Transoceanic migration network of dragonfly *Pantala flavescens*: origin, dispersal and timing

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The awe-inspiring multi-generational, transoceanic migration circuit of dragonfly species, *Pantala flavescens* stretches from India to Africa. Understanding the collective role of wind, precipitation, fuel, breeding, and life cycle driving the migration remains elusive. We identify the transoceanic migration route from years 2002 to 2007 by imposing an energetics-based time-constraint on a modified Dijkstra’s path-planning algorithm incorporating active wind compensation. The prevailing winds play a pivotal role; the Somali Jet enables migration across the Indian Ocean from Africa to India, whereas the return requires stopovers at the disappearing islands of the Maldives and Seychelles. The migration timing, identified using monthly-successful trajectories, life cycle, and precipitation data, corroborates sightings. A branched-network hypothesis connects our sighting in Cherrapunji (North-East India), the likely origin, to the known migration circuit.

**KEYWORDS**
transoceanic migration, *Pantala flavescens*, branched network, migration energetics, wind compensation, insect migration

1 Introduction

Every year millions of insects fly thousands of kilometers for conducive breeding habitats and foraging grounds (Williams, 1957; Holland et al., 2006; Dingle, 2014; Chapman et al., 2015). The sheer volume of the migration has tremendous consequences on the global ecology (Bauer and Hoye, 2014; Chapman et al., 2015; Hu et al., 2016). Migratory insects connect distant ecosystems across oceans, transport nutrients and propagules over long distances, and structure food webs (Bauer and Hoye, 2014; Hu et al., 2016). Although the consequences on the food cycle, distribution of nutrients, pollination, modulating disease-causing micro-organisms (Satterfield et al., 2020), panmixia (Trost et al., 2016) are of universal significance, predicting transoceanic route for insect migration has had limited success (Hedlund et al., 2021; Hobson et al., 2021). Identifying the migration route accurately could help understand gene flow and population dynamics of pests (Cao and Wu, 2019), disease outbreaks due to
parasites (Huestis et al., 2019) and locate the migrant carcass which drives seasonal transport of nutrients like nitrogen and phosphorus (Satterfield et al., 2020). Improving the prediction of the migration route and timing is a prerequisite to applying the concept of “migratory connectivity” to insects (Gao et al., 2020), assisting population management and pest control strategies and aiding full annual cycle studies (Marra et al., 2015). Furthermore, studies show the population of the migrating insects declining (Dirzo et al., 2014; Hallmann et al., 2017; Sánchez-Bayo and Wyckhuys, 2019). An increase in surface temperature shifts the overwintering sites away and stretches the migration range as in the migration of the Agrotes ipsilon (Zeng et al., 2020). Similarly, alterations in the microclimate of the stopover sites and origin have been conjectured to be the cause of declining populations of Pantala flavescens (Cao et al., 2018). Consequently, long-range migration becomes more strenuous, leading to ecological concerns regarding the arrival of migrant species that regulate the local vector population and outbreak of diseases (Huestis et al., 2019; Satterfield et al., 2020). Therefore, assessing the impact of climate change on the survivability of the migration relies intrinsically on route identification.

The annual migration circuit of dragonfly species, Pantala flavescens, is a multi-generational, transoceanic path (Anderson, 2009) spanning 14000–18000 km from India to Africa (Hobson et al., 2012). Intriguingly, P. flavescens crosses the Indian Ocean from Africa to India without stopovers with assistance from winds associated with the Inter-Tropical Convergence Zone (ITCZ); an extraordinary feat for an insect with wings a few inches wide (Anderson, 2009; Hedlund et al., 2021). Anderson (2009) proposed the route India-Maldives-Seychelles-Africa-India (Figure 1) based on his observations. Migratory routes range from simple round trips to complex circuits with merging and splitting branches (Satterfield et al., 2020) and the entirety of the transoceanic migration network of P. flavescens is a subject of multiple contemporary investigations (Borisov et al., 2020; Hedlund et al., 2021; Hobson et al., 2021). Our understanding of transoceanic migration of P. flavescens has been enhanced significantly by wind trajectory analysis (Chapman et al., 2012; Hedlund et al., 2021; Hu et al., 2021) that reveals how wind assists migration and stable isotope analysis (Hobson et al., 2012) that indicates probable origins of the migration circuit. Yet, almost a century later (Fraser, 1924) determining P. flavescens’s migration route still poses a challenge; there remain crucial but inexplicable observations (Hedlund et al., 2021). For instance, the arrival of P. flavescens at the Maldives (Anderson, 2009) and Seychelles (Bowler, 2003) which are sparsely situated islands on the migratory route during a specific period of the year. Similarly, Hobson et al. (2012) hypothesized that the migration originates in North and East India, while Anderson (2009) noted the arrival of P. flavescens in South-East India and Sri Lanka with retreating monsoons. In addition, a comprehensive understanding of how the life-cycle of the migrating dragonflies is linked to the migration timing as well as to breeding at stop-over locations is currently lacking. For instance, the literature indicates that there can be multiple possibilities once the migrants arrive in Maldives; they may breed (Olsvik and Hamalainen, 1992) or replenish energy reserves for the next transoceanic leg (Anderson, 2009). The destination of the next leg may either be Seychelles (Anderson, 2009) or perhaps mainland Africa (Hedlund et al., 2021). Thus, comprehending the complexity and the inherent uncertainties affecting the multigenerational transoceanic migration presents a unique challenge.

Field investigations are challenging in the absence of sophisticated lightweight trackers that can be employed over a large geographical extent for studying the migratory patterns of P. flavescens (Drake and Gatehouse, 1995; Holland et al., 2006; Hedlund et al., 2021; Hobson et al., 2021). Existing migration studies (Stefanescu et al., 2007; Chapman et al., 2008; Chapman et al., 2010; Chapman et al., 2016; Westbrook et al., 2016; Wu et al., 2018; Menz et al., 2022) successfully explain important features like heading and displacement; yet understanding transoceanic migration involves additional challenges. Predicting the migration timing of P. flavescens that is consistent with the precipitation, wind patterns at the reported altitude, and the sightings over the entire migration circuit is daunting for multiple reasons. A self-consistent approach must account for the energetics (Hedlund et al., 2021) of the migrant for estimating the endurance while predicting the wind-assisted trajectory for destinations situated on islands. The trajectory analyses by Hobson et al. (2021); Hedlund et al. (2021) assume that P. flavescens fly downwind; remaining on course during ocean crossings would require wind compensation, alongside wind assistance, to reach stopovers. We propose a complementary approach using an optimal path-planning algorithm (Dijkstra, 1959) applied to micro-aerial vehicles (MAVs) that allows wind compensation. The constraints for MAVs and P. flavescens, indeed, are analogous. The migrants must choose an altitude with favorable tailwind while suitably compensating for wind direction (Srygley, 2003) and fly on an intended path like MAVs. Further, the stopovers can only be on an island similar to waypoints in MAV missions. In addition, the stopovers must be located such that the migrant can reach the destination with limited energy reserves like MAVs. The combination of all these constraints makes the timing of migration the key to deciding the success of the migration. The optimal route does not constitute the migration trajectory per se; rather, migration occurs along a near-optimal path, determining which using trackers is nigh impossible.

**FIGURE 1**
Proposed migration route of Pantala flavescens (Anderson, 2009). The yellow dots indicate locations where Pantala flavescens have been reported. Source: http://www.discoverlife.org.
In the present study, we identify the optimal trajectory and verify the consistency of the timing for the entire migration with the reported observations. In order to determine the trajectory, we present an adapted energetics model (Pennycuick, 2008) for P. flavescens and find that the utmost duration of flight at the optimal velocity, \(4.5 \text{m/s}\), for maximizing range is 90 hours; an estimate consistent with Hedlund et al. (2021); Hobson et al. (2021). In our approach, the total energy expenditure is defined by the velocity at which the migrant is flying. We implicitly assume that the average power consumed in the entire flap-glide cycle, (comprising of a powered and a gliding phase) is identical to that of continuous flapping at the given velocity (Pennycuick, 2008). Thereafter, we apply Dijkstra’s algorithm (Dijkstra, 1959) to obtain the optimal path for minimizing time of flight under the influence of wind for the transoceanic legs of the migration. An optimal trajectory requiring a flight time less than 90 hours is considered successful. The migration time window is identified from the number of monthly-successful trajectories over a period of six years (2002–2007) (Anderson, 2009), life cycle and the local precipitation data and compared to sightings of P. flavescens for each stopover.

2 Materials and methods

2.1 Dragonfly energetics model

We designate the energetics model applied to P. flavescens as the Dragonfly Energetics Model (DEM). Dragonfly species P. flavescens cover large distances in a single flight during the trans-oceanic migration. The energetics involved in their migration though important has not yet been understood. However, there have been significant advances in bird migration theories, which have led to accurate calculations of energetics. In the present study, we develop a computational model for P. flavescens’s migration energetics by adapting the model proposed by Pennycuick (2008). The model is motivated by the long range transport aircrafts that involve energetics that are similar to the migration process, and is generic to any flying species, and has been applied to insects (Warfvinge et al., 2017) and birds (Pennycuick, 2008).

2.1.1 Energetics model for migration range, time of flight and maximum range velocity

The migration range and the time of flight are obtained by numerically integrating the instantaneous migration speed. Figure 2 shows an overview of the migration energetics model. The migration can last up to the point where the fat is completely burnt and defines the upper limit for the time of flight. The migration speed varies over time; for the major part of the course of migration, the migrant flies at a speed close to the maximum range speed (\(V_{mr}\)) that corresponds to the speed at which the lift to drag ratio is maximum. Before achieving \(V_{mr}\), the migrant flies instantaneously at a speed that fulfills the “constant specific work” (work done per unit mass) criterion. The criterion agrees with the observation that flight muscle fraction in migrants is nearly constant (Pennycuick, 2008). The constant specific work is obtained from the chemical power (\(P_{chem}\)), muscle mass (\(m_{mus}\)) and the wing beat frequency (\(f\)). The chemical power (\(P_{chem}\)) is obtained from the mechanical power (\(P_{mech}\)) required to fly and the metabolic power (\(P_{met}\)). Hence the energetics model involves two steps: Calculation of the power curve (the relationship between mechanical power and speed) and then migration range calculation using the power curve. We describe the associated parameters to compute the energetics model.

2.1.2 Power curve calculation

The required inputs for power curve calculation are three morphological parameters (mass (\(m\)), wing span (\(B\)), and wing area (\(S\)), gravity (\(g\)), and air density (\(\rho\)) at the desired height. The power curve is used to determine the total mechanical power (\(P_{mech}\)) required to maintain a horizontal flight, the minimum power speed (\(V_{mp}\)), and maximum range speed (\(V_{mr}\)). The minimum power speed (\(V_{mp}\)) is the speed at which the power required to fly is minimum. The maximum range speed (\(V_{mr}\)) can alternatively be viewed as the speed at which a flyer can cover the maximum distance per unit of fuel consumed. The total mechanical power required to fly at any particular speed consists primarily of three components of power; the induced power (\(P_{ind}\)), the parasite power (\(P_{par}\)), and the profile power (\(P_{pro}\)).

2.1.2.1 Induced power

The induced power is the rate at which the flight muscles of the insect have to provide work to impart downward momentum to the air at a rate that is sufficient to support the weight of the insect. The induced power is estimated using the actuator disc theory assuming a continuous beating of the wings as an actuator disc, and the pressure difference between the upper and the lower surface providing the aerodynamic force. The force multiplied by the induced power factor (\(k\)), which accounts for the loss in efficiency due to the real flapping of the wings, gives the real induced power (Eq. 1).

\[
P_{ind} = \frac{2k(mg)^2}{V_t\pi B^2 \rho}
\]

where \(V_t\) is true air speed.
2.1.2.2 Parasite power
The parasite power is the rate of work required to overcome the drag acting on the insect’s body, excluding the wings. The parasite power can be found from the drag acting on the body (Eq. 2),

\[ P_{par} = \frac{\rho V_m^2 S_b C_{D_b}}{2}, \tag{2} \]

where, \( S_b \) is the body frontal area, which is the maximum cross-sectional area of the insect, and \( C_{D_b} \) is the body drag coefficient.

2.1.2.3 Profile power
A flying insect needs profile power to overcome the drag acting on the wings and it is essentially a consequence of the induced and parasite powers. The profile power is estimated from the minimum of the sum of induced power (Eq. 1) and parasite power (Eq. 2) that is termed as the absolute minimum power \( P_{amen} \) (Eq. 3); the power required to fly at the minimum speed power \( V_m^{\text{amen}} \) (Eq. 4).

\[ P_{amen} = \frac{1.05k^3m^2g^{1/2}B^{3/4}C_{L_b}^{1/4}}{\rho^{3/2}g^{1/2}}, \tag{3} \]

\[ V_m^{\text{amen}} = \frac{0.807k^{1/4}m^{1/2}g^{1/2}}{\rho^{1/2}B^{1/2}S_b^{1/4}C_{L_b}^{1/4}}. \tag{4} \]

The profile power is then set at a fraction \( X_i \) of the absolute minimum power \( P_{amen} \). Here \( X_i = \frac{C_{pro}}{C_{par}} \) and profile power, \( P_{prof} \equiv X_iP_{amen} \). The total mechanical power, \( P_{mech} \), is then given by the summation of individual powers,

\[ P_{mech} \equiv P_{ind} + P_{par} + P_{prof} \tag{5} \]

To maximize the range of migration the maximum range speed, \( V_m^{\text{max}} \), needs to be determined from the power curve. The speed \( V_m^{\text{max}} \) is obtained by drawing a tangent from the origin to the power curve (Pennycook, 2008). The power curve (\( P_{mech} \) versus velocity) is calculated over a range extending from the minimum power speed \( V_m^{\text{amen}} \) (Eq. 4) to the maximum range speed \( V_m^{\text{max}} \).

2.1.3 Migration range calculation
The migration range is obtained by solving the range equation \( dY/dt = V \) where \( Y \) is the distance covered from the source and \( V \) is the instantaneous migration speed. The instantaneous speed, \( V \), varies over time because the mass of the insect changes as a function of time due to burning of fat and protein; consequently the aerodynamic and morphological parameters evolve over time. The model incorporates these changes; the rate at which fuel burns depends on the chemical power \( P_{chem} \), the lift to drag ratio \( (N) \) and the wingbeat frequency \( (f) \). We initialize the migration range calculation with values obtained from the insect’s measured morphological (for instance see Supplementary Material Table S5) and external parameters (p.g).

2.1.3.1 Chemical power
Chemical power is expended by an insect by burning fuel to generate the required mechanical power and support its metabolism. To determine chemical power \( P_{chem} \) (Eq. 6); the mechanical power \( P_{mech} \), basal metabolic rate \( P_{bmr} \), conversion efficiency \( (\eta) \), and respiration ratio \( (R) \) are incorporated. During flight, apart from efforts to maintain flight, insects need to maintain their metabolism at a rate higher than that required at rest (basal metabolic rate). The respiration ratio \( (R) \) accounts for the increase in metabolism due to continuous flight. Only a fraction of the chemical power expended is converted into mechanical power, which is accounted for by the conversion efficiency \( (\eta) \). Combining all these factors final expression for chemical power is given by:

\[ P_{chem} = \frac{R(P_{mech} + P_{bmr})}{\eta}, \tag{6} \]

where the metabolic power \( P_{bmr} \) is defined as \( P_{bmr} = \frac{\eta P_{bmr}}{\eta} \). Here \( P_{bmr} \equiv 3.79(m^{2/3})^3 \). Here \( m_{fat} \) is the fat mass and \( m_{fat} = 0.35m \) at \( t = 0 \).

2.1.3.2 Lift to drag ratio
Lift to drag ratio is a measure of distance covered by the insect per unit fuel energy consumed and can be related to the chemical power (Eq. 7).

\[ N = \frac{mgV_t}{\eta P_{chem}} \tag{7} \]

2.1.3.3 Wingbeat frequency
The wingbeat frequency (wingbeats per second) is a measure of power available from flight muscles, that is generated by the contraction and expansion of the muscle during flapping. Based on dimensional analysis the wingbeat frequency is correlated to the body mass \( (m) \), wing span \( (B) \), wing area \( (S) \), air density \( (\rho) \) and gravity \( (g) \). To some degree, the wingbeat frequency is under the control of the insect, but usually it does not vary much from the natural wingbeat frequency, which is determined by physics of beating wings (Eq. 8).

\[ f = \frac{m^{1/3}g^{1/2}B^{-23/24}S^{-1/3}}{\rho^{3/8}} \tag{8} \]

2.1.3.4 Time marching computation
A MATLAB code was developed for the numerical solution of the energetics model. The time marching computation is performed to compute the morphological parameters and flight parameters at time intervals of 6 minutes. In order to obtain the instantaneous mass of the insect \( m \), we require the rate of mass burnt, \( dm/dt \). It is obtained using the mass burning relation, \( dm/dt = P_{chem}/\epsilon \) where \( \epsilon \) is the energy density of the fuel. We assume that the insect obtains the 95% of the chemical power by burning fat and the rest from muscle mass consisting of protein. Therefore, \( m_{fat} \) is updated after each time step till \( m_{fat} \geq 0 \). A constant specific work criterion (Eq. 9) was used to compute muscle-burning rate.

\[ q_m = \frac{P_{mech}}{\frac{m_{mus}}{1 - \zeta} f} \tag{9} \]

where, \( m_{mus} \) is flight muscle mass and \( \zeta \) is Volume fraction of mitochondria in flight muscles.

Substituting the expressions for \( P_{mech} \) (Eq. 5), wingbeat frequency (Eq. 8) and \( m_{mus} = 0.15m \) a fourth degree polynomial for specific
work \( (q_m) \) is obtained in terms of instantaneous migration velocity. The specific work is computed at \( t = 0 \) and set as a constant for the rest of the time marching procedure to solve for the migration velocity until \( V \leq V_{mr} \); thereafter the migration occurs at \( V_{mr} \).

The flowchart in Figure 3 shows the algorithm used for migration range calculations. The model is validated with the results of Pennycuick (2008) for the bird Great Knot (see Supplementary Material Section 1).

### 2.1.4 Energetics of *Pantala flavescens*

We captured dragonflies at the IIT Kharagpur campus to determine the input parameters for the DEM. We measured morphological parameters (mass, wing span, wing area, and frontal area), using weighing balance and vernier (see further details in SI Table S4; Figure S2). Other relevant data like, aspect ratio \( (AR = B^2/S) \), fat mass \( (mfat = 0.35 m) \), and muscle mass \( (mmusc = 0.15 m) \) were extracted using these parameters. Migratory dragonflies like *Anax junius* (Wikelski et al., 2006; May and Matthews, 2008) and monarch butterfly (Brower et al., 2006) have an estimated fat content in the range 27–33%. Furthermore, long-distance migrants like monarch butterflies and birds (Dingle, 2014) as well as swarming migratory locusts store greater fat content (Du et al., 2022). Hence we have assumed that fat content is 35% for *Pantala flavescens*. Profile power constant and induced power factor for the calculations are identical to that of birds, as they were found to be satisfactorily applicable by May (1991) and Azuma and Watanabe (1988) in their calculations of the power curve for dragonflies and similar U-shaped power curve has been estimated for large insects like moths as well (Warfvinge et al., 2017). The body drag co-efficient \( C_{db} \) was obtained from May (1991) for dragonflies with similar mass. The DEM parameters and results are detailed in the Supplementary Material Section 2.

### 2.2 Dragonfly Path planning model (DPM)

Dijkstra’s algorithm (Dijkstra, 1959) was used to evaluate the role of wind in the migration of *P. flavescens*. The algorithm searches for the shortest path between the starting and the end point, given that at least one such path exists. We marked the stopovers as the starting and the end points. An open-source MATLAB code of Dijkstra’s algorithm (Aryo, 2021) was developed further by modifying the cost matrix for computing the optimal path and the associated time. The combination of the cost matrix along with Dijkstra’s algorithm is designated as the dragonfly path planning model (DPM) here after. Figure 4 shows the flowchart for the DPM.

Either of the two optimization criteria, the time of flight and the distance covered, can be used for generating the cost matrix. However, the primary constraint during migration is the fuel reserve which places an upper limit on the time of flight, but the distance covered depends on the time of flight, the migration velocity of the insect and the local wind velocity. Therefore, the time of flight is a more fundamental constraint associated with the fuel reserve and hence chosen as the cost function function (Warfvinge et al., 2017).

Three key inputs are required for the generation of the cost matrix: the dragonfly migration velocity, which in this case is \( V_{mr} \) and is obtained from DEM; the local wind velocity obtained from NOAA (NOAA, 2019); the global position in terms of latitude and longitude. We select the latitude and the longitude of the starting and end points while initializing the DPM. Thereafter the rectangular area between the starting and end point, with the diagonal as the geodesic distance between the two points, is discretized with 30 grid-points in each direction. The wind data from NOAA is available at various spatial and temporal resolutions.
MATLAB using "velocity data from NOAA was used to create an interpolant in particular day (NOAA, 2019) for the DPM simulations. The internal node (see Figure 5A); the boundary nodes have fewer directions (track). We consider eight possible directions for each any two nodes is calculated. Also, based on dragon directions. Based on the latitude and longitude distance between longitude, dragon points. A separate discretization grid is used for each transoceanic leg of the migration.

Spatial resolution of the data is 2.5°×2.5° on the global grid (144×73) for longitudes 0.0° to 357.5°E, and latitudes 90.0°N to 90.0°S. The data is available at 17 Pressure levels between 1000 and 10 hpa. The temporal resolutions of the available data are 4-times daily, daily mean and monthly mean values. Pantala flavescens has been reported to migrate at heights above 1000m (Anderson, 2009), and the closest corresponding pressure level in the available dataset is 850hpa. We used "daily mean" of "U-wind" and "V-wind" with "multiple pressure level" data file corresponding to 850hpa for a particular day (NOAA, 2019) for the DPM simulations. The velocity data from NOAA was used to create an interpolant in MATLAB using "griddata" to obtain the velocity field on the grid points. A separate discretization grid is used for each transoceanic leg of the migration.

Each node of the grid is assigned wind velocity, latitude and longitude, dragonfly migration velocity, and possible flight directions (track). We consider eight possible directions for each internal node (see Figure 5A); the boundary nodes have fewer directions. Based on the latitude and longitude distance between any two nodes is calculated. Also, based on dragonfly migration speed ($V_d$), local wind velocity, and the track, a resultant velocity between two neighboring nodes is computed. The resultant velocity and the distance determine the time taken to travel between the two nodes that serves as the cost function between any two nodes. We compute the cost matrix using all possible combinations of nodes; that is the time taken to travel for each possible route constitutes the associated entry of the cost matrix. Using the cost matrix in Dijkstra’s algorithm (Dijkstra, 1959), we calculate the optimal time and the optimal route for each leg of the transoceanic migration. We also calculate the total distance and the fuel consumed from the optimal route.

Here $V_w$ is the magnitude of wind velocity calculated from two planar components, $u$-wind ($u$ component of wind velocity, positive in due east) and $v$-wind ($v$ component of wind velocity, positive in due north). The magnitude of dragonfly velocity is $V_f$ (we assume $V_f=V_{max}$), and $V_t$ is the magnitude of resultant velocity, $\theta$ is wind direction, $\alpha$ is dragonfly track, that is the direction relative to the ground and is fixed by the grid (see Figure 5A). Here $\delta$ is dragonfly heading: the direction relative to the wind field that is required to maintain the track. All angles are measured with respect to due east (see Figure 5B).

2.3 Passive tracer trajectory

We simulate the trajectory of a migrating dragonfly species, Pantala flavescens, under the influence of the atmospheric wind field as if it acts as a passive tracer particle and gets purely convected by the wind field. We have developed a MATLAB script for solving the equations of motion (Eq. 10) using the modified Euler method (further details of the numerical method are provided in Pozrikidis (2016); Giri et al. (2022)). The velocity and grid used for passive tracer simulation is the same as for the DPM simulation. The difference in the two simulations is that, the end location is fixed in the DPM simulation, and we calculate the path and the time required to reach that location. Whereas the time of flight is fixed (90 hours) for the passive tracer simulations, and we obtain the end location by time marching at an interval of $\Delta t = 60s$. The time resolution $\Delta t$ is small enough to ensure that the results are converged. The particle at any instant assumes the local wind velocity $(u_t, u_v)$ and the velocity of the particle is updated at each time step ($\Delta t = 60s$) after advancing the

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particle location $X(t)$, $Y(t)$ in time.

$$\frac{dX}{dt} = u_x(X(t), Y(t), t); \frac{dY}{dt} = u_y(X(t), Y(t), t)$$  \hspace{1cm} (10)

3 Results and discussion

3.1 Transoceanic migration route reveals that *Pantala flavescens* actively compensate for wind

Figure 6 shows the optimal path for the transoceanic legs of the *P. flavescens*’s migration circuit and corroborates existing observations (Fraser, 1924; Corbet, 1962; Mitra, 1974; Anderson, 2009; Hobson et al., 2012; Hedlund et al., 2021). The flight from Somalia (Location 5) to India (Location 1) is the longest leg (2802 kms) but requires merely 45 h and no stopovers because of strong tailwinds (~ 15 m/s). The migration from India (Location 1) to Mozambique (Location 4) is more precarious and involves stopovers at Locations 2 (the Maldives) and 3 (Seychelles). The trajectory from India to the Maldives is tortuous and requires 48 h for 1137 kms only. The stark difference between the time required and the distance covered in the two optimal paths highlights the pivotal role of wind. Further, comparing the optimal path with the trajectory of a passive tracer reveals that active wind compensation is imperative for *P. flavescens* to cross the Indian ocean; a passive tracer convecting downwind strays from the optimal paths (see Figures 6C–E). Previous studies have provided evidence of similar active wind compensation during long distance migration by hawkmoths, butterflies, dragonflies and birds (Gibo, 1986; Srygley et al., 1996; Srygley and Oliveira, 2001; Srygley and Dudley, 2008;...
Tarroux et al., 2016; Wynn et al., 2020; Menz et al., 2022). The flight from the Maldives to Seychelles (2565 kms) and the subsequent flight from Seychelles to Mozambique (2056 kms) are even more critical in the success of the migration because the time required is 90h and 81h respectively, nearly extinguishing their energy reserves. Therefore, we deduce that *P. flavescens* actively perform wind compensation (Srygley, 2003; Chapman et al., 2010) while flying at a minimum pace (4.5 m/s) to maximize their range (Hedlund et al., 2021).

### 3.2 Precipitation data, life cycle estimates combined with monthly successful trajectories predict the migration timing window

*P. flavescens* are obligate migrants, whose migration timing shadows the movement of the Inter-Tropical Convergence Zone, pursuing evanescent pools for breeding (Anderson, 2009). Indeed, their sightings at the stopovers reported in the literature (see Table 1) and the months of high precipitation (see Figure 7) approximately coincide. Therefore, migration occurs *only when* both wind assistance and rainfall are available simultaneously, creating the migration time window. We present the total monthly successful trajectories for migration. The arrival of *P. flavescens* in June–July, aided by the Somali Jet, is well documented (Corbet, 2004; Anderson, 2009; May, 2013). Thereafter, the migrants breed and the offsprings typically emerge after 45–60 days (Kumar, 1984; Corbet, 2004; Suhling et al., 2004). Furthermore, Figure 7B reveals that the wind assistance from India to the Maldives is available except from June to August. The ITCZ passes over southern India and the Maldives in October, bringing rainfall, and the onset of migration from India to the Maldives lasting till December. Upon arrival in the Maldives the migrants either breed (Olsvik and Hamalainen, 1992) in evanescent pools created by rainfall (Corbet, 2004) or migrate towards Seychelles (Campion, 1913; Bowler, 2003; Anderson, 2009) to breed (Wain et al., 1999). Breeding in the Maldives might be preferable as crossing the Indian Ocean to reach Seychelles between October and December, although not impossible, remains unfavorable (see Figure 7C). Thereafter the precipitation in and migration to Seychelles is more favorable from January to March, corroborating sightings (Samways, 1998; Samways et al., 2010; Seychelles Bird Records Committee, 2014; Alphonse-Island, 2016). The onward journey to the African mainland is also favored by the wind and precipitation in the period December to March, as seen in Figure 7D. Indeed, sightings from Mozambique and Malawi, South Africa, and Lake Tanganyika support our findings (see Table 1). *Pantala flavescens* may breed twice on the African mainland before migrating to India, thus spawning 4–5 generations every year (Corbet, 2004).

### 3.3 Fresh sighting links the origin of transoceanic migration to branched migration network that provides clue to widespread dispersal

Existing studies (Anderson, 2009; Borisov et al., 2020; Hedlund et al., 2021) suggest an alternate route for *P. flavescens* to cross the Indian Ocean directly from various departure sites in India and the Maldives to arrive in Somalia during November and December. We computed the successful trajectories for emigration from Locations 1 (India), 2 (Maldives) to Location 5 (Somalia), and indeed winds are not favorable until September (see Figure 8A), and most successful trajectories occur in December, consistent with Hedlund et al. (2021). The passage of the ITCZ in October through this region renders the direct crossing feasible. November seems more favorable for migration than December because there is more precipitation in Somalia.

The possibility of the alternate route implies the existence of branching networks (Drake and Gatehouse, 1995; Satterfield et al., 2020). The concept of a branched network allows a complex migratory pattern to emerge naturally and connect unexplained albeit important observations reported hitherto. For instance, *P. flavescens* arrive in South-Eastern India and eastern Sri Lanka along with the NE monsoons from October to December (Corbet, 1988; Anderson, 2009). Furthermore, the origin of *P. flavescens* reaching the

### TABLE 1 Sighting of Pantala flavescens at different locations and the timing reported in the literature.

| Place                  | Reference                        | Months (Remarks)         |
|------------------------|----------------------------------|--------------------------|
| Maldives               | Anderson, 2009                   | Oct–Dec                  |
|                        | Olsvik and Hamalainen, 1992      | Nov (Mating)             |
| Seychelles             | Bowler, 2003                     | Nov                      |
|                        | SBRC                             | Dec–Jan                  |
|                        | alphonse-island.com              | Mar                      |
|                        | Wain et al., 1999                | Nov (Breeding)           |
|                        | Campion, 1913                    | Nov (Location: Aldabra)  |
|                        | Samways, 1998; Samways et al., 2010 | Nov–Apr              |
| South Africa           | Samways and Oshorn, 1998         | Dec–Feb (breeding)       |
| Mozambique/ Malawi     | Dijkstra, 2004                   | Nov and Jan              |
|                        | Bernard and Bąkowski, 2020       | Dec–Mar–Apr              |
| Tanganyika             | Bartenef, 1931                   | Dec–Jan                  |
| Uganda                 | Bartenef, 1931                   | Mar–Apr, Oct             |
| India                  | Fraser, 1924                     | Sep–Nov (Departure on “annual migration”) |
| Amsterdam Island       | Devaud and Lebouvier, 2019       | Feb                      |
| Chagos Archipalegro    | Carr, 2022                       | Oct–Nov                  |
Maldives is speculated as North-Eastern India (Hobson et al., 2012; Hedlund et al., 2021). We sighted *P. flavescens* swarms in Cherrapunji (25.2N 91.7E, NE India (S1)) on 1st and 2nd November 2019 (see Figure 8B). The swarm exhibited a well-coordinated motion predominantly from the north-east to the south-west direction (see Supplementary Movie S1) aligned with the local wind on those days, indicating a destination potentially in South-Eastern India or Sri Lanka. A branched migration network coupled with the sighting in Cherrapunji allows us to conjecture that, indeed, the transoceanic migration of *P. flavescens* plausibly originates in NE India with stopovers in SE India and Sri Lanka. We investigate the branched networks on this pathway further using optimal paths (see Figure 8C; Table 2). A direct flight from Cherrapunji to the Maldives (S6) requires the migrant to cover a distance of around 3000 km, requiring approximately 115 h, which is significantly higher than the threshold of 90 h. There is ample land mass between these two sites, and multiple refuelling stopovers may be anticipated. Figure 8C shows the branching network of various possibilities of reaching S6 from S1 with stopovers at Visakhapatnam (S2), Mangalore (S3), Thiruvananthapuram (S4) and in Sri Lanka (S5). From the figure, it can be observed that S2 and S4 are preferable stopovers as they are in the path of multiple optimal routes, and they are more likely to be selected by migrating *P. flavescens*. The timely sightings at Cherrapunji, SE India and Sri Lanka and the branched networks revealed by the optimal paths lend further credibility to the NE India being the origin of the transoceanic migration of the *P. flavescens*.

The branched network in Figure 8C perhaps provides a glimpse of the complex migratory network of *P. flavescens*, potentially spanning Asia and Africa. The appearance of *P. flavescens* in Japan, China, Indonesia, Sri Lanka, NE India, and southern Islands of the Indian ocean such as Amsterdam Island and Chagos Archipelago (see Figure 1; Table 1) prompts us to speculate that branching and dispersal of migrating *P. flavescens* emerge from all the locations that are part of a more complex migratory circuit spanning Asia, Africa and beyond (Sparrow et al., 2020). The migration significantly impacts global ecology, and its success is linked to any stressors of the climate and local ecological systems. There have been reports of islands of the Maldives (Yamamoto and Esteban, 2010) and Seychelles (Obura et al., 2022) disappearing which can be detrimental to the migration of *P. flavescens* and, in turn, to the global ecology (Liao et al., 2023).

4 Conclusions

In conclusion, we found that the migration from India to Africa commences from October with stopovers in the Maldives and Seychelles, as suggested by Anderson (2009). The arrival timing...
predictions for the Maldives, Seychelles, and Mozambique agree with prior observations (Campion, 1913; Olsvik and Hamalainen, 1992; Wain et al., 1999; Anderson, 2009) (see Table 1). The migration time window indicates the possibility of breeding in the Maldives and Seychelles, that is consistent with Olsvik and Hamalainen (1992)’s and Wain et al. (1999)’s observations, respectively. The direct crossing of the Indian Ocean aided by the Somali Jet is feasible, May onwards, during the migration from Africa to India, corroborating existing reports (Corbet, 2004; Anderson, 2009; Hedlund et al., 2021). An alternate route involving the direct crossing of the Indian Ocean for the onward journey from India and the Maldives to the African continent is also feasible in November and December, as reported by Hedlund et al. (2021). Furthermore, the identification of alternate routes implies the existence of a branched and complex network of migratory routes. We also correlated our spotting of swarms of migrant P. flavescens in Cherrapunji, India on 1st, 2nd November 2019 with two seemingly disconnected results in the literature; Hobson et al. (2012); Hedlund et al. (2021) predicted that the migration of P. flavescens begins in NE India and Corbet (1988); Anderson (2009) reported the arrival of P. flavescens in SE India and Sri Lanka. Indeed all the observations fit into an extensive, branched migration circuit of P. flavescens that originates in NE India and has stopovers spread across the Indian subcontinent and potentially explains the closeness of the local populations (Christudhas and Mathai, 2014). A branched migratory pattern with multiple stopovers provides a clue to the widespread dispersal of P. flavescens throughout SE Asia and Africa (see Figure 1; Table 1). P. flavescens initiate migration from various sites and probably halt at stopovers when they find cues for favorable conditions. Furthermore, the availability of micro-insects (Anderson, 2009) and aerial planktons (May, 2013; Troast et al., 2016) for feeding

### Table 2: Time to cover different section of the branching network.

| Section | Time (hr) | Section | Time (hr) |
|---------|-----------|---------|-----------|
| S1–S6  | 115       | S2–S6  | 66        |
| S1–S2  | 52        | S3–S6  | 43        |
| S1–S3  | 88        | S4–S6  | 31        |
| S1–S4  | 85        | S5–S6  | 49        |

Various possibilities of reaching the Maldives (S6) from Cherrapunji (S1) with stopovers at Visakhapatnam [17.5N 82.5E] (S2), Mangalore [13N 75E] (S3), Thiruvananthapuram [9N 77E] (S4) and in Sri Lanka [9N 81E] (S5).
en route could increase the continuous flight duration and, consequently, the range. The migration in swarms is also likely to reduce the aerodynamic drag on each individual. Thus, the energetics constraints imposed in our model are conservative and underestimate the probability of migration success. Our conclusions raise questions like the importance of the swarm dynamics in insect migration. Furthermore, the impact of the gradual disappearance of the islands of the Maldives (Yamamoto and Esteban, 2010) and Seychelles (Obura et al., 2022) on the survival of the transoceanic migration of *P. flaveiscens* and the consequences on the global ecology have perhaps remained underappreciated so far.

### Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

### Author contributions

The study was conceived and designed by KR and SS. Field work and experimental data collection were done by KR and AP. Codes were written by KR and SS. Simulations and data analysis were done by KR. KR and SS wrote the paper. Valuable insights, feedback, and inputs were provided by AR throughout the work and manuscript drafting. The final work and paper have been shaped by ideas and feedback from all the authors. All authors contributed to the article and approved the submitted version.

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### Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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### Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fevo.2023.1152384/full#supplementary-material
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