Testosterone-mediated behavior shapes the emergent properties of social networks

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16   **Abstract**

1. Social networks can vary in their organization and dynamics, with implications for
18   ecological and evolutionary processes. Understanding the mechanisms that drive social
    network dynamics requires integrating individual-level biology with comparisons across
20   multiple social networks.

2. Testosterone is a key mediator of vertebrate social behavior and can influence how
22   individuals interact with social partners. Although the effects of testosterone on
    individual behavior are well established, no study has examined whether hormone-
24   mediated behavior can scale up to shape the emergent properties of social networks.

3. We investigated the relationship between testosterone and social network dynamics in the
26   wire-tailed manakin, a lekking bird species in which male-male social interactions form
    complex social networks. We used an automated proximity system to longitudinally
28   monitor several leks and we quantified the social network structure at each lek. Our
    analysis examines three emergent properties of the networks: social specialization (the
30   extent to which a network is partitioned into exclusive partnerships), network stability
    (the overall persistence of partnerships through time), and behavioral assortment (the
32   tendency for like to associate with like). All three properties are expected to promote the
    evolution of cooperation. As the predictor, we analyzed the collective testosterone of
34   males within each social network.

4. Social networks that were composed of high-testosterone dominant males were less
36   specialized, less stable, and had more negative behavioral assortment, after accounting
    for other factors. These results support our main hypothesis that individual-level hormone
38   physiology can predict group-level network dynamics. We also observed that larger leks

with more interacting individuals had more positive behavioral assortment, suggesting

40   that small groups may constrain the processes of homophily and behavior-matching.

    5. Overall, these results provide evidence that hormone-mediated behavior can shape the

42   broader architecture of social groups. Groups with high average testosterone exhibit

      social network properties that are predicted to impede the evolution of cooperation.

Keywords: androgens, behavioral endocrinology, collective behavior, cooperation, dynamic

46 networks, social networks, testosterone

Introduction

48   Behavioral interactions are the foundation of social network structures that can vary through

time, among populations, and across species. Network structures play an important role in many

50   ecological and evolutionary processes, including the spread of diseases (Sah *et al.* 2018;

Stroeymeyt *et al.* 2018), the transmission of information and resources (Aplin *et al.* 2012;

52   Maldonado-Chaparro *et al.* 2018), and selection on individual behavior (Ohtsuki *et al.* 2006). A

major challenge is understanding how individual-level factors, such as physiological and

54   behavioral mechanisms, scale up to drive the emergent structural properties of social groups

(Krause & Ruxton 2002). Linking these two levels of analysis is difficult because it requires

56   integrating individual-level data with repeated measures of entire social groups (Sah *et al.* 2017).

Here, we use a comparison of multiple social networks through time to investigate how

58   hormone-mediated behavior shapes the higher-order structure of social networks. Testosterone is

a steroid hormone that is well known for its influence on social behavior and its sensitivity to

60   changes in the social environment (Wingfield *et al.* 1990; Adkins-Regan 2005; Goymann 2009).

Testosterone often promotes physical aggression and other behaviors associated with social

62   dominance (Oyegbile & Marler 2005; Fuxjager *et al.* 2010). Testosterone can also promote

status-seeking behaviors in a non-aggressive context, including cooperative and gregarious

64   behavior (Eisenegger *et al.* 2011; Boksem *et al.* 2013; Ryder *et al.* 2020). Overall, these

hormone-signaling pathways are essential for the development and modulation of complex

66   behavioral phenotypes (Cohen *et al.* 2012).

To investigate how testosterone is associated with social network dynamics, we studied

68   the wire-tailed manakin (*Pipra filicauda*), a bird species in which the males engage in

coordinated displays with each other at sites known as leks. In *P. filicauda*, males exhibit two

70   social status classes: the dominant males who hold territories on the leks, and subordinate,

"floater" males who must acquire a territory on a lek before they can mate (Heindl 2002; Ryder

72   *et al.* 2011a). Male social partnerships in wire-tailed manakins can be remarkably stable and

typically occur between two unrelated males, most commonly, but not exclusively, between a

74   territory-owner and a floater (Ryder *et al.* 2011a; Dakin & Ryder 2020). Two features of the

manakin leks make them especially well-suited to studying the relationship between hormones

76   and group-level social structure. First, testosterone is known to affect the social behavior of

individual male wire-tailed manakins (Ryder *et al.* 2020). Second, these partnerships among

78   males form the basis of complex and dynamic social networks that are replicated across leks,

facilitating a comparative approach (Dakin & Ryder 2018, 2020). Given this background, we

80   sought to test whether testosterone and its effects on male behavioral phenotype could drive the

emergent properties of the social network.

82   The broader function of male-male social behavior and coordinated displays in the

manakin family has been the subject of considerable study (e.g., Prum 1994; DuVal 2007;

84   McDonald 2007; Ryder *et al.* 2008, 2009; Díaz-Muñoz *et al.* 2014). One possible function is that

dyadic displays may be competitive and/or they may serve to maintain an individual's position in

86   a dominance hierarchy (Prum 1994; Heindl 2002). Social interactions may also represent

cooperative coalitions that provide benefits to both parties (Ryder *et al.* 2008, 2009). The

88   potential competitive and cooperative functions of male-male social behavior are not mutually

exclusive, and function may be context- and/or status-specific (Ryder *et al.* 2008). Coordinated

90   displays may also be a vestige of ancestral cooperative behavior (i.e., the behavior may have

been directly beneficial to both parties in the past, and it persists today, even if it no longer has

92   adaptive benefits; Prum 1994). Given this background and recent evidence that testosterone

modulates male social behavior in wire-tailed manakins (Ryder *et al.* 2020), we focused this

94 study on three emergent properties of the social networks formed by male-male interactions that

can influence the evolution and maintenance of cooperation (Fig. 1).

96 The first property we examined, social specialization, seeks to capture the exclusivity of

the relationships between partners in a social network. In the context of manakin behavior,

98 highly specialized networks are well-partitioned among specific territory-owner and floater

relationships, as illustrated in Fig. 1A. In humans, social specialization has been found to

100 maximize the ability of a team to successfully perform a challenging task (Jehn & Shah 1997). In

manakins, we expect specialization to improve the familiarity of social partners and the

102 behavioral coordination of their displays. Greater specialization is also expected to minimize

conflict over mating and territorial ascension opportunities (Schjelderup-Ebbe 1922; McDonald

104 1993).

The second property, network stability (Fig. 1B), quantifies the average persistence of

106 social partnerships through time (Poisot *et al.* 2012; Dakin & Ryder 2020). Coordinated displays

in manakins require the synchronization of complex behaviors, and previous empirical work

108 indicates that longer partnership tenure has a positive effect on display coordination (Trainer &

McDonald 1995; Trainer *et al.* 2002). Greater temporal stability of partnerships also increases

110 the opportunity for familiarity and reciprocity within a social network (Trivers 1971; Roberts &

Sherratt 1998; Croft *et al.* 2006).

112 The third property, behavioral assortment (Fig. 1C), captures the extent to which males

interact with partners who express similar behaviors (i.e., is like associated with like? Croft *et al.*

114 2006; Farine 2014). At the proximate level, positive assortment may represent the outcome of

generalized reciprocity and/or partner choice (Fowler & Christakis 2010; Dakin & Ryder 2018).

116 At the ultimate level, positive assortment has also been shown to promote the evolution of

cooperation (Ohtsuki *et al.* 2006). To quantify the overall behavioral assortment of the manakin

118 networks, we focused on two correlated metrics of social behavior within the network: "strength"

(a male's frequency of daily social interactions) and "degree" (his daily number of social

120 partnerships; Dakin & Ryder 2018). We used a composite measure of assortment that averaged

the assortativity indices of these two phenotypes.

122  As a potential predictor of specialization, stability, and assortment, we quantified the

collective testosterone of the manakin leks (Fig. 2; Akinola *et al.* 2016). Because hormone-

124 behavior relationships are status-specific in manakins and many other species (Eisenegger *et al.*

2011; Boksem *et al.* 2013; Ryder *et al.* 2020), we analyzed collective testosterone for each of the

126 two status classes separately. High individual testosterone levels are associated with reduced

sociality in dominant males, but increased sociality in subordinate males (Ryder *et al.* 2020). We

128 therefore predicted that the average (or collective) testosterone of territorial males in a group

would be negatively associated with its specialization, stability, and assortment. In contrast, we

130 predicted that the collective testosterone of floater males in a group would be positively

associated with these three emergent properties of the social network.

Materials and Methods

134 ***Study Population***

We studied wire-tailed manakins (*Pipra filicauda*) at the Tiputini Biodiversity Station in

136 Orellana Province, Ecuador (0º 38' S, 76º 08' W). This population of *P. filicauda* has been

studied and individuals color-banded annually since 2002 (e.g., Ryder *et al.* 2008, 2009). The

138 present study was conducted on 11 leks during peak breeding activity (December-March) across

three field seasons: 2015-16, 2016-17 and 2017-18. All research was approved by the

140 Smithsonian ACUC (protocols #12-23, 14-25, and 17-11) and the Ecuadorean Ministry of the

Environment (MAE-DNB-CM-2015-0008).

Testosterone Assessment

144 Male manakins were captured using mist-nets on the leks as described in (Ryder *et al.* 2020). We

deployed up to 16 mist-nets simultaneously at a given lek, with the intention of capturing every

146 male on the lek. We rotated the nets between leks with the goal of capturing each male up to

three times per field season. Each mist-net was checked on a 30-minute schedule with variation

148 resulting from capture rate (i.e., multiple birds being caught on the same net run; Vernasco *et al.*

2019; Ryder *et al.* 2020). The amount of time a male spends in the net has a subtle, but

150 significant, negative effect on his circulating testosterone (Vernasco et al. 2019). Therefore, we

used video monitoring to determine the duration of time that each bird was in the mist-net prior

152 to blood sampling, so that it could be accounted for in further analyses (mean = 17.5 minutes, SD

= 10.3 minutes, range = 1–72 minutes). Following removal of a male from the mist-net, a small

154 blood sample (< 125uL) was taken from the brachial vein and stored on ice prior to being

centrifuged at 10,000 rpm for 5 min, as described in previous studies (Ryder *et al.* 2011b;

156 Vernasco *et al.* 2019; Ryder *et al.* 2020). Plasma volume was measured to the nearest 0.25 ul and

stored in 0.75 ml of 100% ethanol (Goymann *et al.* 2007). Testosterone was double extracted

158 from the plasma using dichloromethane. Following extraction, a direct radioimmunoassay was

used to measure the total plasma androgen concentration (ng/ml) adjusted by the extraction

160 efficiency and plasma volume of each sample (Eikenaar *et al.* 2011; Ryder *et al.* 2011b).

Hormone assays were conducted annually, and the detection limits were 0.12, 0.08, and 0.09

162   ng/ml for 2015-16, 2016-17 and 2017-18, respectively; any sample that fell below the assay-specific limit of detection was assigned that limit as its testosterone concentration as a most

164   conservative estimate. The extraction efficiency for all samples was between 62-73%, and the intra-assay coefficients of variation were 6.6%, 11.6%, and 9.2% for 2015-16, 2016-17 and

166   2017-18, respectively; the inter-assay coefficient of variation was 19.5%.

168   *Behavioral Recording*

We used an automated data-logging system to monitor male-male interactions on the display

territories of the leks (Ryder *et al.* 2012; Dakin & Ryder 2018; Ryder *et al.* 2020). The territories on these leks are specifically used for male-male coordinated social displays (as described in

Schwartz & Snow 1978). At the beginning of each field season, male manakins were outfitted with coded nano-tags (NTQB-2, Lotek Wireless; 0.35 g). The tags transmitted a unique VHF

signal ping once per 20 s for three months. In total, 296 tag deployments were performed on 180 individuals (mean 1.7 field seasons per male ± SD 0.7), 178 of whom also had hormone data

(mean number of hormone samples per male = 3 ± SD 1.5). Approximately 10 days (± SD 7) after tagging and sampling was completed at a given lek, a proximity data-logger (SRX-DL800,

Lotek Wireless) was deployed within each display territory at the lek to record tagged males within a detection radius of 30 m (a distance that corresponds to the typical diameter of a

manakin display territory; Heindl 2002; Dakin & Ryder 2018).

    Proximity recording sessions ran from 06:00 to 16:00 for ~6 consecutive days (± SD 1

182   day) and were performed ~3 times per field season at a given lek. Occasionally, the length of a recording session was extended due to extenuating circumstances such as inclement weather.

184   The recording sessions were scheduled to be distributed evenly throughout each field season at

each lek, to minimize any confounding seasonal effects. Each recording session represents an

observation of the social network at a given lek. In total, we conducted 86 recording sessions

(29,760 data-logger hours) representing repeated measures of the social activity at 11 leks during

three field seasons (see Fig. S1 in the supplement for additional details on the sampling regime).

Prior to data-logger deployment, each territory was also observed on non-recording days

to identify the territory-holder based on his color-bands, following previous studies (Ryder *et al.*

2008, 2009). These status assignments were subsequently verified in the proximity data.

Data Processing and Statistical Analysis

All computational and statistical analyses were performed in R (R Core Team 2018). Network

illustrations were made using the igraph package (Csardi *et al.* 2018).

Social Interactions

Social networks were constructed by first defining interactions between two males that occurred

on the display territories. To do this, a computational algorithm was used to identify joint

detections, wherein two males were located at the same display territory within a pre-defined

spatial and temporal threshold (Ryder *et al.* 2008, 2012; Dakin & Ryder 2018). For the temporal

threshold, the two males had to occur < 45 s apart. This temporal threshold was chosen to allow

for the fact that each tag pinged with a 20 second pulse rate, such that overlapping individuals

could have up to a 40 s gap between their respective pings. For the spatial threshold, the two

males had to have a difference in received signal strength values (Δ RSSI) < 10. This threshold

corresponded to a typical distance between 0 to 5 m apart in a ground-truthing experiment

(Dakin & Ryder 2018). Hence, according to our definition, a social interaction is initiated only

208 after males come within this approximate spatial threshold. We chose this spatial threshold

because it is close enough to permit visual and acoustic contact during typical social behaviors

210 (such as those described in Schwartz & Snow 1978). After completing our study, we also

performed a sensitivity analysis to verify that our main results were robust to alternative spatial

212 threshold definitions (see supplement for details).

Any repeated co-occurrence of the two males within 5 minutes was considered to be part

214 of the same social interaction, but after a gap of ≥ 5 minutes, it was considered to be a new

interaction between those two males. The average duration of social interactions defined by this

216 method was 5.1 min ± SD 13.3, 8.8 min ± SD 20.1, and 7.0 min ± SD 20.5 in the three respective

field seasons, further indicating that these were sustained social interactions, rather than random

218 encounters.

In total, we identified 36,885 social interactions over the three field seasons of study.

220 These interactions were used to define a weighted social network for each lek recording session.

The nodes in the network were the individual males, and the edges were weighted by the number

222 of social interactions between each pair of males. An earlier validation study compared social

interactions that were detected by the proximity system with those that were directly observed

224 for 11 males (Ryder *et al.* 2012), and confirmed that all of the interactions detected by the

proximity system were also directly observed.

Null Model Validation of the Social Networks

228 Two broad classes of methods have been described for building social networks: (i) networks

that are built based purely on the proximity of animals unrelated to their behavioral context

230 ("gambit of the group"), and (ii) networks that are built based on specific behavioral criteria that

are directly observed (Franks *et al.* 2010; Croft *et al.* 2011; Farine 2015). Although the

232   interactions in our study were not directly observed, they were recorded at specialized display

perches that have a known function in male-male social interactions (Schwartz & Snow 1978;

234   Ryder *et al.* 2011a). Hence, they do not qualify as the gambit of the group. To further

demonstrate this point, we performed a pre-network permutation of the raw data to determine

236   whether the observed network edges occurred more often than expected by chance (Farine 2017).

This analysis was highly conservative in that it preserved key features of the data including male

238   visit rates to specific territories within each recording session and lek; it is described in detail in

the supplement and Fig. S2. The results demonstrated that 95% of the observed network edges

240   had a greater edge weight than expected under stringent permutation conditions. Moreover, these

preferred edges had edge weights that were 10-fold to 50-fold greater than expected by chance

242   under these stringent conditions (Fig. S2). This provides an additional validation to our approach,

because it indicates that the observed network edges were nearly always preferred, even relative

244   to other possible interactions within the same lek. These results are also consistent with a

previous validation study indicating that these methods capture male-male coalition partnerships

246   that are directly observed (Ryder *et al.* 2012).

Although pre-network data permutations are sometimes used to derive adjusted

248   association indices, or to prune networks prior to further analysis, we did not take this approach

for several reasons. First, the statistical rarity of a relationship in our system does not *a priori*

250   define the importance of any one social interaction, especially given that all of the interactions

occurred on display territories with specialized function. A single interaction between two rarely

252   interacting males may have been highly consequential (e.g., if one of those males had a highly

influential hormone-behavioral phenotype). Conversely, a single interaction between two

254   frequently interacting males may have been relatively unimportant. It would be unwarranted to

assume that partnership rarity indicates the importance of any single interaction in this context.

256   Second, and perhaps more importantly, the goal of this study was a comparison across multiple

networks. Some networks are genuinely less preferential, and more random, than others. If we

258   modified the network edges based on permutation-based indices, it would disproportionately

prune the truly random networks, introducing a source of bias that would be contrary to our main

260   goal. Hence, all further analyses are based on networks where the edge weights are given by the

observed number of interactions, as this method is most appropriate for our study system and

262   aims. Below, we also describe a separate node-label permutation that provided an additional

check on our final statistical analyses.

Social Specialization

266   To quantify social specialization, we sought a metric that would capture the extent to which a

network was partitioned into exclusive social relationships (as opposed to a network made up of

268   non-specific or non-exclusive partnerships). To do this, we used a network metric of

specialization that is commonly used in community ecology called H_2' (Blüthgen *et al.* 2006).

270   An advantage of H_2' is that it is standardized against a theoretical maximum, based on the

overall activity levels of different nodes and Shannon entropy (Blüthgen *et al.* 2006); this makes

272   it possible to compare the extent of specialization across different bipartite networks in a

standardized way. To apply this metric to our manakin data, we converted each lek's social

274   network into its bipartite adjacency matrix (Fig. 1A), with floaters along one axis, and territory-

holders on the other, and then calculated social specialization as H_2' using the bipartite package

276   (Dormann *et al.* 2019). Higher values of specialization indicate that the network is well-

partitioned (i.e., made up of exclusive relationships), as illustrated in Fig. 1A. We chose to focus

278   on floater-territorial specialization because these two social classes are well-defined and floater-

territorial partnerships tend to be the most common in this species (Fig. S3; see also Ryder *et al.*

280   2011a).

Our measure social specialization at the network level, H_2', can also be related to the

282   exclusivity of social partnerships at the individual level, as used in (Sih *et al.* 2009; Edenbrow *et*

al. 2011; Dakin & Ryder 2018). All else being equal, a highly specialized network is expected to

284   be made up of individuals who are relatively more exclusive and/or more important towards their

partners, *sensu* (Sih *et al.* 2009; Dakin & Ryder 2018). Because H_2' has the properties described

286   in the previous paragraph, it is more appropriate as the network-level metric of social

specialization. Note that it is possible, and perhaps even common, for the edge weights in a

288   highly specialized network to be relatively invariant if the network is not fully connected, as

shown in the example manakin networks in Fig. 1A. Hence, specialization based on H_2' is not in

290   principle related to the network-average coefficient of variation of each individual's edge

weights (Maldonado-Chaparro *et al.* 2018).

Network Stability

294   We define network stability as the average persistence of network edges through time (Dakin &

Ryder 2020). To quantify the stability of manakin networks, we compared each lek's social

296   network from a given recording session to its subsequent recording session within the same field

season (Fig. 1B). Network stability was then calculated as the number of male-male partnerships

298   (binary network edges) shared by both time points divided by the number of partnerships at

either time point (Dakin & Ryder 2020). Higher values of stability indicate greater persistence of

300   social relationships within the network, independent of any changes in the representation of

particular males (nodes) (Poisot *et al.* 2012). To focus this measure on the persistence of strong

302   relationships, we computed network stability of partnerships that occurred at least six times

following (Dakin & Ryder 2020). The threshold of six was chosen because it corresponds to an

304   average rate of one social interaction per day in our data. We conducted an additional sensitivity

analysis to verify that alternative thresholds for the stability calculation (greater or less than six)

306   did not change our main results (see supplement for details). Previous work using this metric of

stability has shown that the wire-tailed manakin social networks are far more stable than

308   expected by chance (Dakin & Ryder 2020).

310   ***Behavioral Assortment***

Assortment refers to the extent to which individuals associate with similar partners (Fig. 1C); it

312   can be due to partner choice (homophily), shared environments, or the social transmission of

behavior (Dakin & Ryder 2018). Assortment was quantified using Newman's assortativity index,

314   which is a correlation coefficient for the statistical association among linked nodes within a

network. It ranges from –1 (a negative association), through 0 (no association), and up to +1 (a

316   positive association). To quantify the assortment of social behaviors, we first determined the

daily frequency of two behaviors for each male: his number of social interactions per day

318   (strength), and his number of unique social partnerships per day (degree). Strength and degree

are both repeatable measures of a male's social behavior in our study population (Dakin & Ryder

320   2018). We used the average log-transformed values of each male's strength and degree within

the recording session, and then calculated the assortativity coefficient for the entire social

322   network using the algorithm for weighted networks in the assortnet package (Farine 2016).

Because assortativity values for strength and degree were highly correlated (Pearson's r = 0.78, p

324   < 0.0001, n = 86 networks), we took the average of these two values as the measure of overall

behavioral assortment within the social network. Note that we used log-transformed values of

326   strength and degree because these two variables are strongly positively skewed (Dakin & Ryder

2018; Ryder *et al.* 2020), and assortativity is based on a Pearson's correlation. Finally, we also

328   computed the assortativity of the two discrete status classes (floater and territory-holder), to

ensure that our analysis of behavioral assortment was not solely driven by status-assortment.

Collective Testosterone

332   To understand how hormones might predict network properties, we derived a measure of

collective testosterone of each social network. This was based on the hormonal trait that was the

334   best predictor of social behavior in our previous study, referred to as "mean testosterone" (Ryder

et al. 2020). A male's mean testosterone is his average residual circulating testosterone. It is

336   calculated using a linear regression of log-transformed testosterone to statistically account for the

effects of field season, Julian date, time of day when captured, and duration of restraint, all of

338   which may influence point estimates of a male's baseline hormone level (Vernasco *et al.* 2019).

Hence, mean testosterone represents a standardized measure of a male's circulating testosterone,

340   independent of his capture conditions (Fig. 2A). Next, to determine collective testosterone, we

took the average mean testosterone for each social network, weighted by the interaction

342   frequency (strength) of the males within the network (Fig. 2B). Collective testosterone is thus a

group-level characteristic that is weighted towards the males that made the greatest contribution

344   to group social structure. In other words, networks with low collective testosterone are made up

of mostly low-testosterone individuals, whereas networks with high collective testosterone are

346 made up of mostly high-testosterone individuals. We calculated collective testosterone separately

for each status class, because the effects of hormones on social behavior are status-dependent

348 (Eisenegger *et al.* 2011; Boksem *et al.* 2013; Ryder *et al.* 2020).

350 ***Statistical Analysis***

To evaluate the hypothesis that collective testosterone predicts network properties, we analyzed

352 mixed-effects models of the social network properties in the lme4 package (Bates *et al.* 2018).

The three response variables were social specialization, network stability, and behavioral

354 assortment (Fig. 1). We used Akaike's Information Criterion (AIC) to compare four candidate

models for each response variable, as follows: (1) collective testosterone of territory-holders +

356 collective testosterone of floaters; (2) collective testosterone of territory-holders; (3) collective

testosterone of floaters, and (4) no testosterone predictors. All of the models included additional

358 fixed effects to account for field season (a categorical variable with three levels), the average

Julian date of the recording session, the average number of recorded hours per territory, and the

360 size of the social network (number of individuals), as well as a random effect of lek to account

for repeated measures.

362 Model comparison was performed on models fit with maximum likelihood, and the best-

fit models were re-fit using restricted estimation of maximum likelihood (REML) to derive

364 parameter estimates (Zuur *et al.* 2009). We used the lmerTest package to compute p-values for

parameter estimates in the mixed-effects models based on Satterthwaite's method (Kuznetsova *et*

366 *al.* 2018). We also report Nakagawa and Schielzeth's R^2_{LMM} values as an estimate of effect size

(Nakagawa & Schielzeth 2013). We verified that all models met the assumptions of linear

368 regression analyses. First, we checked that the Pearson residuals met the assumption of

normality. We visually inspected the partial residual plots for each fixed and random effect, to

370   confirm that there were no outliers or departures from the assumption of homoscedasticity. To

check for multicollinearity, we used the performance package (Lüdecke *et al.* 2019) to calculate

372   variance inflation factors (VIFs), and we verified that all VIFs were between 1 – 1.8.

374   ***Sample Sizes and Exclusions***

In two field seasons (2016-17 and 2017-18), we performed an experiment as part of a separate

376   study to test the influence of transiently-elevated testosterone on individuals (n = 5 individuals in

2016-17 and n = 4 in 2017-18; Ryder *et al.* 2020). The results of that experiment demonstrated

378   that elevated testosterone caused a temporary decrease in the frequency and the number of social

partnerships in the altered males (Ryder *et al.* 2020). It is important to note that this experiment

380   was not designed to test emergent properties at the network level, because it was conducted on a

limited scale with only one or two individuals temporarily altered within each lek. We therefore

382   excluded the 6 post-manipulation networks from the main analysis in this study. We verified that

when we included these manipulated leks, all of our main conclusions were unchanged.

384   After excluding the 6 post-manipulation observations, 80 of the original 86 recording

sessions remained. Table 1 summarizes the sample sizes for the network-level analyses. In 10

386   cases, specialization could not be calculated because the bipartite network did not have sufficient

data to determine H_2'. Stability could not be calculated in 26 cases, when either the recording

388   session occurred at the end of a field season, or when there were insufficient partnerships that

met the criteria for the stability calculation.

    Node-label Permutation Analysis

392   To evaluate the possibility that our results could be influenced by other properties of the leks that

were independent of hormonal traits, we also performed a statistical permutation of the post-

394   network data (Farine 2017). The purpose of this analysis was to verify that the results were

driven by (and sensitive to) the relative contribution and social position of different males. To do

396   this, we performed a statistical permutation that randomized the node labels (male IDs) within

each of the social networks, retaining network topology, and leaving each male's testosterone

398   traits unchanged. Hence, this analysis preserved which males were present in which recording

session, but it randomized the relative position and contribution of each male. After generating

400   1,000 of these node-label permutation datasets, we recalculated specialization, stability, and

assortment, and then refit the top models from our mixed-model analysis. We then compared the

402   slope estimates from the observed data to those derived from 1,000 node-label permutations. As

a one-sided p-value, we calculated the proportion of slope estimates from the node-label

404   permutations that were more negative than the corresponding estimate from the observed data.

406   **Results**

The lek social networks had 14 males on average, but there was considerable variation among

408   leks ranging from 3 to 43 individuals (Fig. S1). Additional descriptive statistics are provided in

Table 1. Although network size has a lognormal distribution (Fig. S1), all other emergent

410   properties of the social structure were approximately normally distributed (Fig. S4).

Table 1. Descriptive statistics for manakin social networks. Means and standard deviations (SD) are provided for each field season. The bottom row provides the sample sizes for the number of social networks analyzed (n_{obs}) at each lek (N_{lek}). Additional data is provided in the supplement in Fig. S1 and Fig. S4.

Field season	Network size (# birds)	Social specialization	Network stability	Behavioral assortment	Collective T (terr.)	Collective T (floa.)
15-16	14.5 (8.9)	0.50 (0.22)	0.43 (0.25)	0.01 (0.36)	0.44 (0.42)	−0.31 (0.56)
16-17	17.8 (10.5)	0.30 (0.24)	0.47 (0.20)	0.06 (0.30)	0.49 (0.24)	−0.22 (0.37)
17-18	10.3 (4.5)	0.42 (0.30)	0.42 (0.27)	0.02 (0.37)	0.43 (0.40)	−0.70 (0.46)
Sample size	$n_{obs} = 80$	$n_{obs} = 70$	$n_{obs} = 54$	$n_{obs} = 80$	$n_{obs} = 80$	$n_{obs} = 80$
	$N_{lek} = 11$	$N_{lek} = 11$	$N_{lek} = 11$	$N_{lek} = 11$	$N_{lek} = 11$	$N_{lek} = 11$

We found that the collective testosterone of territorial males could predict all three emergent properties of the social networks (specialization, stability, and assortment). The leks with greater representation of high-testosterone territorial males were less specialized, less stable, and more negatively assorted (Fig. 3; all $p \leq 0.03$ in mixed-effects models). The slope coefficients for these three relationships were also all greater than expected under a node-label permutation (inset panels, Fig. 3; all $p < 0.02$). Model selection results for the observed data indicated considerable uncertainty in the best-fit model for each of the three network properties analyzed (Table S1). However, the collective testosterone of territory-holders was a significant predictor in all of the best-supported models (Table S2). In contrast, the collective testosterone of floater males was not a significant predictor of network properties in any of the best-fit models (Table S2).

We did not detect any significant effects of Julian date on network properties within our study period, but we did observe some significant year-to-year differences (e.g., social specialization and behavioral assortment in Table S2). Assortment was the only property that was significantly related to network size and recording effort (Table S2). All else being equal, behavioral assortment was more positive in larger networks, and it was more negative in networks that had longer recording sessions.

436    $R^2_{LMM(m)}$ provides an estimate of the proportion of variance explained by the fixed effects

in a model (Table S1). The $R^2_{LMM(m)}$ for assortment in the best-fit model was 0.52. This indicates

438   that about 52% the variation in behavioral assortment could be explained by the combined

associations with collective testosterone, network size, sampling effort, and annual variation

440   (Table S2). The $R^2_{LMM(m)}$ for stability was 0.12, indicating that collective testosterone and the

other predictors (network size, sampling effort, year) could explain about 12% the variation in

442   that metric (Table S2). Finally, for specialization, the $R^2_{LMM(m)}$ indicated that the combined

effects of testosterone and these other sources of variation together explained about 20% of the

444   variation (Table S2).

446   **Discussion**

The collective testosterone of the dominant, territory-holding males within a lek was associated

448   with multiple emergent properties of the social network (Fig. 3). Variation in collective

testosterone is a function of both the number of high-testosterone males and their frequency of

450   social interactions. Our results indicate that the hormone-mediated behavior of these individuals

may affect all three social network properties of specialization, stability and assortment. This

452   indicates that the effects of testosterone on dominant males may mediate an extended phenotype

with the power to shape social structure (Dawkins 1982). In contrast, the collective testosterone

454   of floater males was not significantly associated with any of the emergent network properties

(Table S2).

456   How can the relationship between hormone levels and network properties be explained in

terms of individual mechanisms? Given that testosterone has antagonistic effects on the sociality

458   of territorial males (Ryder *et al.* 2020; Vernasco *et al.* 2020), we hypothesize that the behavior of

high-testosterone individuals can cause several features of the social organization to break down.

460   Our previous results showed that high-testosterone dominant individuals have a reduced ability

to attract and maintain social partners (Ryder *et al.* 2020). We propose that this weakening of

462   social relationships may cause floater males to prospect elsewhere for new partners, negatively

affecting both the stability and specialization of the lek network as a whole. Likewise, high-

464   testosterone dominant males may inhibit the processes of social contagion, reciprocity, and/or

behavioral matching that can cause positive behavioral assortment (Dakin & Ryder 2018).

466   Recent studies have found that in other social animals, sparse and specialized social networks

can be associated with fitness benefits (Stroeymeyt *et al.* 2018); hence, a breakdown to this

468   organization may incur costs (Maldonado-Chaparro *et al.* 2018). Testing our proposed

mechanism for the link between hormone-mediated behavior and network dynamics will require

470   direct observation of the individual social behaviors that occur within dyads, and how these

behaviors change through time. Because our current data cannot assess the fine-scale valence of

472   social interactions, further studies are needed that combine direct observation with high-

throughput data on social network dynamics.

474   We included several additional parameters in our analyses to account for social network

size and sampling effort. Although it was not one of our main hypotheses, we noted that network

476   size was positively associated with behavioral assortment (Table S2). In other words, males were

more likely to associate with behaviorally similar partners in larger leks, whereas they were more

478   likely to associate with dissimilar partners in smaller leks. Effects of group size on assortment

have been noted in a few other studies, although the form of this relationship varies (Griffiths &

480   Magurran 1997; Ilmarinen *et al.* 2017; McDonald *et al.* 2017). In manakins and other lekking

systems, larger leks are known to have heightened display activity and higher female visitation

482   rates (Lank & Smith 1992; Durães *et al.* 2009). This raises the possibility that social facilitation

and heightened activity may be associated with increased homophily and/or behavioral matching

484   (either through partner choice, contagion, or reciprocity). Another plausible explanation is that

smaller social groups may constrain these behavioral processes, by making it more difficult to

486   find or match a suitable partner. This hypothesis could be explored in experiments on captive

systems and simulation models.

488   Behavioral assortment was also more positive in networks that were recorded for less

time in our study (in other words, leks where males associated with behaviorally similar partners

490   tended to be recorded for fewer days). Recording time was designed to be approximately even

among leks (Fig. S1), with the most common reason for an extended recording duration being

492   inclement weather that extended the recording session. Therefore, we speculate that inclement

weather may have reduced the amount of behavioral matching (assortment) while also affecting

494   recording time. Although we did not collect data on weather at each lek as this was not our main

goal, the question of how inclement weather affects group-level social dynamics is an interesting

496   one that merits further study.

        Conducting experimental tests of causation at the level of whole social networks remains

498   a major challenge in ecology (Pinter-Wollman *et al.* 2014; James *et al.* 2009). Although our

node-label permutation analysis provides evidence that our results are not merely due to

500   structural differences among networks, we cannot rule out the possibility that high-testosterone

individuals chose to participate in certain networks due to other factors that may also influence

502   the emergent properties of the network (e.g., environmental quality and/or female activity). It is

important to note that in this system, as in many other wild animals, controlled experimental

504   manipulations of the broader social network structure are not yet possible (Zyphur *et al.* 2009;

Akinola *et al.* 2016). Nevertheless, our data here indicate that the increased prevalence of

506 dominant, high-testosterone individuals can predict changes social dynamics and subsequent

higher-order network structure. These findings establish that hormone-behavior relationships are

508 not limited to one individual, but instead hormones have population-level consequences

(McClintock 1981; Robinson 1992).

510   Increasing evidence demonstrates that the structure and stability of social networks is

often associated with benefits–and costs–during foraging, breeding, and disease outbreaks (Silk

512 *et al.* 2010; Maldonado-Chaparro *et al.* 2018; Riehl & Strong 2018; Stroeymeyt *et al.* 2018).

Given the widespread influence of steroid hormones on social interactions across vertebrates

514 (Adkins-Regan 2005), we expect that collective testosterone will be broadly associated with

social network properties in other systems. The direction of these effects may depend on the

516 context of the behavioral interactions that form the social network, and whether these

interactions are primarily cooperative or competitive in nature. Taken together with previous

518 studies, we propose that testosterone-mediated behavior can alter social network dynamics in

ways that often impede the evolution of stable social relationships and cooperation.

Figure Legends

Figure 1. Example social networks illustrating social specialization, network stability, and behavioral assortment. (A) Social specialization is measured using the bipartite form of the

social network. In the example on the left, the associations between floater males and territory-holders are poorly partitioned, creating a network with a relatively low specialization. On the

right, there is greater partitioning, such that each floater male maintains a smaller number of associations with specific territory-holding males. (B) Network stability is measured by

evaluating the persistence of partnerships from one recording session (t_1) to the next (t_2). The example at the top of (B) has a lower stability than the example on the bottom. (C) Behavioral

assortment measures the tendency of like to associate with like. In (C), nodes are shaded to indicate a continuous gradient of daily behavior (in this case, each male's average daily strength

is shown). In the example network on the left, the males tend to associate with behaviorally dissimilar partners, yielding a negative behavioral assortment. In the example on the right, the

males tend to associate with behaviorally similar partners, yielding a positive assortment. See Table 1 for additional data.

Figure 2. Collective testosterone of manakin social networks. (A) Circulating testosterone

538   varies among individual males. This graph shows repeated measures from 210 individual male manakins, sorted along the x-axis by mean testosterone, a standardized measure of a male's

540   average circulating testosterone level (Ryder *et al.* 2020). Opacity is used to denote the two male status classes (with subordinate floaters colored semi-transparent, and dominant territory-holders

542   colored opaque). Color ramping is used to denote each individual's hormone phenotype. (B-D)

These data were used to define collective testosterone in the present study. Node colors indicate

544   an individual's mean testosterone following the scale in (A). The collective testosterone of each

social network is calculated as the average of the individual hormone phenotypes, weighted by

546   strength, and is shown below each example. Collective testosterone was determined for each

status class separately. See Table 1 for additional data.

Figure 3. Collective testosterone predicts the structure of the social network. The networks

550   with higher-testosterone dominant males were less specialized, less stable over time, and had

more negative behavioral assortment. Each plot shows the partial residuals from a statistical

552   analysis that also accounts for field season, the average Julian date of the recording session, the

number of recording hours, network size, and lek identity. Because floater and territorial males

554   can differ in behavior, the analysis of behavioral assortment also accounted for status assortment

within each network. Different symbols are used to indicate repeated measures of 11 different

556   leks, colored according to the collective testosterone scale in Fig. 2. Inset panels show the results

of node-label permutation tests. In each case, the slope in the best-fit model (solid vertical line) is

558   significantly more negative than expected based on the distribution of permuted estimates (grey

distribution, dotted line is at 0). See Tables S1-S2 for additional data.

Acknowledgments

We thank Camilo Alfonso, Brian Evans, David and Consuelo Romo, Kelly Swing, Diego Mosquera, Gabriela Vinueza, and the Tiputini Biodiversity Station of the Universidad San Francisco de Quito. Funding was provided by the National Science Foundation (NSF) IOS 1353085 and the Smithsonian Migratory Bird Center.

Author Contributions

RD, ITM, BMH, and TBR designed the study. ITM, BMH, BJV, and TBR collected the data. RD, ITM, and TBR analyzed the data. RD and TBR wrote the manuscript. All authors edited the manuscript.

Declaration of Interests

The authors declare no competing interests.

Data Availability

All data and R scripts necessary to reproduce this study are available for download at:

https://figshare.com/s/13a311662fee686fa4f3

Supporting Information

Supplemental text, Figures S1-S5, and Table S1-S3 in the attached PDF

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Figure 1

A Social specialization

Low
0.08

High
0.83

Territory
-holders

Floaters

B Network stability

t_1 t_2

Low
0.33

High
0.88

C Behavioral assortment

Negative
-0.52

Positive
0.66

Avg. daily behavior

Figure 2

Figure 3