, respectively (Hill et al., 2014), these methods show potential for screening purposes at winery intake. Hill et al. (2014) noted that compared to visual assessments, both the beforementioned IR techniques appear to be more accurate at predicting severity \geq 50 %, and considerable nonlinearity was observed in low severities of ≤ 25 %.



FIGURE 2 Mechanical harvesting of wine grapes.

Left: Grape bunches still intact on the grapevine before mechanical harvesting. Middle: When grapes are machine-harvested, mechanical actions are exerted on grapevines to remove the berries from the stalks. After mechanical harvesting only the stalks of the grape bunches are left on the grapevine. Interestingly, the brown stalks indicated by a white circle, represent the bunches with rot, compared to the green stalk of a healthy grape bunch. Right: Machine-harvested grapes are delivered to the winery as a shapeless mass of juice, berry skins and pulps called grape mash.

Infrared spectroscopy techniques were especially explored for the detection of rot (Table 2). Detection of grapevine bunch rot could be achieved by following either of two strategies, namely detecting an increase in the concentration(s) of rot-associated disease marker(s) (Versari et al., 2008; Hausinger et al., 2015; Porep et al., 2015a; Porep et al., 2015b; Gelhken et al., 2022), or discriminating between rot-affected and healthy grapes (Beghi et al., 2017; Giovenzana et al., 2017; Giovenzana et al., 2018). The compounds in the beforementioned studies that were targeted to support the rapid detection of rot by increased concentrations include gluconic acid and glycerol (Versari et al., 2008; Hausinger et al., 2015; Porep et al., 2015a), acetic acid (Porep et al. 2015a), the fungal sterol, namely ergosterol (Porep et al., 2014; Porep et al., 2015b), as well as VOC's (Gelhken et al., 2022). Interestingly, the classification studies of Giovenzana et al. (2018) and Beghi et al. (2017) only aimed at discriminating between healthy and rotaffected grapes, which will be downgraded by the winery for high-quality vinification. From an industrial perspective, a two-class grading system could be limiting, especially in a year with extreme weather conditions conducive to rot development. A third group of "classified as infected, but in need of winemaking remedy" is probably a more realistic situation at a producer winery.

The identification of rot pathogens using IR spectroscopy has only recently been explored (Table 2). Although Giovenzana et al. (2018) did report visual spectral differences between grape must infected with botrytis compared to sour rot, no classification modelling was performed. Schmidtke et al. (2019) were the first to explore the use of IR techniques in combination with classification modelling for identifying rot pathogens in grape must. It should be noted that the latter study was conducted on a small sample set consisting of only 30 samples per pathogen. Furthermore, grape berries were inoculated with pure cultures; thus, identification of these rot pathogens will have to be tested with naturally infected berries (Schmidtke et al., 2019). Although the results from this study are promising for identifying rot pathogens, pathogen identification does not provide an answer to the amount of rot present. Knowing the intensity of rot is crucial to guide mitigation practices in wineries.

In addition, some of the reviewed studies listed in Table 2 lack industrial robustness, for example by using only a small sample size per parameter that was modelled. For optimal results with calibration modelling, more than 75 samples might be required (Versari et al., 2008). Furthermore, variation in the sample set must encompass different cultivars, vintages, climatic conditions and more than one type of rot. Rot rarely exhibits a single form under natural conditions. An example of lacking robustness for industrial application could be drawn from Durgun (2010) who explored the use of FTIR-MIR and Raman spectroscopy for rot quantification in the severity range of 0 % to 5 %. Even though this rot intensity range is likely to result in the downgrading of grapes at winery intake, it does not represent the observed in-field severities. Thus, this model would be limiting in sorting grapes into different production streams within the winery. Robust models can perform measurements in the various scenarios encountered under industrial conditions (Giovenzana et al., 2018).

Near-infrared spectroscopy features strongly in rot assessment studies (Table 2). Earlier studies did explore the potential of transmittance FTIR-MIR spectroscopy for rapid detection of rot (Versari *et al.*, 2008; Durgun, 2010). However, the sample preparation requirements for transmission FTIR-MIR could be a drawback for rapid analysis at winery intake. Attenuated total reflectance mid-infrared (ATR-MIR) spectroscopy, which requires minimal or no sample preparation for analysis, could fulfil the requirements for routine application at winery intake. Schmidtke *et al.* (2019) is the only study exploring the potential use of ATR-MIR in rot assessments.

Importantly, spectroscopy methods are coupled with multivariate data analysis (MVDA) techniques, such as partial least square (PLS) regression and partial least square discriminant analysis (PLS-DA) (Table 2). Spectroscopic methods lead to the generation of large volumes and a variety of data, also referred to as big data (Simsek *et al.*, 2019). Although information-rich, this spectral data needs to be transformed into decision-support tools relevant for the problem at hand. The process of extracting information from these datasets is called data mining (Provost & Fawcett, 2013). However, the decision-support information needed

Published studies in the last	15 years (21	007 to 2022) that foc	sussed on the	use of infrared	(IR) spectrosco	opy in grapevine l	ounch rot assessme	ent.	
References					Rot			Reference method	Value for indus-
Countries where research			Sample	Rot	assessment	IR technique		used for rot	trial decision
was conducted	Cultivars	Sample numbers	matrix	pathogen	objective	MVDA	Spectral range	assessment	support
Gelhken <i>et al.</i> (2022) Germany	n = 20 Red and	n = 725 For individual	Naturally infected	n/a	Detection	Vis/NIR (reflectance)	400 to 1700 nm	GC-MS	Rapid, easy, and simultaneous
×	white cultivars	regression modelling of grape aroma	grape mash collected			PLS			quantification of aroma quality of wine grapes
		compounds sample numbers ranged from 47 to 725	at winery intake						which could relate to rot
Schmidtke <i>et al.</i> (2019) Australia	n = 1 White: Chardon- nay	n = 120 30 samples for each of four fun- gal rot pathogens	Grape berry juice from in- oculated fresh ber- ries	Aspergillus niger, Aspergillus carbonarius, Botrytis cinerea, Pencillium expansum	Identifica- tion	ATR-MIR SIMCA, SVR, KNN, RFM	1850 to 375 cm ⁻¹	Pure fungal cul- tures analysed using ATR-MIR	Potential rapid discrimination between Aspergil- lus spp., Botrytis cinerea, and Peni- cillium expansum.
Giovenzana <i>et al.</i> (2018) Giovenzana <i>et al.</i> (2017) Italy	n = 15 Red and white cul- tivars	n = 159 Healthy = 98; Diseased = 88; of which Botrytis cinerea = 23; Powdery mildew = 15; Sour rot = 23.	Naturally infected grape must col- lected by auger from grape load	Botrytis cine- rea, sour rot	Detection	Vis/NIR (re- flectance and transmittance) PLS-DA	400 to 1650 nm	Visual rot assess- ment of grape bunches: Infection level 0 (healthy) versus infection level 1 (extended colonies of > 2 mm with	Potential clas- sification of dis- eased and healthy samples
Beghi <i>et al.</i> (2017) Italy	n = 7 Red and white cul- tivars	n = 2 559 Healthy = 1 235; Diseased = 1 324; of which Botrytis cinerea = 684; Powdery mildew = 395; Sour rot = 245	at winery intake					evident grey- whitish powdery appearance and brown senescent- necrotic lesion)	

TABLE 2

TABLE 2 (CONTINUED)									
References					Rot			Reference method	Value for indus-
Countries where research		Succhanne of Lanco	Sample	Rot	assessment	IR technique	Canadan I and an	used for rot	trial decision
was conducted	Cullivars		IIIaurix	paulogen	objective	MVDA © . 1	opectral range	assessificit	support
Hausinger <i>et al.</i> (2015) Germany	n = 4 Red: Pi- not noir White: Riesling, Müller- Thurgau,	n/a	Naturally infected grape must	Botrytis ci- nerea	Detection	Optical grape sorter	Optical grape sorter with an integrated high- speed red, green and blue camera system and an infrared laser	Enzymatic determination of gluconic acid and glycerol concen- trations	Automated optical grape-sorting gave successful separation between healthy and rot-affected berries which
	LIIIOL BIIS								could field to en- sure and improve wine quality
Porep <i>et al.</i> (2015a) Germany	n = 12 Red and white cul- tivars	n = 1 160 Red = 698; White = 462	Grape mash col- lected after	n/a	Detection	Vis/NIR (reflectance) PLS	Vis: 450 to 850 nm NIR: 1050 to 1650nm	Routine wet chemistry methods: enzymatic,	Vis/NIR potential tool for rapid, multifactorial, and holistic as-
			destem- ming and crushing at winery					automatic titrator, density meter, pH meter	sessment of grape chemical compo- sition. PLS cali- bration models for 10 grape param- eters were built
Porep <i>et al.</i> (2015b) Germany	n = 5 Red: Dornfelder Pinot Meunier Trollinger Lemberger Schwarz- riesling White: Müller-	n = 177	Grape mash collected after destem- ming and crushing at winery	n/a	Detection	Vis/NIR (reflectance) PLS	400 to 1800 nm	HPLC-UV analysis for determination of ergosterol content	Vis/NIR applied as a rapid tool for online determina- tion of ergosterol under industrial working conditions

TABLE 2 (CONTINUED)									
References					Rot			Reference method	Value for indus-
Countries where research			Sample	Rot	assessment	IR technique		used for rot	trial decision
was conducted	Cultivars	Sample numbers	matrix	pathogen	objective	MVDA	Spectral range	assessment	support
Porep et al. (2014)	n = 6	n = 168	Grape	n/a	Detection	Vis/NIR	400 to 1800 nm	Reference analysis	Rapid quan-
Germany	Red and		mash from			(reflectance)		of:	tification of
	white cul-		naturally			FT-NIR	650 to 2500 nm	Ergosterol -	rot-associated
	tivars		infected			(reflectance)		HPLC; laccase	disease markers
			grapes, as well as			PLS		activity - colouri- matric accave	with prediction
			as well as mash from					oliicose friictose	from differen-
			grapes					gluconic acid, ma-	tiation between
			infected					lic acid, glycerol,	low and high
			with pure					acetic acid, etha-	concentrations,
			mould					nol - enzymatic	rough estimates or
			strains					UV test kits;	semi-quantitative,
								titratable acidity	to quantitative
								- automatic titra-	predictions of
								tor; tartaric acid -	glycerol and er-
								Rebelein method;	gosterol
								pH - pH-meter,	
								density – density meter	
Hill <i>et al.</i> (2014)	n = 1	n = 66	Naturally	Botrytis ci-	Quantifica-	NIR	NIR: 1260 to	Visual rot assess-	Suitable alterna-
Australia	White.		infootod		+;	(noff notion on)	1270	mont octimotion	time for anoth
Ausuana	w nue: Riesling		grape	nereu	11011	(renectance) MIR	13/0 mm MIR: 1142 to	of severity (%)	uves for quantum- cation of botrytis
	0		bunches			(reflectance)	1050 cm^{-1}		rot severity in
			and ber-			Digital image	Digital image		white wine grapes
			ries			analysis	analysis: Hue		
							distribution		
							information is		
							used to deter-		
							attributable to		
							diseased and		
							healthy bunch tissue		

TABLE 2 (CONTINUED)									
References					Rot			Reference method	Value for indus-
Countries where research			Sample	Rot	assessment	IR technique		used for rot	trial decision
was conducted	Cultivars	Sample numbers	matrix	pathogen	objective	MVDA	Spectral range	assessment	support
Durgun (2010)	n = 2	n = 174	Grape	Aspergil-	Quantifica-	FTIR-MIR	MIR:	Healthy and	MIR (transmit-
California	Red:	Zinfandel = 66;	must from	lus niger,	tion	(transmit-	3044 to 1887	mouldy homog-	tance) and Raman
	Zinfandel	Chardonnay =	inoculate	Penicillium		tance)	cm ⁻¹ and 1555 to	enates were mixed	spectral analysis
	White:	108	berries	italicum,		Raman	$950 {\rm cm}^{-1}$	in weight:weight	showed promise
	Chardon-			Rhizopus		PLS	Raman: 3400 to	(%) to obtain se-	for objective
	nay			stolonifer			$200~\mathrm{cm}^{-1}$	verities 0.5%, 1%,	quantification of
								2%, 3%, 4%, 5%	grapes' mould
									content
									MIR models con-
									tained white and
									red cultivars
Versari et al. (2008)	n = 2	n = 320	Naturally	Botrytis ci-	Detection	FTIR-MIR	Three spectral	Visual rot assess-	Rapid method for
Italy	Red: San-	Hand-picked in	infected	nerea	Quantifica-	(transmit-	regions:	ment	the simultaneous
	giovese	vineyard = 200 ;	grape		tion	tance)	2971 to 2435	HPLC analysis of	measurement of
	White:	Mechanically	must			PLS	cm^{-1} ;	gluconic acid and	the rot-associated
	Trebbiano	sampled from					2280 to 1717	glycerol concen-	disease markers
		grape loads at					$cm^{-1};$	trations	gluconic acid and
		winery intake =					1543 to 965 cm ⁻¹		glycerol. Models
		120							contained white
									and red cultivars.
									MIR (transmit-
									tance) suitable for
									process control
ATR-MIR = attenuated total rei HPLC-UV = high performance of cultivars or samples. $n/a = n$	flectance mid- liquid chroma ot available. 1	infrared. FTIR-MIR = atography-ultra-violet s NIR = near-infrared. P	Fourier transfi spectroscopy. I 'LS = partial le	orm mid-infrared. R = infrared. KNN sast square. PLS-]	. GC-MS = gas ch N = k-nearest neij DA = partial leas	nromatography-ma ghborhood. MIR = st square discrimin	ss spectrometry. HPI - mid-infrared. MVD, ant analysis. RFM =	JC = high performance l A = multivariate data and random forest model. V	iquid chromatography. lysis. n = total number 'is/NIR = visible/near-

infrared. SIMCA = soft independent modelling of class analogies. SVR = support vector regression.

for industrial implementation will be revealed only if MVDA data are combined with problem-specific interpretation (Esbensen & Julias, 2009). For an introduction to the MVDA methods coupled with spectroscopy in wine and grape analysis, the reader is referred to Cozzolino *et al.* (2009) and Musingarabwi *et al.* (2016).

CONCLUSIONS

Grapevine bunch rots are an inevitable part of grape production. They force attention to the unique challenges of producer wineries where large volumes of grapes need to be assessed for their health status in a relatively short harvest period. The need for rot assessment as part of grape quality evaluation relates to rot's detrimental effect on grape composition and wine quality. Inaccurate rot assessments lead to a loss in income, both for grape growers and wineries. Although grape cultivars naturally differ in susceptibility to rot, climatic conditions have overriding influences on the intensities of encountered infections.

From the published articles on the use of IR methods for grape rot assessments (Table 2), it is evident that the studies predominantly aimed at the detection of rot by rapid quantification of rot-associated disease markers. However, the first step towards the detection of rot at winery intake using rot-associated disease markers would require the identification of disease marker threshold values for rejection.

Infrared spectroscopy may provide sustainable options for rapid and objective rot assessment for routine application at producer wineries. The potential use of ATR-MIR for rapid detection and quantification of rot under industrial working conditions is unexplored. Furthermore, MVDA featuring discriminant analysis for quantifying the severity (%) has not been done before.

Compared to the fruit and vegetable sectors' use of decision support by automated analyses (Giovenzana *et al.*, 2015), the vine and wine sector fail to keep pace. Creating an example for practical application and overcoming practical challenges would benefit sustainable practices in the wine industry.

However, the first step for industrial implementation would be based on having a standardised tool to assess rot at winery intake. As discussed by Hill et al. (2014), using a standardised alternative method for rot assessment would depend on the goal. The question arises whether the alternative method must predict severity (%) directly comparable to visually estimated severity (%)? Thus, to be used as a threshold level upon which penalties would be implied? Or is the goal to find an alternative assessment method that removes the subjectivity of visual assessment regardless of the variable used? The latter is a more generalised strategy and advantageous for implementation under different industrial working situations. The next step would be for wineries customising the assessment method for example, per cultivar and even per cultivar per quality goal. Following this stage, would be aligning process flows for an increase in the potential value of wine by sorting grapes according to health status into different production streams.

LITERATURE CITED

Ashenfelter, O. & Storchmann, K., 2014. Wine and climate change. American Association of Wine Economists (AAWE) Working paper, 152, 1-43.

Australian Wine Research Institute (AWRI), 2023. Fact sheet: Managing *Botrytis*-infected fruit. https://www.awri.com.au Date of access: 06.04.2023.

Barata, A., Campo, E., Malfeito-Ferreira, M., Loureiro, V., Cacho, J. & Ferreira, V., 2011a. Analytical and sensorial characterization of the aroma of wines produced with sour rotten grapes using GC-O and GC-MS: Identification of key aroma compounds. J. Agric. Food Chem. 59, 2543-2553.

Barata, A., Pais, A., Malfeito-Ferreira, M. & Loureiro, V., 2011b. Influence of sour rotten grapes on the chemical composition and quality of grape must and wine. Eur. Food Res. Technol. 233(2), 183-194.

Becker, T. & Knoche, M., 2012. Water induces microcracks in the grape berry cuticle. Vitis 51(3), 141-142.

Beghi, R., Giovenzana, V., Brancadoro, L. & Guidetti, R., 2017. Rapid evaluation of grape phytosanitary status directly at the check point station entering the winery by using visible/near infrared spectroscopy. J. Food Eng. 204, 46-54.

Benkwitz, F., Tominaga, T., Kilmartin, P.A., Lund, C., Wohlers, M. & Nicolau, L., 2012. Identifying the chemical composition related to the distinct aroma characteristics of New Zealand Sauvignon blanc wines. Am. J. Enol. Vitic. 63(1), 62-72.

Berg, H.W., Alley, C.J. & Winkler, A.J., 1958. Investigation of defects in grapes delivered to California wineries: 1957. Am. J. Enol. Vitic. 9(1), 24-31.

Best, S., Leon, L. & Quintana, R., 2015. Development of a differential grape harvesting methodology. Chem. Eng. Trans. 44, 295-300.

Bezuidenhout, F., 2014. South African producer/co-operative wineries. Thesis, Cape Wine Masters, South Africa. Downloaded from https://icwm. co.za/members/bezuidenhout/ Date of access: 18.11.2022.

Bock, C.H., Pethybridge, S.J., Barbedo, J.G.A., Esker, P.D., Mahlein, A.K. & Del Ponte, E.M., 2022a. A phytopathometry glossary for the twenty-first century: towards consistency and precision in intra- and inter- disciplinary dialogues. Trop. Plant Pathol. 47(1), 14-24.

Bock, C.H., Chiang, K.S. & Del Ponte, E.M., 2022b. Plant disease severity estimated visually: a century of research, best practices, and opportunities for improving methods and practices to maximize accuracy. Trop. Plant. Pathol. 47(1), 25-42.

Bock, C.H., Poole, G.H., Parker, P.E. & Gottwald, T.R., 2010. Plant disease severity estimated visually, by digital photography and image analysis, and by hyperspectral imaging. CRC. Crit. Rev. Plant Sci. 29(2), 59-107.

Bramley, R.G.V., Trought, M.C.T. & Praat, J-P., 2011. Vineyard variability in Marlborough, New Zealand: Characterising variation in vineyard performance and options for the implementation of Precision Viticulture. Aust. J. Grape Wine Res. 17(1), 72–78.

Bregaglio, S., Donatelli, M. & Confalonieri, R., 2013. Fungal infections of rice, wheat, and grape in Europe in 2030-2050. Agron. Sustain. Dev. 33(4), 767-776.

Carbajal-Ida, D., Maury, C., Salas, E., Siret, R. & Mehinagic, E., 2016. Physico-chemical properties of botrytised Chenin blanc grapes to assess the extent of noble rot. Eur. Food Res. Technol. 242, 117-126.

Ciliberti, N., Fermaud, M., Roudet, J. & Rossi, V., 2015. Environmental conditions affect *Botrytis cinerea* infection of mature grape berries more than the strain or transposon genotype. Phytopathology 105(8), 1090-1096.

Claus, H., 2020. How to deal with uninvited guests in wine: Copper and copper-containing oxidases. Fermentation 6(38), 1-14.

Claus, H., Sabel, A. & König, H., 2014. Wine phenols and laccase: An ambivalent relationship. In: El Rayess, Y. (eds.). Wine - Phenolic Composition, Classification and Health Benefits, (1st ed.) pp 155-185.

Contini, G. & Peruzzini, M., 2022. Sustainability and Industry 4.0: Definition of a set of key performance indicators for manufacturing companies. Sustainability 14, 1-37.

Cozzolino, D., Cynkar, W.U., Shah, N., Dambergs, R.G. & Smith, P.A., 2009. A brief introduction to multivariate methods in grape and wine analysis. Int. J. Wine Res. 1, 123-130.

Crandall, S.G., Spychalla, J., Crouch, U., Acevedo, F.E., Naegele, R.P. & Miles, T.D., 2022. Rotting grapes don't improve with age: Cluster rot disease complexes, management, and future prospects. Plant Dis. 106(8), 2013-2022.

Dudareva, N., Klempien, A., Muhlemann, J.K. & Kaplan, I., 2013. Biosynthesis, function and metabolic engineering of plant volatile organic compounds. New Phytol. 198(1), 16-32.

Durgun, H., 2010. Quantification of rot in wine grapes. https://scholarworks. calstate.edu/downloads/pz50gx20m?locale=en Date of access: 19.09.2022.

Entling, W., Anslinger, S., Jarausch, B., Michl, G. & Hoffmann, C., 2019. Berry skin resistance explains oviposition preferences of *Drosophila suzukii* at the level of grape cultivars and single berries. J. Pest Sci. 92, 477-484.

Esbensen, K.H. & Julius, L.P., 2009. 4.01 - Representative sampling, data quality, validation – A necessary trinity in chemometrics. In: Brown, S.D., Tauler, R. & Walczak, B. (eds). Comprehensive chemometrics, Chemical and biochemical data analysis, (1st ed.) pp. 1-20.

Espejo, F., 2021. Role of commercial enzymes in wine production: a critical review of recent research. J. Food Sci. Technol. 58, 9-21. https://doi. org/10.1007/s13197-020-04489-0

European Commission, 2006. Commission regulation (EC) No. 1881/2006 of 19 December 2006. Setting maximum levels for certain contaminants in foodstuffs. Official Journal of the European Union, L364/5-L364/24 https://eurlex.europa.eu/LexUriServ/LexUriServ.do?uri=OJ:L:2006:364:0005:002 4:EN:PDF Date of access: 15.11.2022.

Evans, K.J., Dunne, K.J., Riches, D., Edwards, J., Beresford, R.M. & Hill, G.N. 2010. Effective management of botrytis bunch rot for cool climate viticulture. Grape and Wine Research and Development Corporation, Project Number: UT 06/01, University of Tasmania. https://www.wineaustralia. com/getmedia/3eb20c93-e535-4e1b-9989-7ebc16b3ce3c/UT-06-01 Date of access: 30.01.2022.

Ferrer-Gallego, R., Rodríguez-Pulido, F.J., Toci, A.T. & García-Estevez, I., 2022. Phenolic composition, quality and authenticity of grapes and wines by vibrational spectroscopy. Food Rev. Int. 38(5), 884-912.

Fischer, U. & Berger, T., 2007. Objective measurement of grape soundness. Focus 31(1), 18-21.

Francioli, S., Buxaderas, S. & Pellerin, P., 1999. Influence of *Botrytis cinerea* on the polysaccharide composition of Xarel.Lo must and cava base wines. Am. J. Enol. Vitic. 50(4), 456-460.

Furdíková, K., Machyňáková, A., Drtilová, T., Klempová, T., Ďurčanská, K. & Špánik, I., 2019. Comparison of volatiles in noble-rotten and healthy grape berries of Tokaj. LWT - Food Sci. Technol. 105, 37-47.

Gambuti, A., Strollo, D., Genovese, A., Ugliano, M., Ritieni, A. & Moio, L., 2005. Influence of enological practices on ochratoxin A concentration in wine. Am. J. Enol. Vitic. 56(2), 155-162.

Gao, H., Yin, X., Jiang, X., Shi, H., Yang, Y., Wang, C., Dai, X., Chen, Y. & Wu, X., 2020. Diversity and spoilage potential of microbial communities associated with grape sour rot in eastern coastal areas of China. PeerJ 2020(6), 1-17.

Gehlken, J., Pour Nikfardjam, M. & Zörb, C., 2022. Determination of aroma compounds in grape mash under conditions of tasting by on-line near-infrared spectroscopy. Eur. Food Res. Technol. 248(9), 2325-2337.

Giovenzana, V., Beghi, R., Brancadoro, L. & Guidetti, R., 2017. Classification of wine grape based on different phytosanitary status by using visible/near infrared spectroscopy. Chem. Eng. Trans. 58, 331-336.

Giovenzana, V., Beghi, R., Civelli, R. & Guidetti, R., 2015. Optical techniques for rapid quality monitoring along minimally processed fruit and vegetable chain. Trends Food Sci. Technol. 46(2), 331-338.

Giovenzana, V., Beghi, R., Tugnolo, A., Brancadora, L. & Guidetti, R., 2018. Comparison of two immersion probes coupled with visible/near infrared spectroscopy to assess the must infection at the grape receiving area. Comput. Electron. Agric. 146, 86-92.

Hall, M., Loeb, G. & Wilcox, W., 2017. Defining and development management strategies for sour rot. Res. Focus 2017-3 Cornell Viticulture and Enology 30, 1-7. https://grapesandwine.cals.cornell.edu/files/shared/Research%20Focus%202017-3.pdf Date of access: 04.01.2020.

Hall, M.E., Loeb, G.M., Cadle-Davidson, L., Evans, K.J. & Wilcox, W.F., 2018. Grape sour rot: A four-way interaction involving the host, yeast, acetic acid bacteria, and insects. Phytopathology 108(12), 1429-1442.

Hall, M.E., O'Bryon, I., Wilcox, W.F., Osier, M.V. & Cadle-Davidson, L., 2019. The epiphytic microbiota of sour rot-affected grapes differs minimally from that of healthy grapes, indicating causal organisms are already present on healthy grapes. PLoS One 14(3), 1-12.

Hassoun, A., Jagtap, S., Garcia-Garcia, G., Trollman, H., Pateiro, M., Lorenzo, J.M., Trif, M., Rusu, A.V., Aadil, R.M., Šimat, V., Cropotova, J. & Câmara, J.S., 2023. Food quality 4.0: From traditional approaches to digitalized automated analysis. J. Food Eng. 337, 1-16.

Hausinger, K., Lipps, M., Raddatz, H., Rosch, A., Scholten, G. & Schrenk, D., 2015. Automated optical grape-sorting of rotten grapes: effects of rot infections on gluconic acid concentration and glycerol/gluconic acid ratios in must and wine. J. Wine Res. 26(1), 18-28.

Hed, B., Ngugu, H.K. & Travis, J.W., 2009. Relationship between cluster compactness and bunch rot in Vignoles grapes. Plant Dis. 93(11), 1195-1201.

Herzog, K., Schwander, F., Kassemeyer, H.H., Bieler, E., Dürrenberger, M., Trapp, O. & Töpfer, R., 2022. Towards sensor-based phenotyping of physical barriers of grapes to improve resilience of *Botrytis* bunch rot. Front. Plant Sci. 12, 1-17.

Herzog, K., Wind, R. & Töpfer, R., 2015. Impedance of the grape berry cuticle as novel phenotypic trait to estimate resistance to *Botrytis cinerea*. Sensors 15, 12498-12512.

Hill, G.N., Evans, K.J., Beresford, R.M., & Dambergs, R.G., 2013. Near and mid-infrared spectroscopy for the quantification of botrytis bunch rot in white wine grapes. J. Near Infrared Spectrosc. 21(6), 467-475.

Hill, G.N., Evans, K.J., Beresford, R.M., & Dambergs, R.G., 2014. Comparison of methods for the quantification of botrytis rot in white wine grapes. Aust. J. Grape Wine Res. 20(3), 432-441.

Hill, G.N., Jackson, P., Sharp, J.M., Hunt, A.G. & Lewis, K.S.J., 2019. Investigating time and economic costs of botrytis bunch rot sampling using interpolated data. New Zeal. Plant Prot. 72, 166-175.

International Agency for Research on Cancer, 1993. IARC Monographs on the identification of carcinogenic hazards to humans. https://monographs. iarc.who.int/list-of-classifications Date of access: 15.11.2022.

Ioriatti, C., Guzzon, R., Anfora, G., Ghidoni, F., Maxxoni, V., Villegas, T.R., Dalton, D.T. & Walton, V.M., 2018. *Drosophila suzukii* (Diptera: Drosophilidae) contributes to the development of sour rot in grape. J. Econ. Entomol. 111(1), 283-292.

Ioriatti, C., Walton, V., Dalton, D., Anfora, G., Grassi, A., Maistri, S. & Mazzoni, V., 2015. *Drosophila suzukii* (Diptera: Drosophilidae) and its potential impact on wine grapes during harvest in two cool climate wine grape production regions. J. Econ. Entomol. 108(3), 1148-1155.

Jackson, R.S., 2014. Botrytis. Encycl. Food Microbiol. Second Ed. (1), 288-296. http://dx.doi.org/10.1016/B978-0-12-384730-0.00042-2

Jadhav, S.B. & Gupta, A., 2016. Studies on application of β -1,3 glucanase in the degradation of glucans produced by *Botrytis cinerea* and inhibition of fungal growth. Biocatal. Agric. Biotechnol. 7, 45-47.

Jin, X., Wu, X., Liu, X. & Liao, M., 2017. Varietal heterogeneity of textural characteristics and their relationship with phenolic ripeness of wine grapes. Sci. Hortic. 216, 205-214.

Kallitsounakis, G. & Catarino, S., 2020. An overview on botrytized wines. Cienc. e Tec. Vitivinic. 35(2), 76-106.

Kellner, N., Antal, E., Szabó, A. & Matolcsi, R., 2022. The effect of black rot on grape berry composition. Acta Aliment. 51(1), 126-133.

Ky, I., Lorrain, B., Jourdes, M., Pasquier, G., Fermaid, M., Gény, L., Rey, P., Doneche, B. & Teissedre, P.-L., 2012. Assessment of grey mould (*Botrytis cinerea*) impact on phenolic and sensory quality of Bordeaux grapes, musts and wines for two consecutive vintages. Aust. J. Grape Wine Res. 18(2), 215-226.

Landrault, N., Larronde, F., Delaunay, J.C., Castagnino, C., Vercauteren, J., Merillon, J.M., Gasc, F., Cros, G. & Teissedre, P.L., 2002. Levels of stilbene oligomers and astilbin in French varietal wines and in grapes during noble rot development. J. Agric. Food Chem. 50(7), 2046-2052.

Latorre, B.A., Elfar, K. & Ferrada, E.E., 2015. Gray mold caused by *Botrytis cinerea* limits grape production in Chile. Cien. Inv. Agr. 42(3), 305-330.

Lisek, J. & Lisek, A., 2021. Varietal response to sour rot bunch rot in polish grapevine genetic resources. Agronomy 11, 9-11.

Longbottom, M., Simos, C., Krstic, M. & Johnson, D., 2013. Grape quality assessments: a survey of current practice. Wine Vitic. J. 28(3), 33-37.

Madden, A.A., Boyden, S.D., Soriano, J.A.N., Corey, T.B., Leff, J.W., Fierer, N. & Starks, P.T., 2017. The emerging contribution of social wasps to grape rot disease ecology. PeerJ 5:e3223, 1-17.

Marcinkowska, R., Namieśnik, J. & Tobiszewski, M., 2019. Green and equitable analytical chemistry. Curr. Opin. Green Sustain. Chem. 19, 19-23.

Marois, J.J., Bledsoe, A.M., Ricker, R.W. & Bostock, R.M., 1993. Sampling for *Botrytis cinerea* in harvested grape berries. Am. J. Enol. Vitic. 44(3), 261-265.

Marois, J.J., Nelson, J.K., Morrison, J.C., Lile, L.S. & Bledsoe, A.M., 1986. The influence of berry contact within grape clusters on the development of *Botrytis cinerea* and epicuticular wax. Am. J. Enol. Vitic. 37(4), 293-296.

Meneguzzo, J., Miele, A., Rizzon, L.A. & Ayub, M.A.Z., 2008. Effect of bunch rot on the sensory characteristics of the Gewürztraminer wine. J. Int. des Sci. la Vigne du Vin 42(2), 107-11.

Molitor, D., Baus, O., Didry, Y., Junk, J., Hoffmann, L. & Beyer, M., 2020. BotRisk: simulating the annual bunch rot risk on grapevines (*Vitis vinifera* L. cv. Riesling) based on meteorological data. Int. J. Biometeorol. 1571-1582. Molitor, D., Biewers, B., Jünglen, M., Schultz, M., Clementi, P., Permesang, G., Regnery, D., Porten, M., Herzog, K., Hoffmann, L., Beyer, M. & Berkelmann-Löhnertz, B., 2018. Multi-annual comparisons demonstrate difference in bunch rot susceptibility of nine *Vitis vinifera* L. 'Riesling' clones. Vitis - J. Grapevine Res. 57(1), 17-25.

Mundy, D.C., 2008. A review of the direct and indirect effects of nitrogen on botrytis bunch rot in wine grapes. New Zeal. Plant Prot. 61, 306-310.

Musingarabwi, D.M., Nieuwoudt, H.H., Young, P.R., Eyéghé-bickong, H.A. & Vivier, M.A., 2016. A rapid qualitative and quantitative evaluation of grape berries at various stages of development using Fourier-transform infrared spectroscopy and multivariate data analysis. Food Chem. 190, 253-262.

Nutter, F.W., 1991. Disease assessment terms and concepts. Plant Dis. 75(11), 1187-1188.

OIV - International Organisation of Vine and Wine, (2021). Digital trends applied to the vine and wine sector. A comprehensive study on the digitalisation of the sector. https://www.oiv.int/js/lib/pdfjs/web/viewer. html?file=/public/medias/8593/digital-trends-applied-to-the-vine-and-wine-sector.pdf Date of access: 06.09.2022.

Ortt, R., Stolwijk, C. & Punter, M., 2020. Implementing Industry 4.0: assessing the current state. J. Manuf. Technol. Manag. 31(5), 825-836.

Paňitrur-De La Fuente, C., Valdés-Gómez, H., Roudet, J., Acevedo-Opazo, C., Verdugo-Vásquez, N., Araya-Alman, M., Lolas, M., Moreno., Y. & Fermaud, M., 2018. Classification of winegrape cultivars in Chile and France according to their susceptibility to *Botrytis cinerea* related to fruit maturity. Aust. J. Grape Wine Res. 24(2), 145-157.

Paňitrur-De La Fuente, C., Valdés-Gómez, H., Roudet, J., Verdugo-Vásquez, N., Mirabal, Y., Laurie, V.G., Goutouly, J.P., Acevedo-Opazo, C. & Fermaud, M., 2020. Vigor threshold NDVI is a key early risk indicator of botrytis bunch rot in vineyards. Oeno One 54(2), 279-297.

Petrovic, G., Aleixandre-Tudo, J.L. & Buica, A., 2020. Viability of IR spectroscopy for the accurate measurement of yeast assimilable nitrogen content of grape juice. Talanta 206, 1-7.

Pinto, L., Malfeito-Ferreira, M., Quintieri, L., Silva, A.C. & Baruzzi, F., 2019. Growth and metabolite production of a grape sour rot yeast-bacterium consortium on different carbon sources. Int. J. Food Microbiol. 296, 65-74.

Porep, J.U., Erdmann, M.E., Körzendörfer, A., Kammerer, D.R. & Carle, R., 2014. Determination of ergosterol in grape mashes for grape rot indication and further quality assessment by means of an industrial near infrared/visible (NIR/VIS) spectrometer - A feasibility study. Food Control 43, 142-149.

Porep, J.U., Mattes, A., Pour Nikfardjam, M.S., Kammerer, D.R. & Carle, R., 2015a. Implementation of an on-line near infrared/visible (NIR/VIS) spectrometer for rapid quality assessment of grapes upon receival at wineries. Aust. J. Grape Wine Res. 21(1), 69-79.

Porep, J.U., Mrugala, S., Por Nikfardjam, M.S. & Carle, R., 2015b. Online determination of ergosterol in naturally contaminated grape mashes under industrial conditions at wineries. Food Bioprocess Technol. 8(7), 1455-1464.

Price Waterhouse Cooper, 2013. The South African wine industry insights survey 2013. www.pwc.co.za/wine-insights-survey Date of access: 03.12.2022.

Provost, F. & Fawcett, T., 2013. Data science and its relationship to big data and data-driven decision making. Big Data 1(1), 51-59.

Rienth, M., Vigneron, N., Walker, R.P., Castellarin, S.D., Sweetman, C., Burbidge, C.A., Bonghi, C., Famiani, F. & Darriet, P., 2021. Modifications of grapevine berry composition induced by main viral and fungal pathogens in a climate change scenario. Front. Plant Sci. 12, 1-12. Rousseaux, S., Diguta, C.F., Radoï-Matei, F., Alexandre, H. & Guilloux-Bénatier, M., 2014. Non-*Botrytis* grape-rotting fungi responsible for earthy and moldy off-flavors and mycotoxins. Food Microbiol. 38, 104-121.

Santos, H., Augusto, C., Reis, P., Rego, C., Figueiredo, A.C. & Fortes, A.M., 2022. Volatile metabolism of wine grape Trincadeira: Impact of infection with *Botrytis cinerea*. Plants 11, 1-18.

South African Wine Industry Information and Systems (SAWIS), 2021. SA Wine Industry 2021 Statistics, 46, 1-32. https://www.sawis.co.za/info/annualpublication.php Date of access: 02.10.2022.

South African Wine Industry Information and Systems (SAWIS), 2022. Bulk wine and juice prices January - September 2022. https://www.sawis. co.za/info/priceranges_2022.php Date of access: 03.12.2022.

Schmidtke, L.M., Schwartz, L.J., Schueuermann, C. & Steel, C.C., 2019. Discrimination of *Aspergillus* spp., *Botrytis cinerea*, and *Penicillium expansum* in grape berries by ATR-FTIR spectroscopy. Am. J. Enol. Vitic. 70(1), 68-76.

Schueuermann, C., Steel, C.C., Blackman, J.W., Clark, A.C., Schwarz, L.J., Moraga, J., Collado, I.G. & Schmidtke, L.M., 2019. A GC-MS untargeted metabolomics approach for the classification of chemical differences in grape juices based on fungal pathogen. Food Chem. 270, 375-384.

Seem, R.C., 1984. Disease incidence and severity relationships. Ann. Rev. of Phytopathol. 22, 133-150.

Simsek, Z., Vaara, E., Paruchuri, S., Nadkarni, S. & Shaw, J., 2019. New ways of seeing big data. Acad. Manag. J. 62(4), 971-978.

Southey, T., 2022. Presentation: Data-driven decision making tools: the Terraclim integrated data tool applied to Chenin scenarios. 2nd Chenin blanc International Congress, 1-3 November 2022, Stellenbosch, South Africa.

Steel, C.C., Blackman, J.W. & Schmidtke, L.M., 2013. Grapevine bunch rots: Impacts on wine composition, quality and potential procedures for the removal of wine faults. J. Agric. Food Chem. 61(22), 5189-5206.

Steel, C.C., Savocchia, S. & Greer, L.A., 2016. Management of bunch rot diseases of grapes in subtropical vineyards in Australia. Acta Hortic. 1115, 265-271.

Steel, C.C., Schwartz, L.J., Qiu, Y., Schueuermann, C., Blackman, J.W., Clark, A.C. & Schmidtke, L.M., 2020. Thresholds for Botrytis bunch rot contamination of Chardonnay grapes based on the measurement of the fungal sterol, ergosterol. Aust. J. Grape Wine Res. 26(1), 79-89.

Steel, C.C *et al.*, 2018. Report: Determination of thresholds for bunch rot contamination of grapes and techniques to ameliorate associated fungal taints, CSU13-01. Charles Sturt University, Wagga Wagga, NSW 100 pp. https://www.wineaustralia.com/getmedia/ad6cee75-3330-48ba-b2a9-bcc0b2fb6700/CSU-1301-Final-Report Date of access: 15.11.2022.

Swanepoel, M., 2006. Monitoring the quality control chain from vineyard to wine: An industrial case study. Thesis, Stellenbosch University, Private Bag X1, 7602, Matieland (Stellenbosch), South Africa.

Swanepoel, M., du Toit, M. & Nieuwoudt, H.H., 2007. Optimisation of the quantification of total soluble solids, pH and titratable acidity in South African grape must using fourier transform mid-infrared spectroscopy. South African J. Enol. Vitic. 28(2), 140-149.

Tello, J. & Ibáñez, J., 2018. What do we know about grapevine bunch compactness? A state-of-the-art review. Aust. J. Grape Wine Res. 24(1), 6-23.

Ubeda, C., Hornedo-Ortega, R., Cerezo, A.B., Garcia-Parrilla, M.C. & Troncoso, A.M., 2020. Chemical hazards in grape and wine, climate change and challenges to face. Food Chem. 314, 1-10.

VanderWeide, J., Frioni, T., Ma, Z., Stoll, M., Poni, S. & Sabbatini, P., 2020. Early leaf removal as a strategy to improve ripening and lower cluster rot in cool climate (*Vitis vinifera* L.) Pinot Grigio. Am. J. Enol. Vitic. 71(1), 70-79.

Versari, A., Parpinello, G.P., Mattioli, A.U. & Galassi, S., 2008. Determination of grape quality at harvest using Fourier-transform midinfrared spectroscopy and multivariate analysis. Am. J. Enol. Vitic. 59(3), 317-322.

Vignault, A., Pascual, O, Jourdes, M., Moine, V., Fermaud, M., Roudet, J., Canals, J.M., Teissedre, P.L. & Zamora, R., 2019. Impact of enological tannins on laccase activity. OENO One 53(1), 27-38.

Wagner, C. & Esbensen, K.H., 2005. Theory of sampling: Four critical success factors before analysis. J. AOAC Int. 98(2), 275-281.

Welke, J.E., 2019. Fungal and mycotoxin problems in grape juice and wine industries. Curr. Opin. Food Sci. 29, 7-13.

Würz, D.A., Rufato, L., Bogo, A., Allebrandt, R., Pereira de Bem, B., Marcon Filho, J.L., Brighenti, A.F. & Bonin, B.F., 2020. Effects of leaf removal on grape cluster architecture and control of Botrytis bunch rot in Sauvignon Blanc grapevines in Southern Brazil. Crop Prot. 131, 1-6.

Zoecklein, B.W., Williams, J.M. & Duncan, S.E., 2000. Effect of sour rot on the composition of White Riesling (*Vitis vinifera* L.) grapes. Small Fruits Rev. 1(1), 63-77.