Collective thermoregulation in bee clusters

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Swarming is an essential part of honeybee behaviour, wherein thousands of bees cling onto each other to form a dense cluster that may be exposed to the environment for several days. This cluster has the ability to maintain its core temperature actively without a central controller, and raises the question of how this is achieved. We suggest that the swarm cluster is akin to an active porous structure whose functional requirement is to adjust to outside conditions by varying its porosity to control its core temperature. Using a continuum model that takes the form of a set of advection-diffusion equations for heat transfer in a mobile porous medium, we show that the equalization of an effective “behavioural pressure”, which propagates information about the ambient temperature through variations in density, leads to effective thermoregulation. Our model extends and generalizes previous models by focusing the question of mechanism on the form and role of the behavioural pressure, and allows us to explain the vertical asymmetry of the cluster (as a consequence of buoyancy driven flows), the ability of the cluster to overpack at low ambient temperatures without breaking up at high ambient temperatures, and the relative insensitivity to large variations in the ambient temperature. Finally, our theory makes testable hypotheses for how the cluster bee density should respond to externally imposed temperature inhomogeneities, and suggests strategies for biomimetic thermoregulation.

I. INTRODUCTION

Honeybees are masters of cooperative thermoregulation, and indeed need to be able to do so for surviving during winter, raising their brood, frying predatory wasps, or during swarming. Swarming is an essential part of colony reproduction during which a fertilized queen leaves the colony with about 2,000-20,000 bees, which cling onto each other in a swarm cluster, typically hanging on a tree branch, for times up to several days, while scouts search for a new hive location \cite{1}. During this period, the swarm cluster regulates its temperature by forming a dense surface mantle that envelopes a more porous interior core. At low ambient temperatures, the cluster contracts and the mantle densifies to conserve heat and maintain its internal temperature, while at high ambient temperatures the cluster expands and the mantle becomes less dense to prevent overheating in the core. Over this period, when the swarm is in limbo before moving to a new home the cluster adjusts its shape and size to maintain the bees at a tolerable temperature, and is able to regulate the core temperature to within a few degrees of a homeostatic set point of $35^\circ$C over a wide range of ambient conditions.

The swarm cluster is able to control its core temperature without a centralized controller to coordinate behaviour in the absence of any long-range communication between bees in different parts of the cluster \cite{2}. Instead, the thermoregulation behaviour of a swarm cluster emerges from the collective behaviour of thousands of bees\textsuperscript{3} who know only their local conditions. So how can bees in a cluster, each acting on very limited local information, control a core temperature that they are ignorant of? Early work on swarm clusters, and the related problem of winter clusters \cite{3,4}, used continuum models for variations in bee density, and temperature as determined by the diffusion of heat in a metabolically active material. Most of these models \textsuperscript{6,7} assumed that the bee know their location and the size of the cluster, contrary to experimental evidence. However, a new class of models initiated by Myerscough\textsuperscript{10} are based on local information, as experimentally observed; these models are qualitatively consistent with the presence of a core and mantle, but are unable to maintain a high core temperature at low ambient temperatures. Further refinements of these models that account for bee thermotaxis and also use only local information\textsuperscript{11,12}, yield the observed mantle-core formation and good thermoregulation properties at low ambient temperatures, while allowing for an increased core temperature at very low ambient temperatures, observed in some clusters\textsuperscript{2,13,15}. However, the thermotactic mechanism that defines these models of bee behaviour causes the cluster to break up at moderate to high ambient temperatures, unlike what is observed.

Here we present a model for swarm cluster thermoregulation that results from the collective behaviour of bees acting based on local information, yet propagates information about ambient temperature throughout the cluster. Our model yields good thermoregulation and is consistent with experiments at both high and low temperatures, with a cluster radius and mantle density qualitatively similar to observations, without leading to cluster breakup. In Section 2, we outline the basic principles and assumptions behind our model. In Section 3, we formulate our model mathematically, characterize the number...
of parameters in it, and solve the governing equations using a combination of analysis and numerical simulation in Section 4, and compare our results qualitatively with observations and experiments. We conclude with a discussion in Section 5, where we suggest a few experimental tests of our theory.

II. MODEL ASSUMPTIONS

Any viable mechanism for cluster thermoregulation consistent with experimental observations should have the following features: a) The behaviour of a cluster must result from the collective behaviour of bees acting on local information, not through a centralized control mechanism. b) The bee density of a cluster must form a stable mantle-core profile. c) The cluster must expand (contract) at high (low) ambient temperatures to maintain the maximum interior “core” temperature robustly over a range of ambient temperatures.

This suggests that any quantitative model of the behaviour of swarm clusters requires knowledge of the transfer of heat through the cluster, the movement of bees within the cluster, and how these fields couple to each other. The basic assumptions and principles behind our model are as follows:

1. The only two independent fields are the packing fraction of bees in the cluster in the cluster \( \rho \) (equivalent to density) and the air temperature \( T \). These then determine the bee body temperature and metabolism which can be written as a function of the local air temperature, and convection of air in the form of upwards air currents depends entirely on the global bee packing fraction and temperature profiles.

2. We treat the cluster as an active porous structure, with a packing fraction-dependent conductivity and permeability. Bees metabolically generate heat, which then diffuses away through conduction and is also drawn upwards through convection. The boundary of the cluster has an air temperature that is simply the ambient temperature.

3. Cold bees prefer to huddle densely, while hot bees dislike being packed densely. In addition, bees attempt to push their way to higher temperatures. The movement of bees is determined by a behavioural variable which we denote by “behavioural pressure” \( P_b (\rho, T) \), which we use to characterize their response to environmental variables such as local packing fraction and temperature. In terms of this variable, we assume that the bees move from high to low behavioural pressure, a notion that is similar in spirit to that of “social forces” used to model pedestrian movements \[16\]. Here, we must emphasize that behavioural pressure is not a physical pressure. For packing fraction to be unchanging, behavioural pressure must be constant throughout the cluster.

4. We assume that the number of bees in the cluster is fixed and that the cluster is axisymmetric with spherical boundaries. At higher temperatures, the clusters become elongated and misshapen, so that the assumption is no longer accurate; however, this does not change our results for thermoregulation qualitatively. In general, to determine the shape of the cluster, we must account for both heat and force balance, but we leave this question aside in the current study.

A model based on these assumptions can be used to study both the equilibria and dynamics. However, systems which reach equilibrium quickly are best understood in terms of their equilibrium behaviour rather than the specifics of how this equilibrium is reached. Since a swarm cluster is a constantly changing network of attachments between bees, as bees grab onto and let go of nearby bees, and can also detach and reattach themselves at different points on the surface of the cluster, these “microscopic” dynamical processes allow the cluster to quickly equilibrate to changes in ambient conditions \[2, 15\]. Thus, we will mostly focus on the resulting equilibria, but consider the slow dynamical modes that allow it to respond to large scale weak forcing, as they are particularly important in the context of cluster stability.

III. MATHEMATICAL FORMULATION

The two independent fields in our model are bee packing fraction \( \rho(\vec{r}, t) \in [0,1] \), and the air temperature \( T(\vec{r}, t) \); both of these are functions of space and time, we will focus primarily on the equilibrium behaviour of these fields. For a static cluster of bees modeled as an active porous medium that generates heat and is permeable to air, the heat generated metabolically must balance heat lost due to conduction and convection, so that

\[
\rho \mathcal{M}(T) + \nabla \cdot (k(\rho) \nabla T) - C \mathcal{U} \cdot \nabla T = 0 \quad |\vec{r} \in \Omega
\]

\[
T = T_a \quad |\vec{r} \in \delta \Omega, \tag{1}
\]

where \( k(\rho) \) is the packing fraction-dependent conductivity, \( \mathcal{M}(T) \) is the metabolic heat production rate of the bees per unit volume, and \( C \) is the volumetric heat capacity of air. We model the conductivity of the cluster as arising from a superposition of random convection currents within the cluster which are suppressed at high bee packing fraction and the bare conductivity of the bees treated as a solid, and approximate this by a function \( k(\rho) = k_0 \frac{1 - \rho}{\rho} \). Although this form diverges as \( \rho \to 0 \) and random convection currents are unsuppressed, the \( \rho \) never vanishes in the interior of the cluster, so that this limitation is not a problem. Likewise, by bounding
\(\rho\) from above, we prevent \(k\) from vanishing in the cluster. We further assume that the mean flux per unit area \(u\) is determined by Darcy’s law for the flow of an incompressible buoyant fluid through a porous medium, so that:

\[
\begin{align*}
\textbf{u} &= [g\alpha_{\text{air}}(T - T_a)\hat{z} - \nabla P] \cdot \kappa(\rho)/\eta \mid_{\bar{\text{F}}\in\Omega} \\
\nabla \cdot \textbf{u} &= 0, \quad P = 0 \mid_{\bar{\text{F}}\in\partial\Omega}. \tag{2}
\end{align*}
\]

FIG. 1: A cluster at an ambient temperature of a) 11°C and b) 27°C, approximately 12 cm across [Photo courtesy of [17]]. Note that number of bees inside the cluster is nearly constant; change in cluster width is a result of changes in bee packing fraction [11].

Schematic of interior (\(\Omega\)), and boundary (\(\delta\Omega\)), with mantle-core structure. \(\hat{s}, \hat{z}\) are radial, vertical directions in polar coordinates.

Here Eq. (2) relates the flow velocity to the effects of thermal buoyancy and the pressure gradient, while Eq. (3) is just the incompressibility condition, with \(\kappa(\rho)\) the packing fraction-dependent permeability, \(\alpha\) is the coefficient of thermal expansion, \(\rho_{\text{air}}\) is the density of air, \(\eta\) is the viscosity of air, \(g\) is acceleration due to gravity, and \(P\) is air pressure. We assume that the permeability of the cluster may be approximated via the Carman-Kozeny equation \(\kappa(\rho) = \kappa_0(1 - \rho^3)\), used to describe the permeability of randomly packed spheres [18].

In general, the bee metabolic activity is not a constant, and depends on a number of factors such as temperature, age, oxygen and carbon dioxide concentration etc. [2, 19, 20]. To keep our model as simple as possible, we start by assuming that the metabolic rate is temperature-independent, with \(M(T) = M_0\) and show that this is sufficient to ensure robust thermoregulation, setting apart the details of the calculations that show that our model can also yield robust thermoregulation with a temperature-dependent metabolic rate (Append. D).

To close our set of equations, we still must relate \(\rho(\hat{r})\) to \(T(\hat{r})\), which we do by making a hypothesis that bees respond to local packing fraction and temperature changes by changing their packing fraction to equalize an effective behavioural variable which we call the behavioural pressure. Our formulation of this behavioural pressure relies on two assumptions that are based on observations:

1. In their clustered state, bees have a natural packing fraction which is a function of the local temperature. This natural packing fraction decreases with increasing temperature, and increases with lower temperature, until it reaches a maximum packing fraction \(\rho_{\text{max}}\). Effectively, cold bees prefer to be crowded, while hot bees dislike being crowded, consistent with a variety of experiments in the field and in the laboratory [2, 5, 13, 17].

2. In addition to having a temperature-dependent natural packing fraction, bees also like to push their way towards higher temperatures. This will cause areas of equal local temperature to pack more densely at low ambient temperatures, consistent with observations [2].

With these constraints in mind, a minimal model for bee behavioural pressure suggests the piecewise linear function

\[
P_b(\rho, T) = \begin{cases} 
-\chi \cdot T + |\rho - \rho_m(T)| & \rho \leq \rho_{\text{max}} \\
\rho_{\text{max}} & \rho > \rho_{\text{max}},
\end{cases} \tag{4}
\]

where \(\rho_m(T) = \min\{\rho_{\text{max}}, \rho_0 - \rho' T\}\) is the natural packing fraction, with the constants \(\rho_0\) and \(\chi\) having units of temperature\(^{-1}\). We emphasize that our model assumes that individual bees have no independent homeostatic set points for temperature or packing fraction; the behavioural pressure depends on a combination of \(\rho\) and \(T\). Furthermore, we note that the bee packing fraction in a cluster at equilibrium can never be lower than the natural packing fraction; in our formulation, the behavioural pressure is always higher at lower packing fractions and indeed is the cause of the basic aggregation behaviour that creates the cluster. Additionally, the absolute behavioural pressure is of no consequence; only gradients and relative values are important. Further evidence for a behavioural pressure comes from experiments [4] who observed a relation between the ambient temperature and the core density, even when the core temperature remained approximately constant.

To complete the formulation of our problem, we need to specify boundary conditions for the temperature and packing fraction fields. As we have stated earlier, the surface bees will be at the ambient temperature; furthermore, since they can freely expand and contract to minimize their behavioural pressure, we assume that they will be at their natural packing fraction \(\rho_m(T_a)\).

Comparing our model with earlier models such as the Myerscough model [10] and the Watmough-Camazine model [11], we note that both fit into this general behavioural pressure framework (Append. B). However, our work differs from these studies in that we synthesize and generalize the implicit relation between bee behaviour and their environmental variables in terms of the behavioural pressure, choosing a form for behavioural pressure which combines elements from both models.

\[
\]
and is consistent with experimental observations. In addition, we account for the role of fluid flow and convection of heat in the cluster, which breaks the vertical symmetry, and is potentially relevant in heat transport at high Rayleigh number $Ra = \frac{\rho g \alpha T R^2}{\eta k} \cdot (T - T_a) \kappa R$.

**Dimensionless equations:** To reduce the number of parameters in our model, we make our equations dimensionless. Noting that the total number of bees in the cluster is constant allows us to define a dimensionless total bee volume $V = \int \int \int \rho dv / V_0$, where $V_0$ is the total bee volume of a typical cluster, i.e. the average volume of a bee times the average number of bees in a cluster. Upon setting a characteristic length scale to be the radius of a sphere of volume $V_0$, $R_0 = \sqrt[3]{\frac{3V_0}{4\pi}}$, we write $\int \int \int \rho dv = (4\pi/3)V$. Scaling the temperature so that typical ambient temperature of 15°C → 0, and the goal temperature of 35°C → 1, we use the dimensionless variables:

\[
T \rightarrow \frac{T - 15°C}{20°C}, \quad T_a \rightarrow \frac{T_a - 15°C}{20°C}
\]

\[
\kappa_0 \rightarrow \kappa_0 \cdot \frac{g \rho_0 \alpha \epsilon}{\eta M_0 R_0} \cdot (20°C)^2, \quad k_0 \rightarrow \frac{k_0 (20°C)}{R_0^3 M_0}
\]

\[
M_0 \rightarrow 1, \quad \rho' \rightarrow \rho' \cdot (20°C), \quad \chi \rightarrow \chi \cdot (20°C)
\]

and write the dimensionless form of Eqs. (1) to (4) as:

\[
\rho M_0 + \nabla \cdot (k_0 (\rho \kappa_0 \nabla T) - u \cdot \nabla T) = 0 \quad (5)
\]

\[
u = [(T - T_a) \hat{z} - \nabla P] \cdot \kappa (\rho), \quad \nabla \cdot u = 0. \quad (6)
\]

\[
k_0 (\rho) = k_0 \cdot \frac{1 - \rho}{\rho^2}, \quad \kappa (\rho) = \kappa_0 \cdot (1 - \rho)^3 \quad (7)
\]

\[
\chi_{\rho(T)} = \left\{T \rightarrow \chi + \frac{|\rho - \rho_m(T)|}{\rho} \leq \rho_{\max}, \quad \rho > \rho_{\max}ight. \quad (8)
\]

\[
\rho_m(T) = \min \left\{\rho_{\max}, \rho_0 - \rho' T\right\}, \quad (9)
\]

for the packing fraction and temperature profiles of the cluster which depend on the dimensionless parameters $\rho_{\max}, \rho_0, \rho', \chi, T_a, k_0, \kappa_0, N$ [21].

**Information transfer through equalization of behavioural pressure:** On the outer boundary of the mantle, the air temperature is equal to ambient temperature, so that minimizing the behavioural pressure requires that the packing fraction at the mantle be the natural packing fraction at the ambient temperature $\rho_m(T_a)$. At equilibrium, (4) then implies that the behavioural pressure in the mantle and thus throughout the cluster will be $-T_a \cdot \chi$. This means that throughout the cluster, we may write the local bee packing fraction as a function of both the local temperature and the ambient temperature, i.e $P_b(\rho(T, T_a), T) = -T_a \cdot \chi$ which we solve to find

\[
\rho(T, T_a) = \min \{\rho_m(T) + \chi \cdot (T - T_a), \rho_{\max}\}
\]

\[
= \min \{\rho_0 + T (\rho' + \chi) - \chi T_a, \rho_{\max}\}
\]

\[
= \min \{\rho_0 + T \cdot c_0 + T_a \cdot c_1, \rho_{\max}\}, \quad (11)
\]

where we have made substitutions $c_0 = \rho' + \chi, c_1 = -\chi$. Intuitively, the $c_0$ term characterizes the sensitivity of core temperature to ambient temperature, but the cluster cannot fully adapt at low $T_a$ through this term alone; adaptation at lower ambient temperatures requires the $c_1$ term, which is eventually responsible for overheating in the core at very low $T_a$. In Fig. 2 we graph the packing fraction $\rho(T, T_a)$ obtained by tracing contours of equal $P_b$ which allows us to write the equilibrium local packing fraction everywhere in terms of the conditions at the boundary. Bees respond to their local conditions and move accordingly, and these variations in packing fraction propagate information about ambient temperature throughout the entire cluster without long-range communication. Although we have mapped $P_b(\rho, T)$ to $\rho(T, T_a)$ for just one choice of behavioural pressure, our approach will work for any equation of state $P_b(\rho, T)$ which uniquely defines a stable packing fraction, i.e. with $dP_b/d\rho > 0$.

**IV. SIMULATIONS AND RESULTS**

While the boundary of the cluster is assumed to have spherical symmetry, the temperature and packing fraction fields inside do not have to have spherical symmetry owing to the effects of convection. However, the fields still

FIG. 2: Graphical representation of Eqs. (9) to (11), showing how local packing fraction is obtained through behavioural pressure (Online version in colour). At low $T_a$, bees on the mantle will pack at $\rho = \rho_{\max}$ (upper circle), and behavioural pressure is high throughout the cluster leading to higher interior packing fraction (upper solid line) and overpacking. At high $T_a$, bees on the mantle will pack more loosely (lower circle), and low behavioural pressure throughout the cluster leads to low interior packing fraction (lower solid line). The solid lines run along contour lines of equal $P_b(\rho, T)$, as behavioural pressure must be uniform through the cluster. Coefficients used are $\rho_0 = .85, \rho' = .75, \rho_{\max} = .8, \chi = .3$.

The set of equations [21] with the boundary condition that the surface temperature of the cluster is the ambient temperature completes the formulation of the problem to determine $\rho(\vec{r}), T(\vec{r})$. 
have cylindrical symmetry, and therefore we can represent \( \rho(\mathbf{r}) \), \( T(\mathbf{r}) \) as \( \rho(s, z), T(s, z) \), where \( s \) is the distance from the central axis and \( z \) is the height. With this coordinate representation, we solve the governing equations [5, 8, 7, 8, and 11] in a spherically bounded domain using a simple discretization scheme with 30 values of \( s \) and 60 values of \( z \) (Append. C.2).

Our choice of the dimensionless parameter \( k_0 = 0.2 \) is constrained by experiments [19], while we estimate \( \kappa_0 = 1 \) though a simple calculation assuming the bees to be randomly packed spheres, and \( p_{\text{max}} = .8 \), slightly higher than the maximum packing fraction of spheres, as bees are more flexible. (Append. A.2). However, the parameters defining bee movement and behaviour, namely \( c_0, c_1, \) and \( \rho_0 \) are experimentally unknown. Guided by the general observation that physiological performance is often improved by changing parameters while basic mechanisms remain unchanged, we optimize these parameters to achieve robust thermoregulation, i.e. the core temperature remains close to 1(35°C) over a range of scaled ambient temperatures \( T_a \in [-0.7, 0.8] \) corresponding to a real ambient temperature \( T_a \in [0, 30]°C \). For the choice of parameters \( \rho_0 = .85, c_0 = .45, c_1 = .3, \) we find that within this wide range of \( T_a \) and a factor of three in \( V \), the dimensionless core temperature stays in the range \( 0.7 - 1.3 \) corresponding to a real temperature range of \( 29 - 41°C \), while the core temperature itself increases monotonically with \( \rho_0 \) in an analytically solvable way (Append. A.3).

Our simulations also capture the qualitative mantle-core structure of the cluster (Fig. 3) with a dense mantle surrounding a sparse core. We also find that at high ambient temperatures, the cluster expands and the mantle thickens, and at low ambient temperatures the cluster contracts and the mantle thickens. Furthermore, we find that the core temperature, which is also the maximum temperature of the cluster, is higher at low ambient temperatures resulting from “overpacking”, consistent with experiments of [2, 15, 15], and predicted earlier [11]. Finally, we note that the temperature profile is vertically asymmetric due to convection, causing the point of maximum temperature to rise above the geometric center of the cluster, as observed in experiments [2, 15].

\[ \text{FIG. 3: Adaptation with temperature-independent metabolism (Online version in colour).} \]

\[ a) \text{Comparison of temperature and packing fraction profiles at high(8) and low(-7) ambient temperature, where the dimensionless total bee volume} V = 1. \]

\[ b) \text{Core temperature as a function of ambient temperature and total bee volume showing how clusters swell with temperature, consistent with experiment.} \]

\[ c) \text{Cluster radius as a function of ambient temperature and total bee volume showing how clusters swell with temperature, consistent with experiment.} \]

\[ \text{For} V, \rho_0, c_0, c_1, \rho_{\text{max}} = .8. \]

\[ \text{For heat transfer, we choose coefficients of} \]

\[ \rho_0 = .85, c_0 = .45, c_1 = .3, \rho_{\text{max}} = .8. \]

\[ \text{For heat transfer, we choose coefficients of} \]

\[ k_0 = .2, \kappa_0 = 1. \]

\[ \text{V. DISCUSSION} \]

We have shown that it is possible to provide a self-organized thermoregulation strategy in bee clusters over a range of observed ambient temperatures in terms of a few behavioural parameters. Our theory fits into a broader framework for understanding collective behaviour where the organism responds to the environment, but in doing so, changes it, and its behaviour until it reaches a steady state. Here, our model takes the form of a two-way coupling between bee behaviour and local temperature and packing fraction, quantified in terms of an effective behavioural pressure whose equalization suffices to regulate the core temperature of the cluster robustly. Although our choice of the form of the behavioural pressure is likely too simplistic, it is consistent qualitatively with experimental observations, and we think it provides the correct framework within which we can start to quantify collective behaviour. Furthermore, our strategy might be useful in biomimetic settings. Our formulation for behavioural pressure shows that in a cluster, bee packing fraction should depend only on local temperature and the temperature of bees at the boundary, which effectively control the surface packing fraction. This observation provides an immediately testable prediction: a cluster may be “tricked” into overheating its core by warming the bees just below the surface, while exposing the surface bees to a low temperature to increase behavioural pressure. As pointed out earlier, experiments and observations of bee core temperature and bee packing fraction [4] are consistent with these ideas, although a direct experiment of this type does not seem to have been carried out.

Our model adjusts well in response to changes in ambient temperatures, but it doesn’t have the same level of tolerance to different total bee volume that honeybees exhibit. We have used a continuum model where the bees on the surface exposed to the ambient temperature equilibrate to that temperature, but in reality the first layer of bees is hotter than ambient temperature and supports...
a large temperature gradient driven by heat from interior bees. This means that the surface bees feel an average temperature higher than ambient due to the interior bees, and we predict this should give a large cluster a lower behavioural pressure than a small cluster at the same ambient temperature. This should reduce the sensitivity of core temperature to total bee volume, and we have confirmed this in simulations (Append. E).

We close with a description of some possible extensions of our study. Currently, our model ignores changes in shape of the cluster associated with force balance via the role of gravity, and the associated effects on thermoregulation. In reality, the cluster is a network of connections between bees which changes shape and size due to a combination of mechanical forces and heat balance, and a complete theory must couple these two effects as well.

Our heat balance equation and estimates (Append. A.2) suggest that while convective terms are responsible for the asymmetry in the temperature profile, they do not play an important global role in thermoregulation. However, our calculation assumes a uniform packing which does not accurately represent the microscopic structure of the cluster; at high ambient temperatures, swarm clusters are observed to “channelize”, where channels open up to ventilate. Studying a simple “behavioural pressure-taxis” dynamical law (Append. F), we find no linear instability that leads to channelization without an “anemotaxis” mechanism, where behavioural pressure increases with |u|; instead we see only one kind of linear instability which comes from the mathematical necessity of fixing cluster radius. This raises the question of whether there are anemotaxis mechanisms. If not, how can channelization result from bee-level dynamics and mechanics? We have also neglected active cooling, which includes funning, evaporative cooling (which can give up to \( \sim 50^\circ C \) of cooling), diffusion of heat through diffusion of bees, and effects of oxygen and \( CO_2 \). Finally, we have also neglected any implications of bee age distribution, despite knowledge of the fact that younger bees tend to prefer the core and produce less heat, while older bees prefer the mantle and can produce more heat. Accounting for these additional effects will allow us to better characterize the ecological and possibly even evolutionary aspects of thermoregulation.

Thermoregulation is a necessity for a wide variety of organisms. When achieved collectively, individuals expend effort at a cost that accrues a collective benefit. The extreme relatedness of worker bees in a cluster and near-inability to reproduce implies that the difference between the individual and the collective is nearly nonexistent, so that cost and benefit are equally shared. However, many other organisms are faced with the “huddler’s dilemma”, expending individual metabolic effort is costly, and benefits a group that is only partially related. Because genetic relatedness, metabolic costs, individual temperature and spatial positions are all easily measurable, a collective thermoregulatory system is an ideal context in which to study the tangible evolution of cooperation and competition by building on our framework theoretically to account for competition and cooperation.

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Appendix A: Units, parameter estimation, and dimensional analysis

1. Table of quantities and associated units

| Quantity                  | Symbol | Units                          | Typical Values                  |
|---------------------------|--------|--------------------------------|----------------------------------|
| Bee Packing Fraction      | $\rho$ | Dimensionless                  | $\sim 2.\cdot.8$                |
| Bee Metabolic Rate        | $M$    | Power/Volume                   | $\sim .001 - .01\text{W/cm}^2$   |
| Heat Conductivity         | $k$    | Power/(Distance $\cdot$ Temperature) | $\sim .0004 - .0006\text{W}/(\text{cm}^2\cdot\text{C})$ (Append. A 2) |
| Permeability              | $\kappa$ | Distance$^2$                   | $(.05\text{cm})^2$? (Append. A 2) |
| Darcy Velocity           | $u$    | Distance/Time                  | $\sim 1\text{cm/s}$? (Append. A 2) |
| Air Heat Capacity         | $C$    | Energy/Volume $\cdot$ Temperature | $1.2\cdot10^{-3}\text{J}/(\text{mL} \cdot \text{C})$ |
| Gravity                   | $g$    | Distance/time                  | $981\text{cm/s}^2$              |
| Air Density               | $\rho_{air}$ | Mass/Volume                 | $1.2\cdot10^{-3}\text{gram/mL}$ |
| Coefficient of thermal expansion | $\alpha$ | Temperature$^{-1}$           | $1/300\text{C}$                 |
| Air Viscosity             | $\eta$ | Pressure $\cdot$ time         | $1.8\cdot10^{-4}\text{gram/cm} \cdot \text{s}$ |
| Behavioural Pressure      | $P_b$  | Behavioural Pressure/Time     | ?                                |
| Thermotactic Coefficient  | $\chi$ | Behavioural Pressure/Time     | ?                                |

Many sets of dynamical equations for bee movement will result in the same equilibrium where behavioural pressure, whose units depend on the set of dynamical equations used, remains constant. For example, a simple taxis model is one where:

$$\frac{dp}{dt} = -\nabla \cdot \mathbf{J}, \quad \mathbf{J} = -P_b \nabla T$$

would mean that behavioural pressure has units of Distance$^2$/Time. A more complicated evaporation/condensation model would have the form

$$\frac{dp(\mathbf{r})}{dt} = \iiint \left[ (P_b(\mathbf{r}) - P_b(\mathbf{r}')) \frac{e^{\frac{x'^2+y'^2}{2\sigma^2}}}{\sigma^3} T(\rho(\mathbf{r}), \rho(\mathbf{r}')) \right] d^3 \mathbf{r}'$$

where $\sigma$ is the evaporation and condensation radius and $T$ is some transfer coefficient would give units of 1/Time. A yet more complicated model involving mechanical compressibility or viscosity would have yet another set of dimensions for behavioural pressure. All of these models would, however, yield the same static solution.

2. Estimation of heat conductivity, metabolic rate, and permeability

For the metabolic rate and heat conductivity we use estimates from the experiments of Southwick [19], where a cluster of 4250 bees(608 grams) is put into a set of roughly planar, parallel honeycombs, and the temperature profile and oxygen consumption are measured. The bees are roughly uniformly distributed with a bee packing fraction $\rho$ of about .5, for which the metabolic rate is roughly uniform, and the temperature is well approximated with a parabolic profile, with a temperature of $34^\circ\text{C}$ at the core, and $11^\circ\text{C}$ at the edge $9.5\text{cm}$ from the core, parallel to the combs. The combs are very well insulating, so heat transfer occurs primarily in the two directions parallel to the combs. Then, within the cluster:

$$T \approx 34^\circ\text{C} - 23^\circ\text{C} \cdot \frac{x^2 + y^2}{9.5\text{cm}^2},$$

$$\nabla^2 T \approx -4 \cdot \frac{23^\circ\text{C}}{(9.5\text{cm}^2)} = \frac{1.0^\circ\text{C}}{\text{cm}^2}.$$ 

The oxygen consumption rate is measured to be $6.5\text{mL/min}$ which gives a volumetric metabolism of $0.0035 \cdot \text{W}/(\text{cm}^3)$, assuming a bee specific weight of 1 gram/(cm$^3$), and an oxygen to energy conversion of $3.5\text{ml}\text{O}_2/\text{kg}\cdot\text{min} \equiv .0012\text{W/gram}$. This metabolic rate agrees well with the experiments of Heinrich [2]. We now have all the pieces to calculate the
conductivity using the conduction heat balance:

\[ k \nabla^2 T + \rho M = 0 \implies k = \frac{1.7 \cdot 10^{-3} W}{cm^2^\circ C}. \]

At \( \rho = .8 \), the maximum packing fraction we allow, the conductivity becomes close to the value of \( 2.4 \cdot 10^{-3} W/(cm^2^\circ C) \) for fur and feathers [25] as suggested by Southwick which gives us some level of confidence in the functional form \( k_0 \frac{1 - \rho}{\rho} \) we have chosen for conductivity.

To estimate \( k_0 \), we use the Carman-Kozeny equation [18]. The average bee weighs about .14 grams, which corresponds to a sphere of diameter \( \sim .65 cm \). In the absence of any detailed information about the bee structure in the cluster, we treat the cluster as a system of randomly packed spheres. It would be interesting to measure and better understand convective gas and heat transfer within swarm clusters. Using this diameter in the Carman-Kozeny equation, we find:

\[ k_0 \approx \frac{D^2}{180} \approx (.05 cm)^2. \]

A typical cluster has about 10,000 bees, which means about 1.4 kg, giving \( R_0 = 7 cm \). Plugging these values in(Append. [A.3]), we find dimensionless conductivity and permeability to be:

\[ k_0 \approx .2, \quad \kappa_0 \approx 1. \]

3. Dimensional analysis

Our conditions for heat balance imply that:

\[ \rho M(T) + \nabla \cdot (k(\rho)\nabla T) - C \mathbf{u} \cdot \nabla T = 0 \mid \vec{r} \in \Omega \]

\[ T = T_a \mid \vec{r} \in \delta \Omega \]

\[ \mathbf{u} = [g\alpha \rho_{air}(T - T_a) \hat{z} - \nabla P] \cdot \kappa(\rho)/\eta \mid \vec{r} \in \Omega \]

\[ \nabla \cdot \mathbf{u} = 0, \quad P = 0 \mid \vec{r} \in \delta \Omega. \]

while our equation for behavioural pressure reads:

\[ P_b(\rho, T) = \left\{ \begin{array}{ll}
-\chi \cdot T + |\rho - \rho_m(T)| & \rho \leq \rho_{max} \\
\infty & \rho > \rho_{max}
\end{array} \right. \]

\[ \rho_m(T) = \min \{ \rho_{max}, \rho_0 - \rho/T \} \]

Setting our unit of length to be \( R_0 \), we make the transformation \( \nabla \rightarrow \nabla/R_0 \). Then our heat balance equations read:

\[ \rho M(T) + \frac{1}{R_0^2} \nabla \cdot (k(\rho)\nabla T) - \frac{1}{R_0^2} C \mathbf{u} \cdot \nabla T = 0 \mid \vec{r} \in \Omega \]

\[ T = T_a \mid \vec{r} \in \delta \Omega \]

\[ \mathbf{u} = \left[ g\alpha \rho_{air}(T - T_a) \hat{z} - \frac{1}{R_0} \nabla P \right] \cdot \kappa(\rho)/\eta \mid \vec{r} \in \Omega \]

\[ \nabla \cdot \mathbf{u} = 0, \quad P = 0 \mid \vec{r} \in \delta \Omega. \]

On making the substitutions:

\[ T \rightarrow \frac{T - 15^\circ C}{35^\circ C - 15^\circ C}, \quad T_a \rightarrow \frac{T - 15^\circ C}{35^\circ C - 15^\circ C} \]

leads to the goal temperature of 35\(^\circ\)C yielding a dimensionless temperature of unity, and a typical ambient temperature
of 15°C corresponding to a dimensionless temperature that vanishes. Then our system of equations becomes:

$$
\rho M(T) + \frac{35^\circ C - 15^\circ C}{R_0^2} \nabla \cdot (k(\rho) \nabla T) - \frac{35^\circ C - 15^\circ C}{R_0} \cdot \mathbf{C} \mathbf{u} \cdot \nabla T = 0 \mid \mathbf{r} \in \Omega
$$

$$
T = T_a \mid \mathbf{r} \in \partial \Omega
$$

$$
\mathbf{u} = \left[ (35^\circ C - 15^\circ C) \alpha g_{\text{air}} (T - T_a) \hat{z} - \frac{1}{R_0} \nabla P \right] \cdot \frac{\kappa(\rho)}{\eta} \mid \mathbf{r} \in \Omega
$$

$$
\nabla \cdot \mathbf{u} = 0, \quad P = 0 \mid \mathbf{r} \in \partial \Omega.
$$

We divide all terms in the heat equation by the base metabolism $M_0$ to yield:

$$
\rho M(T)/M_0 + \frac{35^\circ C - 15^\circ C}{M_0 R_0^2} \nabla \cdot (k(\rho) \nabla T) - \frac{35^\circ C - 15^\circ C}{M_0 R_0} \cdot \mathbf{C} \mathbf{u} \cdot \nabla T = 0 \mid \mathbf{r} \in \Omega
$$

$$
T = T_a \mid \mathbf{r} \in \partial \Omega
$$

$$
\mathbf{u} = \left[ (35^\circ C - 15^\circ C) \alpha g_{\text{air}} (T - T_a) \hat{z} - \frac{1}{R_0} \nabla P \right] \cdot \frac{\kappa(\rho)}{\eta} \mid \mathbf{r} \in \Omega
$$

$$
\nabla \cdot \mathbf{u} = 0, \quad P = 0 \mid \mathbf{r} \in \partial \Omega.
$$

Finally, making the substitutions for the coefficients

$$
\mathcal{M} \rightarrow \mathcal{M}/M_0, \quad k \rightarrow k \cdot \frac{35^\circ C - 15^\circ C}{M_0 R_0^2}, \quad \kappa \rightarrow \kappa \cdot \frac{(35^\circ C - 15^\circ C)^2 \alpha g_{\text{air}} \cdot \mathbf{C}}{M_0 R_0 \eta},
$$

$$
\rho_0 \rightarrow \rho_0 - \rho' (15^\circ C), \quad \rho' \rightarrow \rho' \cdot (35^\circ C - 15^\circ C), \quad \chi \rightarrow \chi \cdot (35^\circ C - 15^\circ C).
$$
leads to our full dimensionless set of equations for heat balance:

\[ \rho M(T) + \nabla \cdot (k(\rho)\nabla T) - \mathbf{u} \cdot \nabla T = 0 \mid_{\mathbf{r} \in \Omega} \]

\[ T = T_a \mid_{\mathbf{r} \in \partial \Omega} \]

\[ \mathbf{u} = [(T - T_a) \hat{z} - \nabla P] \cdot \kappa(\rho) \mid_{\mathbf{r} \in \Omega} \]

\[ \nabla \cdot \mathbf{u} = 0, \quad P = 0 \mid_{\mathbf{r} \in \partial \Omega}. \]

while the dimensionless behavioural pressure reads:

\[ P_b(\rho, T) = \begin{cases} -\chi + |\rho - \rho_m(T)| & \rho \leq \rho_{\max} \\ \infty & \rho > \rho_{\max} \end{cases}, \]

\[ \rho_m(T) = \min\{\rho_{\max}, \rho_0 - \rho'T\}. \]

Our model has seven parameters \( \rho_{\max}, \rho_0, \rho', \chi, T_a, k_0, \kappa_0 \).

a. Note on further dimensional analysis

We note that, when the metabolic rate is temperature independent, the goal temperature and the typical independent ambient temperature have no bearing on the actual behaviour of the model, only on whether it represents effective thermoregulation. Then, we may set the temperature where the packing becomes maximally dense, \( T_{\text{packed}} = \rho_0 - \rho_{\max} \), to be zero, and the temperature at which the packing fraction becomes zero, \( T_{\text{empty}} = \frac{\rho_0}{\rho'} \), to be 1. We can then make slightly different substitutions for the coefficients:

\[ M \rightarrow 1, \quad k \rightarrow k \cdot \frac{T_{\text{empty}} - T_{\text{packed}}}{M_0 R_0^2} \]

\[ \kappa \rightarrow \kappa \cdot \frac{(T_{\text{empty}} - T_{\text{packed}})^2 \alpha g \rho_{\text{air}} \cdot C}{M_0 R_0^2 \eta}, \]

\[ \rho_0 \rightarrow \rho_{\max}, \quad \rho' \rightarrow \rho_{\max}, \quad \chi \rightarrow \chi \cdot (T_{\text{empty}} - T_{\text{packed}}). \]

and find that there are now only five free parameters \( \rho_{\max}, \chi, T_a, k_0, \kappa_0 \). We do not carry out this extra analysis in the main body of the paper because this causes us to lose sight of what the goal core temperature and typical ambient temperatures are.

Appendix B: Behavioural pressure formalism and its antecedents in previous models

To understand how our behavioral pressure formalism fits in with previous models, we compare them within this framework. The Myerscough model assumes that the bee density depends only on local temperature, and thus can be written as \( P_b(\rho, T) = |\rho - \rho_m| \), where \( \rho_m = 8 \text{bees/cm}^3 \cdot (1 - \frac{T}{40^\circ \text{C}}) \). The Watmough- Camazine model defines a dynamical law for the bee density via the equations:

\[ \dot{\rho} = -\nabla \cdot \mathbf{J} \]

\[ \mathbf{J} = -\mu(\rho) \nabla \rho - \rho \chi(T) \nabla T, \]

where \( \mu(\rho) > 0 \) is a motility function, and \( \chi(T) \) is a thermotactic function. This may be written as

\[ \mathbf{J} = -\rho \cdot \left[ \frac{\mu(\rho)}{\rho} \nabla \rho + \chi(T) \nabla T \right] = -\rho \nabla P_b, \]
where the behavioural pressure $P_b$ is defined as:

$$P_b(\rho, T) = \int_{\rho'} \frac{\mu(\rho')}{\rho'} d\rho' + \int \chi(T') dT'.$$

We note that as $\mu(\rho) > 0$ for all $\rho$, the density component is minimized as $\rho \to 0$, and the density at the surface must be fixed to prevent the cluster from falling apart, unlike in our behavioral pressure framework. Additionally, because the behavioural pressure can be divided into temperature and density components, the point of highest density will always be at the same temperature, regardless of size or ambient temperature, which is not observed experimentally. This is in contrast with our formalism.

**Appendix C: Numerical methods**

1. Method of solving for equilibrium

To reach equilibrium, we use an iterative scheme, described by the following pseudocode:

- $T(\vec{r}) \leftarrow T_a$
- $\mathbf{R} \leftarrow R_0/3^{1/3}$
- $\rho(\vec{r}) \leftarrow \begin{cases} \rho_{\text{max}} : |\vec{r}| \leq \mathbf{R} \\ 0 : |\vec{r}| > \mathbf{R}, \end{cases}$

**repeat**

- $\rho_{\text{new}}(\vec{r}) \leftarrow \rho(T(\vec{r}), T_a)$
- $\rho(\vec{r}) \leftarrow \rho(\vec{r}) + c_\rho (\rho_{\text{new}}(\vec{r}) - \rho(\vec{r}))$

**Expand or contract the bee packing fraction and temperature profiles to normalize total bee volume**

- $\mathbf{R} \leftarrow \mathbf{R} / \sqrt[3]{\mathbf{V}}$, Scaling ratio
- $\rho(\vec{r}) \leftarrow \rho(\vec{r}/\mathbf{R})$
- $T(\vec{r}) \leftarrow T(\vec{r}/\mathbf{R})$

**Solve for $u$ at fixed temperature and density, then solve for $T$ at fixed $M$ and $u$**

- Find $u$ such that: $u = [(T - T_a) \hat{z} - \nabla P] \cdot \kappa(\rho), \nabla \cdot u = 0$
- Find $T_{\text{new}}(\vec{r})$ such that: $\rho M(T) = -\nabla \cdot (k(\rho) \nabla T_{\text{new}}) - u \cdot \nabla T_{\text{new}}$
- $T(\vec{r}) \leftarrow T(\vec{r}) + (T_{\text{new}}(\vec{r}) - T(\vec{r}))$

**until** converged

The intermediate steps can be solved as a system of linear equations. The solution is considered to be converged when $\rho_{\text{new}} = \rho, T_{\text{new}} = T, \mathbf{R} = 1$ to within $10^{-10}$.

2. Discretization of space

To solve for the temperature and density profiles, we must first discretize the system. While spherical symmetry is broken due to convection, the system retains rotational symmetry about its axis. We therefore use cylindrical coordinates, where each cell is given indices $(i, j)$, and has coordinates which represent the distance from the central axis $s_{ij}$ and the vertical coordinate $z_{ij}$, where

$$s_{ij} = \left(i + \frac{1}{2}\right) \frac{\mathbf{R}}{n}, \quad z_{ij} = \left(j - \frac{1}{2}\right) \frac{\mathbf{R}}{n}.$$

$n$ is the radius of the cluster in cells. All cells with $s_{ij}^2 + z_{ij}^2 \leq \mathbf{R}$ are in the interior of the cluster, while all cells with $s_{ij}^2 + z_{ij}^2 \leq \mathbf{R}$ are at the exterior of the cluster, subject to the boundary conditions $T_{ij} = T_a, P_{ij} = 0, \rho_{ij} = 0$.

The volume of each cell with coordinates $(i, j)$ is $V_{ij} = 2\pi w^2 s_{ij}$, where $w = \frac{\mathbf{R}}{n}$ is the width of each cell. Each cell $(i, j)$ neighbors four other cells, $(i, j+1), (i, j-1), (i+1, j), (i-1, j)$, with the exception of cells which border the axis.
(i = 0), which only have three neighbors. The area shared by cell (i, j) and its outside horizontal neighbor (i + 1, j) is $2\pi w \left( s_{ij} + \frac{w}{2} \right)$. The area shared by cell (i, j) and its vertical neighbor (i, j + 1) is $2\pi ws_{ij}$.

**FIG. 4:** Cell layout of a cluster 30 cells in radius. Interior cells are coloured white, exterior cells are coloured grey.

**a. Heat equations**

For heat balance, we discretize our heat equations as

$$\frac{dQ_{ij}}{dt} = M(T_{ij}) \rho_{ij} V_{ij} - \sum_{(i',j') \in \text{network}} \frac{A_{ij,i'j'}}{w} \cdot H(k(\rho_{ij}), k(\rho_{i'j'})) \left( T_{i'j'} - T_{ij} \right) + \sum_{(i',j')} u_{ij,i'j'} \left[ -\theta(u_{ij,i'j'}) T_{ij} + \theta(-u_{ij,i'j'}) T_{i'j'} \right] = 0.$$

The first term corresponds to metabolic heat generation, and the second term is heat conduction, where $\sum_{(i',j')} A_{ij,i'j'}$ is the sum of all $(i',j')$ neighboring $(i,j)$, with the heat conductance between two neighboring cells depending on the harmonic mean of the conductance of each cell: $H(a,b) = \frac{2}{\frac{1}{a} + \frac{1}{b}}$. The third term represents convective heat transfer, with $u_{ij,i'j'}$ the air flow from cell $(i,j)$ to $(i',j')$.

We have chosen an “upwinding” scheme [26]: when air flows out of a cell, the outwards heat flux is determined by the temperature of that cell. When air flows into a cell from a neighboring cell, the inwards heat flux is determined by the temperature of the neighboring cell where the air originates. This scheme uses the Heaviside step function:

$$\theta(x) = \begin{cases} 0 & x \leq 0 \\ 1 & x > 0. \end{cases}$$

**b. Solving for buoyancy driven flow**

The air flow from cell $ij$ to neighboring cell $i'j'$ is given by:

$$u_{ij,i'j'} = H(\kappa(\rho_{ij}), \kappa(\rho_{i'j'})) \cdot \frac{A_{ij,i'j'}}{w} \cdot \left[ (P_{ij} - P_{i'j'}) + (z_{i'j'} - z_{ij}) \cdot \left( \frac{T_{ij} + T_{i'j'}}{2} - T_a \right) \right]$$

where the air conductance between two cells again depends on the harmonic mean of the permeability of each cell, and the pressure is set so that air flow is conserved in every cell i.e. $\sum_{(i',j')} u_{ij,i'j'} = 0$ for all cells $(i,j)$. This yields a set of linear equations that can be solved easily.
Appendix D: Temperature-dependent metabolic rate

To formulate our temperature-dependent metabolic rate, we note that bees have a higher base metabolic rate at high temperatures than at low temperatures. At moderate temperatures, bees on the mantle keep their body temperatures approximately $3^\circ$C above ambient temperature, and at below $15^\circ$C, they will “shiver” to keep their body temperature at $18^\circ$C\[15\]. Assuming a constant coefficient of thermal transfer between a bee and the surrounding air, this leads to a formulation of metabolism by shivering at air temperatures below $15^\circ$C, and gives us the full piecewise function for metabolic rate. Our metabolic rate for high temperatures comes from experiments involving oxygen consumption in swarm clusters \[15\].

\[
M(T) = M_0 \cdot \begin{cases} 
1 + \frac{15^\circ C - T}{3^\circ C} & : T < 15^\circ C \\
1 + \frac{T - 15^\circ C}{10^\circ C} & : T \geq 15^\circ C, 
\end{cases}
\]

where the base metabolism, $M_0$ has units of power/volume.

\[\text{FIG. 5: Metabolic rate as a function of temperature.}\]

The simulated cluster with temperature-dependent metabolic rate has the same qualitative features as when we use temperature-independent metabolic rates. We set $M_0$ to be half as high as when $M$ is temperature-independent, because now it represents the minimal metabolic rate, not the uniform one (Fig. 4).

\[\text{FIG. 6: Adaptation with temperature-dependent metabolic rate (Online version in colour). a) Comparison of temperature and packing fraction profiles at high(0.8) and low(-0.7) ambient temperature, where the dimensionless total bee volume $V$ is 1. b) Adaptation is nearly the same as for temperature independent metabolic rate. $T_a$ is also plotted to guide the eye. c) Cluster radius is also nearly the same as for temperature independent metabolic rate. For packing fraction, we choose coefficients of $\rho_0 = .85, c_0 = .5, c_1 = .25, \rho_{\text{max}} = .8$. For heat transfer, we choose coefficients of $k_0 = .4, \kappa_0 = 2$.}\]
Appendix E: Accounting for the role of finite bee size on thermoregulation

In the paper, we have defined the temperature at the boundary to be the ambient temperature, and we assumed the surface bees feel the ambient temperature, which sets the behavioural pressure accordingly. In reality, bees point their heads inwards, and feel a temperature gradient driven by the heat produced by interior bees. Therefore, it may be more realistic that the behavioural pressure is set by the temperature a slight distance inwards from the surface. If we include the effects of convection, this implies that spherical symmetry must be broken, and the temperature becomes not just a function of distance, but also dependent on angle. Therefore, to close our set of equations without having to delve deeper into the question of cluster shape, we must neglect upwards convection and only consider conduction. This gives us the system of equations:

\[ \rho \mathcal{M}_0 + \nabla \cdot (k(\rho) \nabla T) = 0, \quad k(\rho) = k_0 \cdot \frac{1 - \rho}{\rho}, \]

\[ T = T_a |_{\vec{r} \in \delta \Omega}, \]

where the behavioural pressure is now set by the temperature a distance of \( L_{\text{bee}} \) inside the cluster:

\[ \rho_0 \cdot T + T(\mathbf{R} - L_{\text{bee}}) \cdot \mathbf{c}_1, \quad \rho_{\text{max}} \]. \hfill (E1)

Assuming \( L_{\text{bee}} = 1 \text{ cm} \), slightly shorter than the body length of a worker bee, for an average cluster radius of 7 cm we find that the dimensionless \( L_{\text{bee}} \) of \( \approx .14 \). We simulate the system for dimensionless total bee volumes of .5, 1, and 3, with the same parameters of \( c_0 = .45, c_1 = .3, \rho_{\text{max}} = .8 \) as were used in paper. In the first system, to test thermoregulation with this effect, we choose \( L_{\text{bee}} = .14 \), and compensate for the slightly lower average behavioural pressure by increasing \( \rho_0 \) to .95. In the second system, we test thermoregulation without this effect, and so we choose \( L_{\text{bee}} = 0, \rho_0 = .85 \), as we did in the main body. We find that for large clusters, the temperature gradient created by the interior bees lowers behavioural pressure and loosens the cluster, mitigating overheating in the core and sensitivity to total bee volume.

FIG. 7: Comparison of core temperatures(Online version in colour). a) Behavioural pressure is set by the temperature slightly below the surface of the cluster, and sensitivity to total bee volume is mitigated. b) Behavioural pressure is set by ambient temperature, and overheating of the core is much worse. \( T_a \) is also plotted in each to guide the eye.

Appendix F: Results of linear stability analysis

Solving for linear stability using simple “behavioural pressure-taxis” dynamics (Append. F1), we find that all clusters simulated at a temperature-independent metabolic rate are stable. However, at low ambient temperatures,
clusters with a temperature-dependent (Append. D) metabolic rate can be linearly unstable via an overheating instability, which we believe to be a relic of fixing the boundaries of the cluster. In this instability, bees from the core move to the mantle, increasing the mantle thickness and insulation. This causes the core to heat up, increasing its behavioural pressure causing even more bees to move from the core to the mantle, leading to eventual runaway. We only see this instability when using a temperature-dependent metabolic rate, where an increased core temperature results in a greater net metabolic rate, aggravating the problem.

![Cluster Profile and Eigenvector](image)

**FIG. 8**: a) Cluster profile. b) Contour plots of instable eigenvector of linear response matrix (Online version in colour). Circle represents boundaries of the cluster, and a temperature-dependent metabolic rate is used. 

\[ J_0 = 1, V = 1.5, T_a = -7, \rho_0 = .85, c_0 = .5, c_1 = .25, \rho_{max} = .8 \]. For heat transfer, we choose coefficients of \[ k_0 = 4, \kappa_0 = 2 \]. Note that the packing fraction very close to the boundary is fixed because \( \rho = \rho_{max} \).

Dynamical behavior which allows the cluster radius to vary would suppress this instability, as bees moving from the core to the mantle would result in an expansion of the cluster, increasing the surface area and cooling the core. However, a dynamical model which allows the boundaries of the cluster to change requires a better understanding of the bee-level structure and the mechanics within a swarm cluster, and we leave this aside here.

1. **Methods for calculating linear stability**

Having solved for the equilibrium state, we want to find if this state is stable or unstable. To do so, we must define some dynamical laws for bee movement. Choosing a simple “behavioural pressure-taxis” behaviour, where \( J \propto -\rho \nabla P_b, \frac{d\rho}{dt} = -\nabla \cdot J \) gives us the complete set of dynamical equations:

\[
\begin{align*}
\rho \dot{T} &= \rho M(T) + \nabla \cdot (k(\rho) \nabla T) - C \mathbf{u} \cdot \nabla T = 0 \quad |_{\mathbf{r} \in \Omega} = \rho F[\rho, T], \\
\mathbf{u} &= \left[ ga \rho_{air} (T - T_a) \right] \hat{z} - \nabla P \cdot \kappa(\rho)/\eta \quad |_{\mathbf{r} \in \Omega}, \\
\nabla \cdot \mathbf{u} &= 0, \\
\mathbf{J} &= -J_0 \cdot \rho \nabla P_b, \\
\mathbf{J} &= 0 \quad |_{\mathbf{r} \in \partial \Omega}, \\
\frac{d\rho}{dt} &= -\nabla \mathbf{J} \quad |_{\mathbf{r} \in \Omega} = G[\rho, T]
\end{align*}
\]

Here we vary \( J_0 \) over a wide range to reflect the large variations in bee movement and temperature time scales, and define \( F, G \) to be the functionals that determine the dynamics of the system. We also emphasize some issues about these choices: a) Our form assumes a substrate that the bees move on. In reality, the bees \textit{are} the substrate, and changes on one side of the cluster propagate mechanically to other parts of the cluster at a rate faster than the taxis rate. b) Bee movement is not necessarily a local movement down pressure gradients. Bees can disconnect from the cluster and reattach at a different point, which doesn’t fit into the local gradient picture. c) There is a discontinuity in behavioural pressure and packing fraction at the boundary of the cluster, so this taxis model does not explain how the boundary of the cluster can change. Within these limits then, this choice of behaviour gives us a full set of differential equations. To determine if the equilibrium state of the system is linearly stable, we must determine
whether the linear response matrix:

\[
M = \begin{pmatrix}
\frac{dF}{dT} & \frac{dF}{dG} \\
\frac{dG}{dT} & \frac{dG}{d\rho}
\end{pmatrix}
\]

has positive eigenvalues. We note that the equilibrium temperature and bee packing fraction profile we solved for is symmetric with respect to rotation about the central axis (φ direction), and therefore we may partition the space of perturbations into subspaces defined by wave number \( k_{\phi} \):

\[
\Delta T(s, z, \phi) = e^{i k_{\phi} \phi} \Delta T(s, z), \quad \Delta \rho(s, z, \phi) = e^{i k_{\phi} \phi} \Delta \rho(s, z).
\]

These temperature and bee packing fraction perturbations will in turn give a change in airflow, pressure, and behavioural pressure:

\[
\Delta u(s, z, \phi) = e^{i k_{\phi} \phi} \Delta u(s, z), \quad \Delta P(s, z, \phi) = e^{i k_{\phi} \phi} \Delta P(s, z), \quad \Delta P_b(s, z, \phi) = e^{i k_{\phi} \phi} \Delta P_b(s, z).
\]

All of these will change the temperature and bee packing fraction time derivatives which will be proportional to \( e^{i k_{\phi} \phi} \). For each wave number \( k_{\phi} \), we construct the stability matrix and study its spectrum. To first order:

\[
\dot{T} = \left[ -u \cdot \nabla_s + \nabla_s \cdot k \nabla_s - k \cdot \left( \frac{k_{\phi}}{s} \right)^2 + \rho M_T \right] \Delta T + [M \Delta \rho + \nabla_s k \rho \nabla_s] \Delta \rho - \Delta u \cdot \nabla_s T
\]

\[
\dot{\rho} = \nabla_s \cdot (\rho \nabla_s \Delta P_b) - \left( \frac{k_{\phi}}{s} \right)^2 \rho \Delta P_b
\]

\[
\Delta P_b = \frac{dP_b}{d\rho} \Delta \rho + \frac{dP_b}{dT} \Delta T,
\]

where \( \nabla_s \) is the gradient in the \( s \) and \( z \) directions, \( M_T = \frac{dM(T)}{dT} \), \( k_{\rho} = \frac{d k_{\rho}}{d \rho} \). Regions where \( \rho = \rho_{\text{max}} \) are locked at \( \rho_{\text{max}} \) and not allowed to vary in bee packing fraction.

\[a. \text{ Solving for } \Delta u, \Delta P\]

The above equations depend on \( \Delta u \), which is determined by:

\[
\Delta u = \kappa [\Delta T \hat{z} - \kappa \nabla (\Delta P)] + \Delta \rho [(T - T_a) \hat{z} - \nabla P].
\]

\( \Delta u \) must have the form:

\[
\Delta u = \left[ \Delta u_{sz} - \kappa \nabla_{\phi} \Delta P_{\phi} \right] = \left[ \Delta u_{sz} - \kappa \frac{i k_{\phi}}{s} \Delta P_{\phi} \right],
\]

where \( \Delta u_{sz} \) is the \( sz \) component of \( \Delta u \). We solve for pressure \( \Delta P \) using the condition \( \nabla \cdot \Delta u = 0 \):

\[
\nabla \cdot \Delta u = \nabla_{sz} \cdot u_{sz} + \nabla_{\phi} \cdot u_{\phi} = \nabla_{sz} \cdot u_{sz} - \kappa \Delta P \left( \frac{i k_{\phi}}{s} \right)^2 = 0 \Rightarrow \Delta P \cdot \kappa \left( \frac{k_{\phi}}{s} \right)^2 = -\nabla_{sz} \cdot u_{sz}.
\]

At \( k_{\phi} = 0, \Delta u_{\phi} = 0 \), this condition simply becomes \( \nabla_{sz} \cdot u_{sz} = 0 \).

Therefore, at a given \( \Delta T, \Delta \rho \), we can solve \( \Delta u, \Delta P \) from this set of linear equations.

\[b. \text{ Numerical computation of linear response matrix}\]

To solve for stability, we discretize the system in the \( s, z \) directions in the same way that we did when solving for the equilibrium state. Because we are only solving for linear stability and the system starts off uniform in the \( \phi \) direction, we don’t need to discretize in the \( \phi \) direction. For perturbations at a certain wavenumber \( k_{\phi} \), the temperature derivative is, to first order:
\[
\frac{dT_{ij}}{dt} \rho_{ij} \cdot e^{-i k_{s} \phi} = M \left[ (T + \delta T)_{ij} \right] (\rho + \delta \rho)_{ij} - \\
\delta T_{ij} \left( \frac{k_{s}}{s_{ij}} \right)^{2} k(\rho_{ij}) - \frac{1}{V_{ij}} \sum_{(i',j')} A_{ij,i',j'} \cdot H[k(\rho_{ij} + \delta \rho_{ij}), k(\rho_{i',j'} + \delta \rho_{i',j'})] \cdot \left[ (T + \delta T)_{i',j'} - (T + \delta T)_{ij} \right] \\
\left[ \frac{1}{V_{ij}} \sum_{(i',j')} (u_{ij,i',j'} + \delta u_{ij,i',j'}) \cdot [-\theta \left( u_{ij,i',j'} + \delta u_{ij,i',j'} \right) \cdot (T_{ij} + \delta T_{ij}) + \theta \left( -[u_{ij,i',j'} + \delta u_{ij,i',j'}] \right) \cdot (T_{i',j'} + \delta T_{i',j'})] \\
\left[ (P + \delta P)_{ij} - (P + \delta P)_{i',j'} \right] + (z_{i',j'} - z_{ij}) \cdot \left( \frac{T_{ij} + \delta T_{ij} T_{i',j'} + \delta T_{i',j'} - go}{2} \right). \\
\delta P \text{ is set such that the divergence in the } \phi \text{ direction negates the divergence in the } s, z \text{ directions,} \\
\sum_{(i',j')} \delta u_{ij,i',j'} = \delta P_{ij} k(\rho_{ij}) w^{2} \left( \frac{k_{s}^{2}}{s_{ij}} \right) \\
\text{for all cells } (i, j). 
\]