

Genetic parameters for methane emissions in Australian sheep measured in portable accumulation chambers in grazing and controlled environments

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ABSTRACT

Context. Genotype by environment interaction or sire re-ranking between measurements of methane emission in different environments or from using different measurement protocols can affect the efficiency of selection strategies to abate methane emission. **Aim.** This study tested the hypothesis that measurements of methane emission from grazing sheep under field conditions, where the feed intake is unknown, are genetically correlated to measurements in a controlled environment where feed intake is known. **Methods.** Data on emission of methane and carbon dioxide and uptake of oxygen were measured using portable accumulation chambers from 499 animals in a controlled environment in New South Wales and 1382 animals in a grazing environment in Western Australia were analysed. Genetic linkage between both environments was provided by 140 sires with progeny in both environments. Multi-variate animal models were used to estimate genetic parameters for the three gas traits corrected for liveweight. Genetic groups were fitted in the models to account for breed differences. Genetic correlations between the field and controlled environments for the three traits were estimated using bivariate models. **Key results.** Animals in the controlled environment had higher methane emission compared to the animals in the field environment ($37.0 \pm \text{s.d. } 9.3$ and $35.3 \pm \text{s.d. } 9.4$ for two protocols vs $12.9 \pm \text{s.d. } 5.1$ and $14.6 \pm \text{s.d. } 4.8$ mL/min for lambs and ewes ($\pm \text{s.d.}$; $P < 0.05$) but carbon dioxide emission and oxygen uptake did not significantly differ. The heritability estimates for methane emission, carbon dioxide emission and oxygen uptake were 0.15, 0.06 and 0.11 for the controlled environment and 0.17, 0.27 and 0.35 for the field environment. The repeatability for the traits in the controlled environment ranged from 0.51 to 0.59 and from 0.24 to 0.38 in the field environment. Genetic correlations were high (0.85–0.99) but with high standard errors. **Conclusion.** Methane emission phenotypes measured using portable accumulation chambers in grazing sheep can be used in genetic evaluation to estimate breeding values for genetic improvement of emission related traits. The combined measurement protocol-environment did not lead to re-ranking of sires. **Implication.** These results suggest that both phenotypes could be used in selection for reduced methane emission in grazing sheep. However, this needs to be consolidated using a larger number of animals and sires with larger progeny groups in different environments.

Keywords: enteric emission, feed intake, grazing environment, heritability, measurement of methane emission, portable accumulation chamber, sheep, repeatability.

Introduction

Methane emission from livestock is increasingly becoming a societal concern, due to its effect on global warming (Boucher *et al.* 2009) and the financial implication for producers if a carbon price were applied to livestock products (Alcock and Hegarty 2011; Browne *et al.* 2011). Enteric emissions from livestock account for 73% of the total

agriculture emissions and grazing sheep contribute 20% of the enteric fermentation emissions in Australia (Department of the Environment and Energy 2019). Emissions are projected to increase due to an increasing demand for livestock products. Australia has dedicated substantial resources to abate enteric methane emission in livestock. Several strategies have been suggested to reduce gas emissions involving either feeding practices and dietary additives or long-term strategies such as genetic improvement (Smith *et al.* 2008; Eckard *et al.* 2010). Previous studies have reported a modest heritability for methane emission adjusted for liveweight per head in Australian sheep (Goopy *et al.* 2016; Paganoni *et al.* 2017; Robinson *et al.* 2020) and New Zealand (Pinares-Patiño *et al.* 2013; Jonker *et al.* 2018; Rowe *et al.* 2019), so that it would be possible to reduce emission by selection. For direct selection on the trait, a practical measurement protocol needs to be established to obtain phenotypic information at scale on selection candidates, or on animals in a reference population that could be used for genomic prediction of breeding value.

Different methods and measurement protocols exist to measure phenotypes on methane emission from sheep. Closed-circuit respiration chambers (Blaxter and Clapperton 1965) and portable accumulation chambers (PAC) have been used to measure emission in controlled and field conditions respectively (Goopy *et al.* 2011; Goopy *et al.* 2016; Bond *et al.* 2019; Robinson *et al.* 2020). A measurement protocol to be adopted by breeders or in large scale phenotyping of reference populations for the purpose of genetic evaluation should be cost effective and practical whilst being predictive of lifetime rate of emission of an animal (Robinson *et al.* 2015). Under similar management conditions, PAC has been recommended as an effective, low-cost way to measure methane. The method is highly correlated with phenotypes obtained from respiration chambers (Robinson *et al.* 2020). However, the correlation between the different protocols used with PACs across environments needs to be evaluated as per Robinson *et al.* (2020).

Production of methane is determined by the amount and composition of feed eaten, the feeding schedule before measurement and the digesta flow rate from the rumen (Blaxter and Clapperton 1965; Pinares-Patiño *et al.* 2003; Goopy *et al.* 2014). These variables are typically not known during unsupervised recording of methane production under field conditions. Methane production measured under pasture grazing conditions would be potentially cheap and relatively practical, however measurement of feed intake is expensive and difficult for grazing animals. Also, measuring feed intake before testing for methane emission could increase stress levels in sheep due to isolation and confinement which disrupts their feeding behaviour and hence feed intake (Llonch *et al.* 2016). This poses the question whether methane measurement under grazing conditions could be a valid assessment of variation in methane production.

Genetic correlations between PAC measurement of methane emission under an environment with known feed intake and measurements under grazing conditions are important to establish whether these traits are genetically similar and whether sire rankings remain the same between the two measurement conditions. In addition to measuring methane emission, PACs can be used to simultaneously measure carbon dioxide production and oxygen uptake. Carbon dioxide and oxygen measured in PAC have been reported to be more heritable and have a higher genetic correlation with feed intake than methane emission for sheep at post-weaning, hogget and adult ages (Paganoni *et al.* 2017). These traits can consequently be cheaply measured in sheep using PAC and could be used as a proxy for feed intake. The objective of the study was to compare genetic parameters for methane emission, carbon dioxide emission and oxygen uptake measured using PACs in multiple breeds under different feeding protocols and environments, with controlled feed intake in New South Wales (NSW) vs field measurement with unknown feed intake in Western Australia (WA). Genetic correlations for the three traits were estimated between the two locations to evaluate whether sires' estimated breeding values for these traits are re-ranked between the two feeding protocols and environments.

Materials and methods

Data

All protocols were approved by the University of New England Animal Ethics Committee (Approval AEC 15-021) and the Department of Primary Industries and Regional Development of Western Australia (WA) (Approval AEC 4-14-10). The climatic conditions, diet and measurement protocols, however, varied as described in detail below. The main difference in the measurement protocols was that in New South Wales (NSW), gas measurements were taken using PACs in a controlled environment where feed intake was known. In WA, the animals were measured under grazing conditions and the amount of feed intake was not known. In both locations, the liveweight of the animals was measured. The PACs and devices used to measure the concentration of methane, carbon dioxide and oxygen were as described by Goopy *et al.* (2016).

Controlled environment

In NSW, a total of 510 ewes from the Sheep Cooperative Research Centre Information Nucleus flock (van der Werf *et al.* 2010) kept at the University of New England's Kirby Research station, in Armidale (30°28'S, 151°40'E), Australia, were measured in seven batches (groups of animals measured around the same time) between April 2015 and March 2016. The ewes were born between 2007 and 2013 so they were

aged 2–8 years of age at the time of measurement. Armidale has a temperate to cool climate with warm summers and summer dominant rainfall. Ewes were tested multiple times using different measurement protocols. Animals were either measured immediately after coming off feed (PAC0) or 1 h after coming off feed (PAC1). Prior to the commencement of the trial ewes were first placed in individual pens, and habituated for at least 1 week for the PAC0 measurement, and at least 3 weeks for the PAC1 measurement. The diet included equal parts of lucerne chaff and cereal hay (9.6 MJ metabolisable energy (ME) per kg dry matter (DM), 33.3% acid detergent fibre, 52.2% neutral detergent fibre and 140 g crude protein per kg DM) at 1.5 (Batches 1 and 2) or 1.6 (Batches 3–7) times their maintenance requirement, calculated from the weight of the animals before transport (Robinson *et al.* 2020). Feed was provided daily at 08:00 hours. Repeat measurements for both PAC0 and PAC1 were spaced 2 weeks apart and methane, carbon dioxide and oxygen were measured as described in Oddy *et al.* (2019). In some instances, ewes were measured more than twice. Repeat measures of 58 animals were removed because they were recorded multiple times on the same day. After removing outlier phenotypes, records from 499 animals were retained. Preparatory analysis of the data established that the phenotypic correlation between PAC0 and PAC1 measurement protocols ranged from 0.87 to 1.00, therefore measurements by the two protocols were deemed to describe the same trait and data were combined for further investigation.

Grazing environment

The WA data included measurements of 1385 ewes and ewe lambs from the Sheep Cooperative Research Centre Information Nucleus flock at the Great Southern Agricultural Research Institute, Department of Agriculture and Food at Katanning (33°41'S, 117°35'E). The climate at Katanning is Mediterranean with mild winters and hot dry summers with 70–80% of the rainfall occurring between May and October. Animals were measured in six batches with three cohorts (Lambs, Ewe 1 and Ewe 2) of animals measured twice between 2014 and 2015. Ewe 1 (290) and Ewe 2 (464) cohorts included ewes born between 2007 and 2013, while Lambs (249) were female offspring of Ewe 1 and 2 born in 2014. All lambs were born between April and June 2014 and weaned mid-October while the ewes were born between July and August of their respective birth years. Each cohort was measured twice, approximately 1 month apart. To provide some linkage between measurements some animals were measured repeatedly across days and cohorts. Each cohort of sheep grazed (dry pasture 7.0–7.7 MJ ME/kg of DM, 43.5% acid detergent fibre and 72.9% neutral detergent fibre) together for 1 month prior to testing and were supplemented if there was insufficient pasture for liveweight maintenance. Lambs were supplemented with 100 g of lupins (14.1 MJ ME per kg DM, 21.3% acid

detergent, 34.7% neutral detergent fibre and 317 g crude protein per kg DM) per day between the first and second measurement. The Ewe 1 cohort was supplementary fed with hay (8.2 MJ of ME per kg DM, 34.1% acid detergent fibre and 64.9% neutral detergent fibre and 56 g crude protein per kg DM) in February and March 2015 while the Ewe 2 cohort sheep grazed with 1500 kg DM per hectare of spring pasture and were not supplementary fed.

On the day of measurement, ewes were tested in up to 10 runs per day with up to 24 ewes per run. Animals were mustered into a holding paddock the day prior to measurement, weighed at 06:00 hours on the day of measurement and then draughted into two groups, one measured in the morning and the other in the afternoon. The afternoon group was returned to pasture before measurement. Ewes were off-pasture for no longer than 4 h, but at least 1 hour prior to methane measurement. A few ewes were measured more than twice and repeat measurements of animals that were measured multiple times on the same day were removed.

The first cohort of sheep measured were the lambs in November and December of 2014 (6–8 weeks post-weaning). Each day between 6 and 8 runs with 16–20 lambs per run were conducted. Measurement time ranged from 33 to 64 min. The Ewe 1 cohort was a mixture of pregnant and dry ewes measured in February and March 2015. Of the total of 594 ewes that were joined prior to measuring methane, 327 ewes were pregnant (first trimester) with singles, 167 with twins and 100 were dry. Each day ewes were tested in 2–9 runs with 20–24 animals per run. Most ewes were measured twice across the two measurement periods, with a small number that was only measured once or 3–5 times. The measurement time for the ewes ranged between 25 and 74 min. The Ewe 2 cohort had dry ewes measured in September and October of 2015. Each day 7–10 runs were conducted with 20–24 animals in each run with the measurement time ranging between 37 and 44 min. After removing phenotypic outliers across the whole dataset, records from 1382 animals were retained.

Statistical analyses

A summary of the data is provided in Table 1. The systematic effects used to model the three traits are described in Table 2. The fixed effects included: PAC protocol used (PAC0 or PAC1), batch, PAC chamber, birth year, test day and time of the day in the controlled environment and batch, PAC chamber, birth year, test day, time of the day, position of the PAC chamber in the experimental layout and pregnancy status in the grazing environment data. Only the significant fixed effects (described above) were fitted in the final model. Liveweight was fitted as a quadratic covariate in both flocks to account for the non-linear relationship between the liveweight and the three gas measurement traits as a

Table 1. Data structure, means and standard deviations (in brackets) for methane emission (CH₄), carbon dioxide emission (CO₂) and oxygen uptake (O₂) for sheep tested by portable accumulation chamber in controlled environment for two protocols (PAC0 and PAC1) and grazing environments for lambs and ewes.

Variable	Controlled environment		Grazing environment	
	PAC0	PAC1	Lambs	Ewes
Records	520	553	772	2137
Animals	488	489	368	1014
Sires	176	176	3	282
Dams	449	451	344	843
CH ₄ (mL/min)	37.0 (9.3)	35.3 (9.4)	12.9 (5.1)	14.6 (4.8)
CO ₂ (mL/min)	447.0 (83.7)	396.0 (71.3)	299.0 (60.8)	421.0 (82.2)
O ₂ (mL/min)	-468.0 (75.7)	-436.0 (75.3)	-310.0 (57.8)	-375.0 (88.4)
LWT (kg)	50.1 (6.8)	50.2 (6.8)	33.4 (5.2)	60.0 (7.9)

Variables: CH₄, methane; CO₂, carbon dioxide; O₂, oxygen; LWT, liveweight; mL/min, millilitres per minute. To convert mL/min into grams/day use: CH₄ (g/day) = 1.03114 CH₄ (mL/min); CO₂ (g/day) = 2.82915 CO₂ (mL/min); O₂ (g/day) = 2.05701 O₂ (mL/min). Controlled environment: PAC0, feed was available until the time of testing; PAC1, feed was withdrawn 1 h before testing.

Table 2. Environmental effects with a significant effect on measurements for emission methane, carbon dioxide emission and oxygen uptake in the controlled and grazing environments.

Model term	Description
Method	In the controlled environment, two test protocols were used: PAC0 – feed was not removed prior to measurement; PAC1 – feed was removed 1 h before measurement
Chamber	12 portable accumulation chambers for PAC0 and PAC1 in the controlled environment and 29 in the grazing environment
Chamber position	24 levels indicating the position of the PAC during measurement in grazing environment
Test batch	Seven and five cohorts of animals brought from the field for measurement in the controlled and grazing environments, respectively
Test-day	In the controlled environment, PAC0 measurements were conducted in 2 days while PAC1 measurements were conducted in 4 days In the grazing environment, measurements were conducted in 21 days
Test-time	PAC0 had four test runs: 09:30 h, 11:00 h, 12:30 h, 14:00 h; PAC1: 09:30 h, 11:00 h or 11:30 h, 13:00 h or 13:30 h (fitted separately for each method). In the controlled environment, 7–8 runs were made on different test days with varied times based on the cohorts and runs conducted in a day ranging from 07:00 h to 16:30 h
Birth year	2007–2013 in the controlled environment and 2007–2014 in grazing environment
Liveweight	Liveweight recorded soon after arrival, used to calculate feed, offered at 1.5 (Batches 1 and 2) or 1.6 (Batches 3–7) times the maintenance requirements in the controlled environment Liveweight recorded during measurement in the grazing environment
Pregnancy status	Dry, single bearing and twin bearing in the grazing environment
Genetic groups	Terminal, maternal and merino in both the controlled and grazing environments

PAC0, feed was available until the time of testing; PAC1, feed was withdrawn 1 h before testing.

proxy for feed intake. An interaction term between test day and time of the day was also fitted in both controlled and grazing environment datasets. The effect of rear type and the interaction term between age and liveweight were not significant for any of the gas traits and therefore, were not included in the model. Random effects for all the traits under the controlled and grazing environments included animal genetic effect to estimate the additive genetic variance and permanent environmental effect of the animal to account for correlations between repeated records, along with residual error. The amount of variation explained by year of birth, batch, chamber, date of measurement, PAC

method, position of the chamber and pregnancy status (Table 3) were subsequently estimated by fitting these effects as random instead of fixed as described above. These estimates on the systematic environmental effects are important to inform the control of these effects in future experiments.

The statistical analyses was carried out in ASReml (Gilmour et al. 2015) to estimate (co)variances for methane emission, carbon dioxide emission and oxygen uptake. The pedigree for the animals in the controlled environment included 2649 animals over 18 generations, with 1012 sires, 1123 dams, 556 sires of sires, 462 sires of dams, 398

Table 3. Variation explained by different environmental factors in measurements in the controlled environment and grazing environment as a proportion of the total variance.

Model terms	Controlled environment			Grazing environment		
	CH ₄	CO ₂	O ₂	CH ₄	CO ₂	O ₂
Year of birth	0 ± 0	0.07 ± 0.09	0.13 ± 0.13	0.01 ± 0.01	0.05 ± 0.03	0.02 ± 0.01
Batch	0.10 ± 0.06	0.08 ± 0.07	0.22 ± 0.14	0.38 ± 0.18	0.43 ± 0.18	0.58 ± 0.17
Chamber	0 ± 0	0.02 ± 0.01	0.01 ± 0.01	0 ± 0	0 ± 0	0 ± 0
Date	0.05 ± 0.02	0.07 ± 0.04	0.10 ± 0.04	0.17 ± 0.17	0.14 ± 0.07	0.11 ± 0.06
Date time of day	0.05 ± 0.02	0.06 ± 0.03	0.04 ± 0.01	0.05 ± 0.02	0.09 ± 0.03	0.08 ± 0.03
Method	0.04 ± 0.06	0.29 ± 0.3	0.09 ± 0.13			
Chamber position				0 ± 0	0.02 ± 0.01	0 ± 0
Pregnancy status				0.01 ± 0.04	0 ± 0	0 ± 0

Blank spaces mean the effect was not present or recorded in the location.
CH₄, methane; CO₂, carbon dioxide; O₂, oxygen.

dams of sires, 276 dams of dams and three genetic groups. To account for variation in breeds the base animals were classified into maternal, terminal and Merino genetic groups. The sire breeds included pure-bred Merino, Border Leicester, Dorset Horn, Dorset, Texel, White Suffolk, Dohne, composite maternal whereas the dams were either pure-bred Merino or a Border Leicester × Merino cross. In the grazing environment dataset, the pedigree included 4396 animals over 19 generations, with 1317 sires, 2002 dams, 646 sires of sires, 752 sires of dams, 469 dams of sires, 662 dams of dams and three genetic groups as described above. The gas measurements (y) from animal j were modelled as:

$$y = Xb + Z_1u + Z_1Qg + Z_2pe + e,$$

where y is the vector of trait observations for methane emission, carbon dioxide emission and oxygen uptake; b is the fixed effect vector; u is the vector of random animal effects; g is the vector of fixed genetic group effects; pe is the vector containing random permanent environmental effects of animal and e the vector of random residual effects. The matrices X , Z_1 and Z_2 are incidence matrices that link observations to levels of fixed effects, additive genetic effects and permanent environmental effects, respectively. Q is a matrix of genetic groups allocating animals with unknown pedigree to either maternal, terminal or Merino populations. Trivariate models between the three gas traits were also used to estimate correlations between the gas traits within the controlled and grazing environments. Genetic correlations for methane emission, carbon dioxide emission and oxygen uptake between the controlled and grazing environments were estimated using both animal and sire models while fitting the significant fixed effects within each environment and genetic groups as described above with the residual covariance set to zero.

A substantial number of sires (140) provided a genetic link between the two environments.

Results

Methane emission, carbon dioxide emission and oxygen uptake were on average higher for sheep in a controlled environment than under grazing conditions (Table 1). Ewes in the controlled environment were on average 50.2 kg (34.4–74.5) while in the grazing environment the ewes were 60 kg (36.0–90.2) and the lambs 33.4 kg (17.6–66.5). In the controlled environments, sheep consumed on average 1.4 kg of feed on the day before measurement and 0.9 kg on the day of measurement. The genetic group solutions and liveweight coefficients for the three traits in the controlled and grazing environments are shown in Table 4. The genetic group solutions for the three gas traits were not significantly different within the environments. For both environments, positive linear coefficients were estimated for liveweight on methane and carbon dioxide emission and negative for oxygen uptake. The quadratic coefficients in the controlled environment were not significantly different from zero. In the grazing environment, quadratic terms for methane emission (−2.0) and carbon dioxide emission (−48.2) were negative whereas they were positive for oxygen uptake (43.4).

Overall, there were larger phenotypic variances for all three gas traits in the controlled than in the grazing environment as shown in Table 5. The estimated genetic variances and heritabilities for the three gas traits were not significantly different from zero in the controlled environment. The heritabilities for methane emission, carbon dioxide emission and oxygen uptake in the grazing environment were $0.17 \pm \text{s.e } 0.05$, $0.26 \pm \text{s.e } 0.05$ and $0.32 \pm \text{s.e } 0.05$, respectively. Methane emission ($0.59 \pm \text{s.e } 0.03$ vs $0.23 \pm \text{s.e } 0.03$), carbon dioxide emission ($0.51 \pm \text{s.e } 0.03$ vs $0.35 \pm \text{s.e } 0.03$),

Table 4. Genetic group solutions and liveweight regression coefficients for methane emission, carbon dioxide emission and oxygen uptake in controlled environment (CE) and grazing environment (GE).

Model terms	Environment	Levels	Trait		
			CH ₄ ± s.e	CO ₂ ± s.e	O ₂ ± s.e
Genetic groups (mL/min)	CE	Maternal	38.0 ± 2.88	353.4 ± 51.93	-343.8 ± 54.07
		Terminal	43.8 ± 14.20	352.2 ± 11.32	-334.8 ± 11.80
		Merino	34.2 ± 2.25	355.9 ± 8.61	-340.5 ± 8.93
	GE	Maternal	7.3 ± 3.89	327.7 ± 50.73	-313.5 ± 50.17
		Terminal	6.3 ± 0.97	333.6 ± 12.61	-344.6 ± 12.41
		Merino	6.7 ± 0.82	335.5 ± 10.45	-348.8 ± 10.18
Liveweight (mL/min/kg)	CE	Linear	8.9 ± 0.92	93.5 ± 6.34	-81.4 ± 5.25
		Quadratic	1.1 ± 1.22	-5.1 ± 8.47	6.9 ± 6.98
	GE	Linear	8.0 ± 0.34	151.3 ± 4.46	-155.6 ± 54.35
		Quadratic	-2.0 ± 0.35	-48.2 ± 4.58	43.4 ± 4.38

CE, controlled environment and protocol (NSW); GE, grazing environment and protocol (WA); CH₄, methane; CO₂, carbon dioxide; O₂, oxygen.

0.03) and oxygen uptake ($0.53 \pm \text{s.e } 0.05$ vs $0.38 \pm \text{s.e } 0.02$) were more repeatable in the controlled compared to the grazing environment. The estimates of the genetic correlations between methane and the other traits in the controlled environment were not reliable due to the high standard errors. However, the phenotypic correlations between the three traits followed the same trend in both environments but correlations were generally higher in the controlled environment. Positive phenotypic correlations were estimated between methane emission and carbon dioxide ($0.52 \pm \text{s.e } 0.02$ and $0.73 \pm \text{s.e } 0.02$). Both methane emission ($-0.41 \pm \text{s.e } 0.02$ and $-0.67 \pm \text{s.e } 0.02$) and carbon dioxide emission ($-0.83 \pm \text{s.e } 0.01$ and $-0.92 \pm \text{s.e } 0.01$) had negative phenotypic correlations with oxygen uptake. In the grazing environment, high positive genetic correlations were found between methane emission and carbon dioxide emission ($0.77 \pm \text{s.e } 0.09$) whereas strong negative genetic correlations were estimated between

methane emission and oxygen uptake ($-0.67 \pm \text{s.e } 0.11$) and between carbon dioxide emission and oxygen uptake ($-0.99 \pm \text{s.e } 0.01$).

Apart from the additive and permanent environmental variance, the effects of batch, date of measurement and the interaction between date and time of day explained more variation than the other effects. Although chamber, year of birth, pregnancy status and the position of the PAC chamber were significant, they explained the lowest proportion of the phenotypic variance for all the three gas traits in both environments. Batch, date and time of measurement was associated with higher variation in the grazing compared to controlled environment.

The animal and sire models estimated similar values for the genetic correlations as shown in Table 6. High genetic correlations were estimated for methane emission, carbon dioxide emission and oxygen uptake estimated between the controlled and grazing environments. However, these

Table 5. Variance components, heritabilities (bold diagonals), repeatabilities (bold diagonals in brackets), genetic correlations (above diagonal) and phenotypic correlations (below diagonals) for methane emission, carbon dioxide emission and oxygen uptake in the controlled environment and grazing environment from a multi-trait model.

Environment	Trait	Variance components			Correlations and variance ratios		
		Additive ± s.e	PE ± s.e	Phenotypic ± s.e	CH ₄ ± s.e	CO ₂ ± s.e	O ₂ ± s.e
CE	CH ₄	8.5 ± 8.3	26.5 ± 8.2	59.6 ± 2.6	0.14 ± 0.14 (0.59 ± 0.03)	0.18 ± 0.78	-0.06 ± 0.76
	CO ₂	337.7 ± 402.0	1195.7 ± 401.2	3025.5 ± 121.7	0.73 ± 0.02	0.11 ± 0.13 (0.51 ± 0.03)	-0.99 ± 0.06
	O ₂	297.6 ± 278.2	775.6 ± 276.0	2020.6 ± 83.1	-0.67 ± 0.02	-0.92 ± 0.01	0.15 ± 0.14 (0.53 ± 0.03)
GE	CH ₄	1.5 ± 0.4	0.5 ± 0.4	8.7 ± 0.3	0.17 ± 0.05 (0.23 ± 0.03)	0.77 ± 0.09	-0.67 ± 0.11
	CO ₂	389.0 ± 80.4	136.4 ± 70.7	1514.0 ± 46.4	0.52 ± 0.02	0.26 ± 0.05 (0.35 ± 0.03)	-0.99 ± 0.01
	O ₂	461.4 ± 84.0	91.1 ± 70.7	1445.9 ± 45.6	-0.41 ± 0.02	-0.83 ± 0.01	0.32 ± 0.05 (0.38 ± 0.02)

CE, controlled environment and protocol (NSW); GE, grazing environment and protocol (WA); CH₄, methane; CO₂, carbon dioxide; O₂, oxygen; PE, permanent environment.

Table 6. Genetic correlations between methane emission, carbon dioxide emission and oxygen uptake measured in sheep using the portable accumulation chambers in the controlled environment and grazing environment using an animal and sire model.

Trait	Animal model \pm s.e	Sire model \pm s.e
CH ₄	0.85 \pm 0.79	0.85 \pm 0.65
CO ₂	0.93 \pm 0.79	0.97 \pm 0.81
O ₂	0.99 \pm 0.59	0.98 \pm 0.74

CH₄, methane; CO₂, carbon dioxide; O₂, oxygen.

correlations should be interpreted with caution due to the high standard errors. Common sires had on average 2.68 (a range of 1–9) and 3.75 (a range of 1–13) progeny in the controlled and grazing environments, respectively. Although they are not shown here the additive, permanent environment and residual variances were similar to what was estimated using the trivariate models in this study.

Discussion

This study estimated genetic parameters for methane emission, carbon dioxide emission and oxygen uptake traits measured using PACs in a field grazing environment and the genetic association with measurements in a controlled environment where feed intake was known and controlled. Although the dataset used in this study was not large, the results indicate that PAC measurements for methane emission from sheep under grazing and controlled environments are heritable, and suggest that the three traits are genetically correlated. There would appear to be potential for commercial industry application of PAC measurements on grazing sheep for genetic evaluations to abate methane emission through selection. In the design of this study the measurement protocols were confounded with location and Armidale and Katanning constitute significant environmental differences in terms of temperature, humidity and feed availability. However, the high genetic correlations (acknowledging the high standard errors) between the two recording environments are reassuring in the sense that neither the natural environment nor the measurement protocol result in a large re-ranking of breeding values for emissions traits. This gives some confidence that PAC measurement of emission traits are relatively robust across such protocols as well as environment.

The methane emissions quantified in the grazing environment (14.1 mL/min) were similar to those reported for grazing sheep in New Zealand using PACs (12.6 mL/min) (Jonker *et al.* 2018). Methane emission in the controlled environment (36.1 mL/min) were also the same as those measured by Robinson *et al.* (2020) and Dominik *et al.* (2017) using PACs. The lower methane emissions in the grazing environment are principally due to decreased feed

intake during the period immediately prior to measurement. This is accounted for in part by the fixed effects of test day and test time. The effect of disruption to feed intake has been observed in a controlled environment as higher phenotypic variance for feed intake on the day of measurement compared to the day before measurement (Robinson *et al.* 2020). In addition, animals remained off feed for at least 1–4 h on the test day in the grazing environment while the ewes in the controlled environment were habituated prior to measurement with a fixed amount of feed based on their maintenance requirements derived from their liveweight. In addition, a lower ME concentration in the diet was available for the animals in the grazing environment compared to the controlled environment. The average ewe liveweight was not significantly different between environments and, therefore, ewes' carbon dioxide emission and oxygen uptake measurements were not significantly different between the grazing and controlled environments' measurements. However, in the grazing environment, lambs were lighter compared to ewes and therefore had lower but not significantly different carbon dioxide emission and oxygen intake. Records on these systematic effects that influence the phenotypes are therefore important to ensure unbiased genetic evaluation.

In both the controlled and grazing environments, a kilogram increase in liveweight can be associated with a higher methane and carbon dioxide emission. A non-linear relationship however, exists between the three gas traits and liveweight based on the quadratic regression coefficients estimated in this study. We hypothesise that the negative quadratic coefficients in the grazing environment could be due to a reduction in feed intake per unit weight. The value of using liveweight to account for feed intake needs to be explored further since it does not account for all the variation due to feed intake. A model fitting liveweight and feed intake on the day of measurement for the controlled environment (results not presented) reduced the phenotypic variance for methane emission by 60%. After accounting for liveweight and feed intake the remaining variation still has a heritable component ($h^2 = 0.12 \pm$ s.e 0.11). Carbon dioxide emission has also been recommended as a proxy for feed intake and feed efficiency in beef cattle (Herd *et al.* 2016; Arthur *et al.* 2018; Renand *et al.* 2019). Fitting liveweight and carbon dioxide as covariates in the grazing environment, reduced the phenotypic variance of methane emission by 53% and 27% for the controlled and grazing environments respectively. This indicates that in the absence of information on feed intake an alternative would be to use both liveweight and carbon dioxide.

The estimated genetic group solutions were not significantly different for all the three traits suggesting that there are no significant differences in methane emission, carbon dioxide emission and oxygen uptake between maternal, terminal and Merino breeds. Since the genetic group solutions were not significantly different producers and breeders can

focus on genetic selection within breed to reduce methane emission.

Heritability and repeatability

This study builds onto the existing research findings that PAC measurements on methane emission in the grazing and controlled environments are heritable (Robinson *et al.* 2014, 2020; Goopy *et al.* 2016; Jonker *et al.* 2018). Selection strategies can therefore utilise the available genetic variation to reduce methane emission from grazing sheep in Australia in the longer term. The heritabilities of carbon dioxide emission (0.01–0.22) and oxygen uptake (0.08–0.38) estimated in this study are within the range of what is reported in other studies (Paganoni *et al.* 2017). It was difficult to accurately separate the animal additive and permanent environmental variance (especially in the controlled environment) with the current small data set, and so, reflecting the high standard errors, these heritability estimates should be interpreted with caution.

The shorter interval between the repeated measurements in the controlled environment led to significantly higher repeatabilities compared to a month interval in the grazing environment. Elsewhere, a repeatability of 0.55 has been reported for methane emission measure in respiration chambers with a 10–15-day interval and as high as 0.94 for measurements on consecutive days (Pinares-Patiño *et al.* 2013). The moderate repeatabilities also align with what was estimated for Merino ewes in a controlled environment (0.17–0.40) (Dominik and Oddy 2015), lambs and ewes grazing pastures (0.33–0.55) in New Zealand (Jonker *et al.* 2018), ewe lambs measured for 17 days in a controlled environment (0.36) in Ireland (O'Connor *et al.* 2021) and the phenotypic correlation (0.59) between two PAC measurement protocols in a controlled environment in Australia (Robinson *et al.* 2020). Moderate repeatabilities for methane emission (0.23 and 0.59), carbon dioxide emission (0.34 and 0.51) and oxygen uptake (0.38 and 0.53) in both environments suggest that there is benefit in having repeated measurements for genetic evaluation to rank animals for selection. In fact, the heritability based on n measurements is equal to $\text{heritability}/[\text{repeatability} + (1 - \text{repeatability})/n]$, based on two measurements, the heritability would increase by 33% for repeatability = 0.5 whereas the increase would be only 5% for repeatability = 0.9. However, for the purpose of building reference populations for genomic selection, it is more beneficial to measure more animals rather than multiple measurements per animal. Additionally, an optimal measurement protocol also needs to be evaluated to find a balance between cost and accuracy for routine genetic evaluation of methane emission to aid in selection and genetic improvement. This is in consideration of the high cost involved in recording the gas phenotypes.

Correlations within the environments

Positive genetic and phenotypic correlations between methane and carbon dioxide emission show that sheep with high methane emission also emit higher levels of carbon dioxide and consume more oxygen. These correlations are consistent with previous estimates for Australian sheep (Paganoni *et al.* 2017). Correlated response in methane emission could also be achieved by selecting for animals with a genetic potential for lower carbon dioxide emission. Since a positive genetic correlation is reported between carbon dioxide emission and feed intake, inclusion of the emission traits and production traits in the selection index may improve feed efficiency without compromising productivity (Paganoni *et al.* 2017; Robinson *et al.* 2020). High genetic correlations between PAC methane emission and feed intake (86–95%) also suggest that reducing methane and feed intake relative to production or liveweight is more effective than reducing methane relative to feed intake to reduce methane emission without compromising on production (Robinson *et al.* 2020). The high genetic correlations between carbon dioxide emission and oxygen uptake (–0.99 and –0.82) indicate that the two traits are the same genetically. Despite the high standard errors, the negative correlations were expected since carbon dioxide is exhaled (positive values) from the inhaled oxygen (negative values) from the lungs.

Correlations between the environments

In this study, strong genetic correlations (0.85–0.99) were estimated between methane emission, carbon dioxide emission and oxygen uptake in sheep measured in different locations and using different measurement protocols (field and controlled). This provides an indication that selection for methane emission, carbon dioxide emission and oxygen uptake could lead to a similar result, whether based on PAC measurements in a controlled environment or in a grazing environment. However, the genetic evaluation models need to account for possible differences in mean and variance that may result from differences in management and measurement protocols in the separate environments. The trait measurements should also be expressed in the same units and link sires included in the analysis. The slightly lower genetic correlations between methane emission in both environments show that methane (0.85) is more sensitive to the measurement protocol and environment compared to carbon dioxide emission (0.93–0.97) and oxygen uptake (0.98–0.99). These results need to be verified using a large dataset before implementation for routine genetic evaluation due to the high standard errors. Future studies should also consider evaluating the genetic correlation between grazing environment methane emission in multiple environments because the measurement protocols were confounded with the environment in this study.

Conclusions

Measurements of methane emission, carbon dioxide emission and oxygen uptake in the field using PACs can be used for genetic improvement. Although the values were associated with high standard errors, positive and favourable genetic correlations were estimated between methane emission, carbon dioxide emission and oxygen uptake measured in PACs in field and controlled environments. Methane emission phenotypes measured using PACs in a field environment and in controlled environments can therefore be used in genetic evaluation to estimate breeding values for genetic improvement of emission related traits.

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