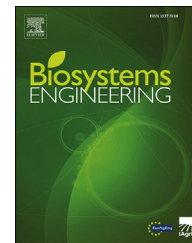




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## Research Paper

# Thermal environment sensor array: Part 1 development and field performance assessment

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Current thermal environment (TE) monitoring and control strategies for livestock and poultry facilities require enhanced measurement capabilities to provide an optimum TE based on the animals' thermal demands. Further, techniques for combining additional parameters are needed to adequately assess the total impact of the TE on the animals. Hence, two papers introduce a spatial network of 44 Thermal Environment Sensor Arrays (TESAs), each with a custom data acquisition system (Part 1) and a technique for evaluating the TE as a function of mean body temperature difference from thermally comfortable pigs using estimated body mass and TESA measurements as inputs (Part 2). The TESAs and new thermal index were deployed in a commercial pig facility to perform a preliminary assessment of robustness and capabilities under production settings. Each TESA measured dry-bulb temperature ( $T_{db}$ ), black globe temperature, airspeed, and relative humidity (RH), and required a custom circuit board with a microcontroller, signal conditioning, and communication hardware. After closeout (completion of the production cycle), TESAs were validated with a reference system to determine individual time constants and assess if a significant bias correction was needed (except airspeed). Total number of usable measurements for subsequent analysis for all sensors per TESA averaged (95% CI) 202,310 (199,187; 205,437). In summary, 7%  $T_{db}$  thermistor, 9% digital  $T_{db}$ , and 27% RH sensors required correction after 170 d inside the facility. Utilisation of low-cost sensors, open-source software, and microcontrollers allowed this novel network to provide sufficient measurement density to promote future queries on TE data in animal facilities.

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## 1. Introduction

The growing global population is projected to increase by 2.4 billion people from 2015 to 2050 (UN, 2015) and will require a

secure animal-based protein supply raised in energy, water, and feed efficient housing systems that do not adversely impact the environment. A housing system operating within the animal's optimum Thermal Environment (TE) is one approach to enhance animal well-being and growth

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Nomenclature			
ADC	Analogue to Digital Converter	$x'_{TESA}$	bias corrected future measure value ( $^{\circ}\text{C}$ or % RH)
BGT	Black Globe Thermometer	$x_{TESA}$	future measured value ( $^{\circ}\text{C}$ or % RH)
CDMS	Custom Data Management Software	$Z_{calc}$	z-statistic
CI	Confidence Interval	$\Delta R_{ref}$	divider resistor standard uncertainty (reference; $\Omega$ )
CTA	Constant Temperature Anemometer	$\Delta^{\circ}\text{RH}$	relative humidity zeroth-order standard uncertainty
DAQ	Data Acquisition	$\Delta^{\circ}\text{RH}_{ref}$	reference RH zeroth-order standard uncertainty
MTRHR	Mobile Temperature and Relative Humidity Reference	$\Delta^{\circ}\text{T}_{db}$	dry-bulb temperature zeroth-order standard uncertainty (analogue)
$n$	number of steady-state measurements	$\Delta^{\circ}\text{T}_{db,d}$	dry-bulb temperature zeroth-order standard uncertainty (digital)
NTC	Negative Temperature Coefficient	$\Delta^{\circ}\text{T}_g$	globe temperature zeroth-order standard uncertainty
OTA	Omnidirectional Thermal Anemometer	$\Delta^{\circ}\text{T}_{ref}$	reference $\text{T}_{db}$ zeroth-order standard uncertainty
PCB	Printed Circuit Board	$\Delta^{\circ}\text{u}$	airspeed zeroth-order standard uncertainty
RH	Relative Humidity (%)	$\Delta V_{ref}$	analogue voltage standard uncertainty (reference; $V_{DC}$ )
$s_{ref}$	steady-state standard deviation for reference sensor ( $^{\circ}\text{C}$ or % RH)	$\Delta x$	difference between $x_0$ and $x$ at steady-state ( $^{\circ}\text{C}$ or % RH)
$s_{TESA}$	steady-state standard deviation for TESA sensor ( $^{\circ}\text{C}$ or % RH)	$\Delta x_{ref}$	single sample reference standard uncertainty ( $^{\circ}\text{C}$ or % RH)
$t$	time (s)	$\Delta \bar{x}_{ref}$	mean steady-state reference combined standard uncertainty ( $^{\circ}\text{C}$ or % RH)
$t_0$	initial time (s)	$\Delta^{\circ}x_{ref}$	zeroth-order standard uncertainty for reference sensor ( $^{\circ}\text{C}$ or % RH)
$\tau$	time constant (s)	$\Delta \bar{x}_{TESA}$	mean steady-state TESA combined standard uncertainty ( $^{\circ}\text{C}$ or % RH)
$\text{T}_{db}$	dry-bulb temperature ( $^{\circ}\text{C}$ )	$\Delta^{\circ}x_{TESA}$	zeroth-order standard uncertainty for TESA sensor ( $^{\circ}\text{C}$ or % RH)
TE	Thermal Environment	$\Delta \bar{x}_{TESA}$	mean TESA sensor steady-state value ( $^{\circ}\text{C}$ or % RH)
TESA	Thermal Environment Sensor Array		
TESA DAQ	Thermal Environment Sensor Array Data Acquisition		
$\text{T}_g$	globe temperature ( $^{\circ}\text{C}$ )		
$\text{T}_{mr}$	mean radiant temperature ( $^{\circ}\text{C}$ )		
$x(t)$	sensor response as a function of time ( $^{\circ}\text{C}$ or % RH)		
$x_0$	initial mean sensor value at time $t_0$ ( $^{\circ}\text{C}$ or % RH)		
$\bar{x}_{ref}$	mean reference steady-state value ( $^{\circ}\text{C}$ or % RH)		

performance (Curtis, 1983; Renaudeau, Gourdine & St-Pierre, 2011), while simultaneously reducing facility resource usage, as well as total feed consumed and days on feed. The TE describes the parameters that influence heat exchange (i.e., convective, conductive, radiative, and evaporative) between an animal and its surroundings (ASHRAE, 2013; Curtis, 1983; DeShazer, Hahn & Xin, 2009); however, all required parameters that describe the TE a housed animal experiences are rarely quantified, resulting in a lack of accurate TE control that is optimal for the animal. Hence, there is a need for advanced techniques to accurately assess and, ultimately, control the TE based on how the animal exchanges heat with its surroundings (Fournel, Rousseau & Laberge, 2017).

The parameters used to describe the TE include dry-bulb temperature ( $\text{T}_{db}$ ), relative humidity (RH), airspeed, and mean radiant temperature ( $\text{T}_{mr}$ ). Dry-bulb temperature is frequently the main parameter used to describe and control TE in commercial animal production systems; however, it exclusively impacts the convective (with airspeed) and evaporative (with airspeed and RH) modes of heat loss. The RH must be known with  $\text{T}_{db}$  to estimate latent heat loss (i.e., by respiration or wetted skin evaporation) by determining the water vapour pressure gradient between surrounding air and the saturated surface. Airspeed influences convective and evaporative heat transfer rates, and can substantially increase

heat loss (beneficial in a hot  $\text{T}_{db}$ ; unfavourable in a cold  $\text{T}_{db}$ ). Lastly,  $\text{T}_{mr}$  is the uniform temperature of the surroundings in which radiant heat transfer from the animal's surface equals that in the actual surroundings. Due to the instrumentation difficulties,  $\text{T}_{mr}$  and airspeed are often neglected in livestock facilities, despite Bond, Kelly and Heitman (1952), Mount (1964), Mount (1967), and Beckett (1965) having shown radiative heat losses to be a substantial source of heat loss from pigs.

The incorporation of these four parameters into a single Thermal Environment Sensor Array (TESA) that is robust and practical for application in livestock and poultry facilities would allow the integration and application of advanced techniques. For human occupied buildings, many commercially available TE measurement systems exist to quantify indoor thermal comfort statistical values (e.g., draught rate, predicted mean vote, and predicted percentage dissatisfied; ASHRAE, 2013). These systems are prohibited by cost from use in multi-point Data Acquisition (DAQ) systems, feature proprietary hardware and software that limit flexibility, and are designed for relatively clean, low airspeed environments. In animal production systems, various combinations of  $\text{T}_{db}$ , RH, airspeed, and/or  $\text{T}_{mr}$  have been monitored (Brown-Brandt et al., 2014; Hayes et al., 2013; Vilela et al., 2015), but rarely all together. There is a unique opportunity, specific to animal

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