# UNIVERSITY<sup>OF</sup> BIRMINGHAM

# University of Birmingham Research at Birmingham

### Shared environmental responses drive cooccurrence patterns in river bird communities

Royan, Alexander; Reynolds, S. James; Hannah, David; Prudhomme, Christel; Noble, David G.; Sadler, Jonathan

DOI:

10.1111/ecog.01703

Document Version Early version, also known as pre-print

Citation for published version (Harvard):

Royan, A, Reynolds, SJ, Hannah, D, Prudhomme, C, Noble, DG & Sadler, J 2015, 'Shared environmental responses drive co-occurrence patterns in river bird communities', *Ecography*, vol. 39, no. 8, pp. 733-742. https://doi.org/10.1111/ecog.01703

Link to publication on Research at Birmingham portal

Publisher Rights Statement: Checked for eligibility: 27/01/2017

**General rights** 

Unless a licence is specified above, all rights (including copyright and moral rights) in this document are retained by the authors and/or the copyright holders. The express permission of the copyright holder must be obtained for any use of this material other than for purposes permitted by law.

•Users may freely distribute the URL that is used to identify this publication.

- •Users may download and/or print one copy of the publication from the University of Birmingham research portal for the purpose of private study or non-commercial research.
- •User may use extracts from the document in line with the concept of 'fair dealing' under the Copyright, Designs and Patents Act 1988 (?)

•Users may not further distribute the material nor use it for the purposes of commercial gain.

Where a licence is displayed above, please note the terms and conditions of the licence govern your use of this document.

When citing, please reference the published version.

Take down policy

While the University of Birmingham exercises care and attention in making items available there are rare occasions when an item has been uploaded in error or has been deemed to be commercially or otherwise sensitive.

If you believe that this is the case for this document, please contact UBIRA@lists.bham.ac.uk providing details and we will remove access to the work immediately and investigate.

Download date: 10. Apr. 2024

## Shared environmental responses drive co-occurrence

2	patterns in river bird communities		
3	Alexander Royan <sup>a*</sup> , S James Reynolds <sup>b</sup> , David M Hannah <sup>a</sup> , Christel		
4	Prudhomme <sup>c</sup> , David G Noble <sup>d</sup> , and Jonathan P Sadler <sup>a,b*</sup>		
5			
6	<sup>a</sup> School of Geography, Earth & Environmental Sciences, College of Life &		
7	Environmental Sciences, University of Birmingham, Birmingham B15 2TT, UK		
8			
9	<sup>b</sup> Centre for Ornithology, School of Biosciences, College of Life & Environmental		
10	Sciences, University of Birmingham, Birmingham B15 2TT, UK		
11			
12	<sup>c</sup> Centre for Ecology and Hydrology, Wallingford OX10 8BB, UK		
13			
14	<sup>d</sup> The British Trust for Ornithology, The Nunnery, Thetford IP24 2PU, UK		
15			
16	*Corresponding authors: <a href="mailto:axr196@bham.ac.uk">axr196@bham.ac.uk</a> : <a href="mailto:j.p.sadler@bham.ac.uk">j.p.sadler@bham.ac.uk</a>		
17			
18	Submitted to Ecography – Final accepted draft (13.5.2015)		

### Abstract

Positive or negative patterns of co-occurrence might imply an influence of biotic
interactions on community structure. However, species may co-occur simply because
of shared environmental responses. Here, we apply two complementary modelling
methodologies - a probabilistic model of significant pairwise associations and a
hierarchical multivariate probit regression model - to (i) attribute co-occurrence
patterns in 100 river bird communities to either shared environmental responses or to
other ecological mechanisms such as interaction with heterospecifics, and (ii) examine
the strength of evidence for four alternative models of community structure. Species
co-occurred more often than would be expected by random community assembly and
the species composition of bird communities was highly structured. Co-occurrence
patterns were primarily explained by shared environmental responses; species'
responses to the environmental variables were highly divergent, with both strong
positive and negative environmental correlations occurring. We found limited
evidence for behaviour-driven assemblage patterns in bird communities at a large
spatial scale, although statistically significant positive associations amongst some
species suggested the operation of facilitative mechanisms such as heterospecific
attraction. This lends support to an environmental filtering model of community
assembly as being the principle mechanism shaping river bird community structure.
Consequently, species interactions may be reduced to an ancillary role in some
avifaunal communities, meaning if shared environmental responses are not quantified
studies of co-occurrence may overestimate the role of species interactions in shaping
community structure.

- **Key-words**: Assembly rules; Bayesian; Coexistence; Community assembly;
- Community ecology; Ecohydrology; Hydroecology; River birds; Spatial analysis

#### Introduction

46

47 Species distributions are determined by a range of factors, including climate, land 48 cover, landscape barriers, dispersal ability and species interactions (MacArthur, R. H. 49 1972). Mechanistic models of community assembly predicate that the impacts of these 50 biotic and abiotic factors on the trait composition of communities manifest through 51 assembly rules (i.e. biological filters) and the ability of species to tolerate local 52 conditions (i.e. environmental filters) (Belmaker, J. and Jetz, W. 2013). 53 Environmental filters restrict community membership to species possessing a 54 particular set of functional traits (Petchey, O. L. et al. 2007). Conversely, a number of 55 positive (e.g. mutualisms and commensalisms), negative (e.g. competition) and 56 neutral models of species interaction may describe overlapping patterns in species 57 distributions (Araújo, M. B. and Rozenfeld, A. 2013). Such interactions may leave 58 imprints on species distributions that are detectable at large spatial scales (Baselga, A. 59 et al. 2012, Gotelli, N. J. et al. 2010, Heikkinen, R. K. et al. 2007); both positive and 60 negative interactions can be discernible across scales of hundreds of kilometres 61 (Araújo, M. B. and Rozenfeld, A. 2013). However, positive/negative associations can 62 result from species having similar/dissimilar habitat requirements as well as from 63 direct or indirect interactions (Ovaskainen, O. et al. 2010), meaning (dis)similarity in 64 habitat preference is a legitimate ecological explanation for why two species might 65 have positive (or negative) association. Consequently, disentangling the relative roles 66 of environmental and biotic factors in community assemblage processes and 67 understanding their link to spatial patterns in species distributions is a considerable 68 challenge and is a prominent unresolved issue in ecology (Barnagaud, J. Y. et al. 69 2014, Kraft, N. J. B. et al. 2014, MacArthur, R. and Levins, R. 1964, Macarthur, R. H. 70 1958, Peron, G. and Altwegg, R. 2015).

Associations between species have particularly pervasive impacts on bird populations, where interspecific competition may influence the distribution, habitat choice, abundance and reproduction of species (Cody, M. L. 1974), and interactions between them have been detected in the distributions of species at large scales (Gotelli, N. J. et al. 2010, Heikkinen, R. K. et al. 2007, Robertson, O. J. et al. 2013). Community-wide patterns of spatial segregation in the Danish avifauna could not be attributed to heterogeneity in the distribution of habitat or differences in habitat utilisation, suggesting that the large-scale operation of species interactions (e.g. interspecific territoriality and conspecific attraction) can cause behaviour-driven assembly patterns (Gotelli, N. J. et al. 2010). However, it has been suggested that habitat variability, disturbance regimes and food availability structure bird communities at finer spatial scales, with geomorphic landforms and climatic envelopes becoming more important in determining neighbourhood associations at coarser scales (Börger, L. and Nudds, T. D. 2014, Gotelli, N. J. et al. 1997, Kroll, A. J. et al. 2014, Mattsson, B. J. et al. 2013). Considering the impact of both environmental and biotic effects, there are four prevalent hypotheses that explain the distribution of heterospecifics (Mattsson, B. J. et al. 2013): (1) a null hypothesis that species are distributed entirely randomly and patterns of species occurrence do not conform to variability in habitat condition or the distribution of heterospecifics (Hubbell, S. P. 2001); (2) the environmental filtering hypothesis where community structure relates to variability in abiotic factors (e.g. disturbance) with species partitioned according to abiotic constraints (Macarthur, R. H. 1958). Community membership will therefore be restricted to those species possessing a particular set of functional traits (Keddy, P. A. 1992, Petchey, O. L. et al. 2007, Weiher, E. and Keddy, P. A. 1995) and one might expect species occurrences to

71

72

73

74

75

76

77

78

79

80

81

82

83

84

85

86

87

88

89

90

91

92

93

94

be highly correlated with environmental variables; (3) the interspecific interaction hypothesis, where species distributions are primarily influenced by the occurrence of heterospecifics and species either aggregate in patches occupied by heterospecifics or segregate to occupy empty patches. Individuals may cluster with heterospecifics because of the benefits of group membership in terms of increased vigilance, reduced predation risk and enhanced assessment of habitat quality (Ward, P. and Zahavi, A. 1973). Conversely, segregation may occur due to mechanisms such as competitive exclusion (MacArthur, R. and Levins, R. 1967). In this interspecific interaction hypothesis we would expect to find irregular distributions of species that are poorly explained solely by environmental variables; (4) a pluralistic hypothesis in which species not only interact and aggregate/segregate according to the distribution of heterospecifics but they do so in a manner reflecting the patchiness of habitats that promote fitness. These hypotheses can be termed 'surrogate hypotheses' as they are assessed through an examination of patterns in data rather than by controlled experimentation (Araujo, M. B. and Luoto, M. 2007) and provide a strong analytical framework in macro-ecological studies where the manipulation of experimental conditions is not possible (Gotelli, N. J. and McGill, B. J. 2006). Investigation into non-random patterns of association between pairs of species has largely centred on the comparison of presence/absence matrices with null models

96

97

98

99

100

101

102

103

104

105

106

107

108

109

110

111

112

113

114

115

116

117

118

119

120

has largely centred on the comparison of presence/absence matrices with null models (Gotelli, N. J. et al. 1997, Gotelli, N. J. et al. 2010, Ulrich, W. and Gotelli, N. J. 2010, Weiher, E. et al. 1998). Analyses are based on inferences as to whether an observed matrix differs from those produced by random processes or from a known ecological mechanism. Investigations of these matrices have led to the analysis of empirical patterns in species distributions and the development of ecological hypotheses for community organisation, including the community assembly rules of Diamond, J. M.

(1975). However, it is not clear if it is possible to use such approaches to discriminate between spatial patterns caused by species interaction and those caused by affinities for particular habitats. An alternative method in modelling biotic interactions is to restrict the distribution of one species by including the abundance of another as a predictor alongside environmental variables (Meier, E. S. et al. 2010). However, the abundance of heterospecifics could act as a surrogate for absent important environmental variables (Araújo, M. B. and Luoto, M. 2007). Additionally, twospecies occupancy models can use species distribution data to model the probability of occurrence of species pairs and show how the detection probability of either species may change in the presence of the other (Richmond, O. M. W. et al. 2010). This approach is well suited to the analysis of the co-occurrence of subordinate (e.g. prey) species and dominant (e.g. predator) species (Robinson, Q. H. et al. 2014) but has less applicability to macro-ecological studies of multiple species distributions. In this study we aimed to: (a) investigate pairwise patterns in species cooccurrence and partition co-occurrence into correlative responses to environmental variables or potential species interactions, and (b) evaluate the evidence for the operation of different community assembly processes. We applied two complementary models to a large-scale dataset of river bird records for mainland Great Britain, relating river bird occurrence data to variables that quantify land use and long-term variability in climate and river flow. First we used the probabilistic model of species co-occurrence developed by Veech, J. A. (2013) to calculate significant positive, negative or random associations between species, which represent all of the ways in which species can co-occur (Veech, J. A. 2014). We then compared the output of the probabilistic model to that of a joint species distribution model (JSDM) (Pollock, L. J. et al. 2014), which attributes co-occurrence patterns to either

121

122

123

124

125

126

127

128

129

130

131

132

133

134

135

136

137

138

139

140

141

142

143

144

shared environmental responses or other ecological processes (e.g. species interactions) in a single integrated modelling process.

We addressed three specific objectives which were to: (i) examine patterns of significant positive or negative co-occurrence in river bird communities across mainland Great Britain; (ii) assess the strength of shared environmental responses between species pairs relative to the strength of evidence for species interactions; and (iii) characterise the composition of river bird communities and assess the importance of species interactions in determining species distributions across different guilds.

#### Material and methods

River bird data

Bird survey data were obtained from the British Trust for Ornithology's (BTO's) Waterways Breeding Bird Survey (WBBS), an annual large-scale national survey of birds on rivers in Great Britain. Waterway locations for the WBBS are randomly selected according to a stratification procedure (Marchant, J. H., Noble, D.G., Leech, D.I., Freeman, S.N. 2002). First, national grid squares at 2 × 2 km resolution are selected at random and those without a waterway are discarded. Waterways within remaining squares are then identified according to the definition of any double blue line, with shaded in-fill, on the OS 1:25,000 Pathfinder map series. A single waterway within each square is then randomly selected, with the start and end points of the survey location determined by the observer. Each year two visits are made to WBBS survey locations by the same observer, one in the first half of the breeding season and one in the second half; typically early April and late June, respectively. The WBBS comprises transect methodology, with survey locations divided into 10 linear transects of fixed 500 m length situated along one side of the waterway. The observer counts all

birds seen or heard within 100 m of each 500 m linear transect, with the total number of transects surveyed (up to a maximum of 10) counted and recorded so that heterogeneity in survey effort may be quantified and accounted for in subsequent statistical analyses. As multiple visits to survey locations provide more reasonable estimates of species occupancy and abundance which, in turn, reduce bias associated with detection probability (McCarthy, M. A. et al. 2012), we selected WBBS survey locations with a minimum of four repeated visits between 1998 and 2011 (inclusive). In total, 100 WBBS locations were used, achieving wide coverage across mainland Great Britain (Fig.1).

We used data for 19 bird species which were selected because they are largely ubiquitous across riverine landscapes in Great Britain and sufficiently characterised diversity in river bird communities (Royan, A. et al. 2013, Royan, A. et al. 2014). Additionally, these species also possess an array of ecological traits, which provides the opportunity to study a broad range of responses to the environmental variables and interaction with heterospecifics (Newbold, T. et al. 2014, Vandewalle, M. et al. 2010), whilst the relationship between their distribution in Great Britain and key environmental variables is well characterised (Royan, A. et al. 2013, Vaughan, I. P. et al. 2007). Our response variable was defined as the presence/absence of a bird species at each of the 100 locations. Because waterbird distributions may shift across years (Lehikoinen, A. et al. 2013), river bird occurrence data could potentially be influenced by the date of survey. Occurrence data were, therefore, combined across visits to the survey locations, whereby a species was recorded as present if it was observed during surveys at any time between 1998 and 2011 (the presence/absence matrix is provided in Table A1).

Environmental data

Bird distributions can be influenced by a range of environmental factors, including hetereogeneity in land use and habitat (Luck, G. W. et al. 2013), variability in climate (Renwick, A. R. et al. 2012) and, in the case of river birds, variability in river flows (Royan, A. et al. 2013). We therefore paired river bird occurrence data with variables that quantify each of these environmental factors (Table 1).

Annual river flow data for the period 1998 to 2011 were obtained from the National River Flow Archive (NRFA), organised by the Centre for Ecology and Hydrology (CEH). River flow variables were calculated using long-term averages between 1998 and 2011 (inclusive) and characterised variability around average, high and low flows in the UK and were defined *a priori* according to (Royan, A. et al. 2013). To ensure the relevance of flow variables to the river bird data, all 100 WBBS survey locations were situated within 10 km of a river flow gauging station. We also ensured that there were no major tributary inflows or anthropogenic barriers located between station-survey pairings.

Land use data were produced by the CEH as part of the Countryside Survey 2000 and were derived from the satellite-generated Land Cover Map 2000 (LCM2000) with a resolution of 50 m. Land use variables quantified the percentage of the riverine catchment comprising woodland, arable, urban, heathland and grassland habitats. Climate data for each survey location were obtained from the UKCP09 gridded observation datasets at 5 × 5 km resolution at monthly timescales and climate variables were calculated using data between January 1998 and December 2011 (inclusive). Further details on how these data were derived can be found on the UK Met Office website (http://www.metoffice.gov.uk/climatechange/science/monitoring/ukcp09/index.html).

We calculated mean breeding season temperature (°C) and total breeding season

rainfall (mm), with the breeding season defined as April to July (inclusive). We chose to use climate variables calculated during the breeding season rather than annual or winter variables as they have been shown to be better predictors of bird distributions in Great Britain (Renwick, A. R. et al. 2012).

#### *Modelling procedure*

Initially, we used the probabilistic modelling approach developed by Veech, J. A. (2013) to investigate statistically significant pairwise patterns in species co-occurrence. The model calculates the expected frequency of co-occurrence between each pair of species based on the distribution of one species being independent of the second one. It then compares the expected frequency to the observed frequency and returns the probability that a lower or higher value of co-occurrence could have been obtained by chance. The probabilities can be interpreted as *p* values as the model classifies species pairs into categories of significant positive, negative or random association based upon an alpha threshold of 0.05.

We further analysed co-occurrence patterns using the JSDM developed by Pollock, L. J. et al. (2014). This hierarchical modelling approach attributes co-occurrence patterns to shared environmental responses and residual patterns of co-occurrence and so can be used to investigate the mechanisms that influence the structure and dynamics of species assemblages. For instance, evidence of strong environmental correlation would support the hypothesis for the operation of environmental filtering assembly mechanisms. Weak environmental correlation and strong residual correlation would lend support to the interspecific interaction hypothesis, although one cannot completely discount the possibility of some influence of unmeasured habitat relationships (Börger, L. and Nudds, T. D. 2014). Evidence of strong environmental correlation and also a broad range of residual correlation would

provide evidence for a pluralistic hypothesis of community organisation. The null hypothesis of random community organisation would be supported if no significant positive or negative pairwise associations were detected.

246

247

248

249

250

251

252

253

254

255

256

257

258

259

260

261

262

263

264

265

266

267

268

269

270

Full details of the JSDM can be found in Pollock, L. J. et al. (2014) but below we summarise this approach in brief. Co-occurrence patterns were modelled using a hierarchical probit regression model in which the linear environmental predictors were related to a binary response variable using a latent variable formulation. The probability of occurrence of each species at a site is determined by the mean of a normal distribution (the latent variable), assuming a standard deviation of one, whereby species are modelled as being present when the distribution mean is greater than zero and modelled as absent when less than zero. Associations amongst species are modelled by changing the locations of the latent multivariate normal distribution and probabilities of co-occurrence by changing the correlations of the latent distribution. The number of species being modelled determines the number of dimensions of the multivariate normal distribution. The means of the normal distribution were modelled using regression equations and a matrix of regression coefficients. To relate species occurrences to the environmental variables, we calculated the coefficients by fitting a Bayesian implementation of a binomial generalized linear model (Gelman, A., Yu-Sung, S., Yajima, M., Hill, J., Pittau, M.G., Kerman, J., Zheng, T., Dorie, V. 2014), specifying a logit link function and a Cauchy prior distribution for the coefficients to prevent coefficient inflation from complete separation of occurrences across predictor variables (a common problem in logistic regression) (Gelman, A. et al. 2008). Residual correlation is controlled by a matrix of correlation coefficients in the latent distribution between species. This was calculated by re-scaling the variance/covariance matrix of the multivariate normal distribution by dividing the terms by the corresponding standard deviations and defining with an inverse-Wishart prior. A second correlation matrix was also calculated to account for the component of between-species correlation that is due to shared environmental responses. The model estimates posterior distributions for four parameters: correlations between species due to the environment, the residual correlation between species, regression coefficients, and the predicted probability of occurrence of a species at each site.

The model was fitted using the Markov Chain Monte Carlo Bayesian software JAGS v3.4.0 in R v3.0.2 via R2jags v0.03-11 (Plummer, M. 2014). We ran five chains for 100,000 iterations with the first 10,000 discarded as burn-in and the remaining samples thinned by a factor of 10 such that 9,000 samples were retained for analysis. We used vague normal priors for all model parameters (mean = 0, standard deviation = 1).

We characterised the composition of river bird communities by converting the matrix of species occurrences across the 100 river locations into a binary dissimilarity matrix and by then using hierarchical cluster analysis, with Ward's clustering, to create a dendrogram that illustrates the clustering of species occurrences. Non-parametric analysis of variance tests (Kruskal-Wallis Rank Sum) were used to investigate differences in the distribution of residual correlations between species within each guild identified by clustering.

#### **Results**

Probabilistic modelling of species co-occurrence revealed instances of positive (species co-occur significantly more frequently than expected), negative (species co-occur significantly less frequently than expected) and random species associations (observed frequency of co-occurrence does not significantly depart from expected).

Positive associations were more common than negative associations (Fig. 2). Because the geographic range of all 19 species encompasses the whole of Great Britain, all 100 locations were used in the analysis of 171 species pairs; 42% of species associations were positive, 11% were negative, and 47% were random. This indicates that the species composition of bird communities largely followed non-random patterns of community assembly, although random co-occurrence patterns were prevalent.

The output of the JSDM revealed that these pairwise associations could be largely attributed to shared environmental responses. Species' responses to the environmental variables were highly divergent, with both strong positive and negative (e.g. R > 0.7 and R < -0.7) environmental correlations occurring (Fig. 3). The range of residual correlation, however, was comparatively narrower and tended to be positive. The strength of environmental correlation was considerably greater than that of residual correlation (Fig. 4); where both environmental and residual correlation were positive (n = 110), environmental correlation was stronger (i.e. closer to 1) in 83 of the pairwise associations, and where both environmental and residual correlations were negative (n = 23), environmental correlation was stronger (i.e. closer to -1) in 20 of the pairwise associations.

Hierarchical cluster analysis revealed that the communities can be decomposed into three guilds, whereby species within each guild were likely to co-occur due to similar environmental tolerances (Fig. 5). Species within each guild displayed similar characteristics for foraging habitat and resource acquisition. These were species that forage on fast-flowing rivers (i.e. common sandpiper [Calidris hypoleucos], white-throated dipper [Cinclus cinclus], common merganser [Mergus merganser], grey wagtail [Motacilla cinerea]), those that forage on large, slow-flowing rivers (i.e. Eurasian coot [Fulica atra], great cormorant [Phalacrocorax

carbo], great crested grebe [Podiceps cristatus], common kingfisher [Alcedo atthis], common moorhen [Gallinula chloropus], mute swan [Cygnus olor], Eurasian reed warbler [Acrocephalus scirpaceus], sedge warbler [A. schoenobaenus], western yellow wagtail [Motacilla flava]), and species that forage at riparian margins across a range of rivers (i.e. northern lapwing [Vanellus vanellus], Eurasian oystercatcher [Haematopus ostralegus], common redshank [Tringa totanus], common reed bunting [Emberiza schoeniclus], white wagtail [Motacilla alba], and sand martin [Riparia riparia]). Further analysis of the residual correlation between species within these guilds revealed additional complexities in species covariance patterns. Residual correlation was significantly higher between species within the riparian guild than between species within the slow-flowing guilds (Kruskal-Wallis:  $\chi^2 = 15.70$ , p < 0.001) (Fig. 6).

#### **Discussion**

The diversity and complexity of abiotic and biotic factors that influence species distributions present considerable challenges in the exploration of alternative community assembly processes. The processes themselves may be too difficult and complex to monitor directly and so inference is limited to indirect approaches, such as the analysis of co-occurrence (Ovaskainen, O. et al. 2010) and the testing of surrogate hypotheses (Araújo, M. B. and Luoto, M. 2007). Species interactions affect the distribution of many bird species (Cody, M. L. 1974). However, the importance of interactions between species in structuring bird communities at macro-ecological scales is hotly debated in the ecological literature (Araújo, M. B. and Luoto, M. 2007, Araújo, M. B. and Rozenfeld, A. 2013, Gotelli, N. J. et al. 2010, Heikkinen, R. K. et al. 2007, Kroll, A. J. et al. 2014, Meier, E. S. et al. 2010, Ovaskainen, O. et al. 2010).

Here, we analysed co-occurrence in river bird communities across Great Britain and assessed the degree to which patterns could be attributed to shared environmental responses. This allowed us to test alternative community assembly hypotheses: a null hypothesis of random assembly, environmental filtering, interspecific interaction, and a pluralistic model of assembly. However, our discussion is not limited to mechanistic models of community assembly but we also consider the influence of the filtering of species from regional species pools on community structure. We found that species co-occurred more often than random and that cooccurrence patterns were primarily explained by shared environmental responses. Species co-occurrences were strongly related to the environmental predictor variables, whilst residual correlation was comparatively weak. These findings suggest that environmental filtering is the dominant mechanism operating to structure river bird assemblages and that interspecific interaction is reduced to an ancillary role. Whilst our results corroborate the findings of some studies in suggesting that species interactions are of secondary importance relative to habitat structure and resource availability (Börger, L. and Nudds, T. D. 2014, Gotelli, N. J. et al. 1997, Kroll, A. J. et al. 2014, Mattsson, B. J. et al. 2013, Petchey, O. L. et al. 2007), they contrast markedly with others (Gotelli, N. J. et al. 2010, Heikkinen, R. K. et al. 2007, Robertson, O. J. et al. 2013, Sebastian-Gonzalez, E. et al. 2010). One reason for this might be that we focussed on bird communities on riverine systems which are highly dynamic, disturbance-prone environments in which river flow is the dominant driver of changes in ecosystem structure (Junk, W. J. et al. 1989). High levels of environmental disturbance result in rapid turnover of biota through flow-induced

spatial and temporal heterogeneities in the prevalence of key foraging and breeding

habitats (Ward, J. V. et al. 2002). Perturbations in aquatic-riparian food webs also

346

347

348

349

350

351

352

353

354

355

356

357

358

359

360

361

362

363

364

365

366

367

368

369

cause irregularity in ecological processes such as production and biotic interactions (Ledger, M. E. et al. 2012). Heterogeneity in external factors such as land-cover complexity and the availability of foraging habitat also exerts strong formative pressure on the structure of bird communities (Devictor, V. et al. 2007, Mattsson, B. J. et al. 2013, Robertson, O. J. et al. 2013). Consequently, in riverine systems the influence of abiotic factors in constraining bird species distributions is likely to be strengthened, with species interactions such as competitive filtering reduced to a minor role. Moreover, it is plausible that those species with the greatest exposure to abiotic disturbance are the least likely to be connected to other species in the community and so are less influenced in distribution by interspecific interaction (Araujo, M. B. et al. 2011). Evidence from bird communities indicates that when species richness is depressed by disturbance, abiotic constraints are more important than species interactions in structuring species assemblages (Belmaker, J. et al. 2012). However, the predictive performance of the JSDM varied between bird

foraging guilds, with residual correlation most positive amongst riparian species. It may be the case that a few species in the community have close links to others in the community network with most species poorly connected to others, with a tendency for individualistic behaviour and partitioning (Araújo, M. B. et al. 2011). This perhaps indicates a reduced impact of environmental filtering and an enhanced influence of species interactions on riparian species distributions. Therefore, a pluralistic model of community organisation may be more appropriate for some avian taxa. Riparian species such as the common redshank, Eurasian oystercatcher and northern lapwing forage in ephemeral patches of habitat. It is plausible that competition amongst these functionally similar species for transitory food resources is likely and that this would lead to segregation (MacArthur, R. and Levins, R. 1967). Yet, co-occurrence between

these species was observed to be positive, suggesting the operation of an alternative model of interspecific interaction such as heterospecific attraction. However, positive residual correlation could also be caused by an unmeasured variable for habitat quality which might influence species distributions. It is possible that the habitat, river flow and climate variables used to model species distributions did not fully capture variability in ephemeral habitat patches, which are utilised by species such as the Eurasian oystercatcher and northern lapwing. It is also possible that an enhanced influence of interspecific interaction might have been detected by using abundance rather than presence data as there is some evidence that mechanisms underlying abundance variations occur at community scales where species interaction is greatest (Boulangeat, I. et al. 2012).

Hetereospecific attraction can lead to positive patterns of co-occurrence as species preferentially select patches of habitat already occupied by heterospecifics, whereby individuals may use the presence of other species as an indicator of patch quality (Monkkonen, M. et al. 1990). Such patterns have previously been observed in some waterbird species including riparian waders (Sebastian-Gonzalez, E. et al. 2010). Facilitative processes like heterospecific attraction are thought to be more prevalent in disturbed environments by increasing the average fitness of individuals of a species that could not survive in isolation (Villarreal-Barajas, T. and Martorell, C. 2009). However, this requires intraspecific to be stronger than interspecific competition, allowing for niche partitioning (MacArthur, R. and Levins, R. 1967), or for resources to be sufficiently abundant that competitive exclusion is minimised. Interspecific competition between the sympatric riparian species in our study is likely to be reduced, in part, because of the disparity between body sizes which allows for different foraging strategies and the exploitation of different food resources, thus

facilitating co-existence through spatial niche separation (Leyequien, E. et al. 2007).

Furthermore, temporal niche separation, as exhibited by nocturnally foraging northern lapwings, may also enable co-existence.

424

425

426

427

428

429

430

431

432

433

434

435

436

437

438

439

440

441

442

443

444

Patterns of random co-occurrence were relatively prevalent amongst river bird communities. There are several reasons why such patterns might be observed. First, random patterns of co-occurrence may indicate a strong influence of the filtering of species from the regional species pool which may be greatest at locations with lower productivity and reduced competitive exclusion (Houseman, G. R. and Gross, K. L. 2006). Variability in river flows exerts a strong disturbance regime that can limit productivity in aquatic-riparian environments which, in turn, prevents competitive exclusion (Bunn, S. E. and Arthington, A. H. 2002, Ledger, M. E. et al. 2012, Poff, N. L. R. et al. 1997). Consequently, the natural disturbance regime of rivers may enhance ecological filtering from species pools into local communities via heightened abiotic filtering which maintains productivity at relatively low levels. In support of this theory, the percentage of co-occurrence patterns that were random was higher at river locations with greater river flow variability than at locations with more stable river flow regimes (Figure A1). Secondly, patterns of random co-occurrence could also be caused by the operation of multiple environmental factors which cause both aggregation and segregation of species distributions, as it has been shown experimentally that contrasting environmental processes might counterbalance each other to co-produce patterns of random species co-occurrence (García-Baquero, G. and Crujeiras, R. M. 2015). Finally, patterns of random co-occurrence could also arise as a consequence of imperfect detection of some species which leads to false absences in the species presence/absence matrix (MacKenzie, D. I. et al. 2004). However, we

aimed to minimise the potential for bias caused by species detection probability by only using survey locations that were visited on several occasions.

In summary, co-occurrence patterns were primarily driven by shared environmental responses. We found limited evidence for behaviour-driven assemblage patterns in bird communities at a relatively large spatial scale.

Consequently, by underestimating the prevalence of shared environmental responses studies based solely on analyses of null model matrices or probabilistic models may overstate the influence of species interactions on community structure. However, we also found that pluralistic models of community assembly may be more appropriate for some avian taxa and that the high disturbance regimes of rivers may enhance random ecological filtering of species into avian assemblages. Therefore, our results still highlight the necessity to consider biotic interactions in the modelling of species distributions, especially in environments where gradients of disturbance exist and facilitative mechanisms such as interspecific attraction may operate to promote positive associations amongst some species. Our analyses also highlight the value of long-term and large-scale bird monitoring programmes for the collation of data that allow for macro-ecological studies of community-level interaction strengths.

#### Acknowledgements

We would like to thank Michael McCarthy, Miguel Matias, Joseph Veech and two anonymous referees for reviewing and providing comments on previous versions of this manuscript, and Nick Golding for assistance with the JSDM code. We also extend gratitude to the BTO for access to WBBS datasets, Mark Hulme and John Marchant for preparing raw data, and the many BTO volunteers who collected WBBS data during surveys between 1998 and 2012 (inclusive), with funding for the survey

- 470 provided by the Environment Agency and the BTO. National River Flow Archive
- 471 (NRFA) data were provided by the Centre for Ecology & Hydrology (CEH). This
- 472 research was made possible by financial support to AR from a National Environment
- 473 Research Council (NERC) studentship (NE/J500240/1).

475

496

497

498

#### References

- 476 Araújo, M. B. and Luoto, M. 2007. The importance of biotic interactions for modelling species distributions under climate change. Global Ecology and Biogeography 16: 743-753.
- Araújo, M. B. and Rozenfeld, A. 2013. The geographic scaling of biotic interactions.
   Ecography 37: 406-415.
- 481 Araújo, M. B. et al. 2011. Using species co-occurrence networks to assess the impacts of climate change. Ecography 34: 897-908.
- Barnagaud, J. Y. et al. 2014. Habitat filtering by landscape and local forest composition in native and exotic New Zealand birds. Ecology 95: 78-87.
- Baselga, A. et al. 2012. Global patterns in the shape of species geographical ranges reveal range determinants. Journal of Biogeography 39: 760-771.
- Bayliss, A. C. and Davies, H. N. 2003. Evaluation of the Mapping and Assessment of Urban and Suburban Areas. Final Report, R&D Project FD 1919. Report to Defra/Environment Agency. 35 pp.
- Belmaker, J. and Jetz, W. 2013. Spatial scaling of functional structure in bird and mammal assemblages. American Naturalist 181: 464-478.
- Belmaker, J. et al. 2012. Global patterns of specialization and coexistence in bird assemblages. Journal of Biogeography 39: 193-203.
- Börger, L. and Nudds, T. D. 2014. Fire, humans, and climate: modeling distribution dynamics of boreal forest waterbirds. Ecological Applications 24: 121-141.
  - Boulangeat, I. et al. 2012. Accounting for dispersal and biotic interactions to disentangle the drivers of species distributions and their abundances. Ecology Letters 15: 584-593.
- Bunn, S. E. and Arthington, A. H. 2002. Basic principles and ecological consequences
   of altered flow regimes for aquatic biodiversity. Environmental
   Management 30: 492-507.
- 502 Cody, M. L. 1974. Competition and the Structure of Bird Communities. Princeton University Press.
- Devictor, V. et al. 2007. Functional homogenization effect of urbanization on bird communities. Conservation Biology 21: 741-751.
- Diamond, J. M. 1975. Assembly of species communities. In: Diamond, M. L. C. J.
   M. (ed), Ecology and Evolution of Communities. Harvard University Press,
   pp. 342–444.
- García-Baquero, G. and Crujeiras, R. M. 2015. Can environmental constraints
   determine random patterns of plant species co-occurrence? Ecology and
   Evolution Early View:

- Gelman, A. et al. 2008. A weakly informative default prior distribution for logistic and other regression models. The Annals of Applied Statistics 1360-1383.
- Gelman, A., Yu-Sung, S., Yajima, M., Hill, J., Pittau, M.G., Kerman, J., Zheng, T.,
   Dorie, V. 2014. arm: Data Analysis Using Regression and
   Multilevel/Hierarchical Models. R package version 1.7-07. Available at
   <a href="http://CRAN.R-project.org/package=arm">http://CRAN.R-project.org/package=arm</a>.
- Gotelli, N. J. et al. 1997. Co-occurrence of Australian land birds: Diamond's assembly rules revisited. Oikos 80: 311-324.
- Gotelli, N. J. et al. 2010. Macroecological signals of species interactions in the Danish
   avifauna. Proceedings of the National Academy of Sciences of the United
   States of America 107: 5030-5035.
- Gotelli, N. J. and McGill, B. J. 2006. Null versus neutral models: what's the difference? Ecography 29: 793-800.

528

- 525 Griffith, D. M., Veech, J.A. & Marsh, C.J. In press. co-occur: Probabilistic species co-526 occurrence analysis in R. — Journal of Statistical Software
  - Heikkinen, R. K. et al. 2007. Biotic interactions improve prediction of boreal bird distributions at macro-scales. Global Ecology and Biogeography 16: 754-763.
- Houseman, G. R. and Gross, K. L. 2006. Does ecological filtering across a
   productivity gradient explain variation in species pool-richness relationships?
   Oikos 115: 148-154.
- Hubbell, S. P. 2001. The Unified Neutral Theory of Biodiversity and Biogeography.
   Princeton University Press.
- Junk, W. J. et al. 1989. The flood pulse concept in river-floodplain systems. —
   Canadian Special Publication of Fisheries and Aquatic Sciences 106: 110-127.
- Keddy, P. A. 1992. Assembly and response rules 2 goals for predictive community ecology. Journal of Vegetation Science 3: 157-164.
- Kraft, N. J. B. et al. 2014. Community assembly, coexistence, and the environmental filtering metaphor. Functional Ecology Early view:
- Kroll, A. J. et al. 2014. Avian community composition associated with interactions
   between local and landscape habitat attributes. For. Ecol. Manage. 326: 46 57.
- Ledger, M. E. et al. 2012. Drought alters the structure and functioning of complex food webs. Nature Climate Change 3: 223-227.
- Lehikoinen, A. et al. 2013. Rapid climate driven shifts in wintering distributions of three common waterbird species. — Global Change Biology 19: 2071-2081.
- Leyequien, E. et al. 2007. Influence of body size on coexistence of bird species. —
   Ecological Research 22: 735-741.
- Luck, G. W. et al. 2013. Changes in bird functional diversity across multiple land uses: Interpretations of functional redundancy depend on functional group identity. PLoS ONE 8: e63671.
- MacArthur, R. and Levins, R. 1964. Competition, habitat selection, and character
   displacement in a patchy environment. Proceedings of the National
   Academy of Sciences of the United States of America 51: 1207.
- MacArthur, R. and Levins, R. 1967. The limiting similarity, convergence, and divergence of coexisting species. American naturalist 101: 377-385.
- Macarthur, R. H. 1958. Population ecology of some warblers of northeastern coniferous forests. Ecology 39: 599-619.
- MacArthur, R. H. 1972. Geographical Ecology: Patterns in the Distribution of
   Species. Harper & Row.

- MacKenzie, D. I. et al. 2004. Investigating species co-occurrence patterns when species are detected imperfectly. Journal of Animal Ecology 73: 546-555.
- Marchant, J. H., Noble, D.G., Leech, D.I., Freeman, S.N. 2002. River Habitat Survey
   and Waterways Breeding Bird Survey 1998–2000: Final Report. British
   Trust for Ornithology.
  - Mattsson, B. J. et al. 2013. Explaining local-scale species distributions: Relative contributions of spatial autocorrelation and landscape heterogeneity for an avian assemblage. PLoS ONE 8: e55097.
- 570 McCarthy, M. A. et al. 2012. The influence of abundance on detectability. Oikos 571 122: 717-726.
- Meier, E. S. et al. 2010. Biotic and abiotic variables show little redundancy in explaining tree species distributions. Ecography 33: 1038-1048.

569

574

575

576

577

578579

580

589

590

592

593

594

595

596

597

598

- Monkkonen, M. et al. 1990. Numerical and behavioral-responses of migrant passerines to experimental manipulation of resident tits (*Parus* spp) heterospecific attraction in northern breeding bird communities. Oecologia 85: 218-225.
- Newbold, T. et al. 2014. Functional traits, land-use change and the structure of present and future bird communities in tropical forests. Global Ecology and Biogeography 23: 1073-1084.
- Ovaskainen, O. et al. 2010. Modeling species co-occurrence by multivariate logistic regression generates new hypotheses on fungal interactions. Ecology 91: 2514-2521.
- Peron, G. and Altwegg, R. 2015. The abundant centre syndrome and species distributions: insights from closely related species pairs in southern Africa. — Global Ecology and Biogeography 24: 215-225.
- Petchey, O. L. et al. 2007. Low functional diversity and no redundancy in British avian assemblages. Journal of Animal Ecology 76: 977-985.
  - Plummer, M. 2014. rjags: Bayesian graphical models using MCMC. R package version 3-12. Available at: <a href="http://CRAN.Rproject.org/package=rjags">http://CRAN.Rproject.org/package=rjags</a>.
- Poff, N. L. R. et al. 1997. The natural flow regime. BioScience 47: 769-784.
  - Pollock, L. J. et al. 2014. Understanding co-occurrence bymodelling species simultaneously with a Joint Species DistributionModel (JSDM). Methods in Ecology and Evolution 5: 397-406.
  - Renwick, A. R. et al. 2012. Modelling changes in species' abundance in response to projected climate change. Divers. Distrib. 18: 121-132.
  - Richmond, O. M. W. et al. 2010. Two-species occupancy models: a new parameterization applied to co-occurrence of secretive rails. Ecological Applications 20: 2036-2046.
- Robertson, O. J. et al. 2013. Influence of interspecific competition and landscape structure on spatial homogenization of avian assemblages. PLoS ONE 8: e65299.
- Robinson, Q. H. et al. 2014. The application of occupancy modeling to evaluate intraguild predation in a model carnivore system. Ecology 95: 3112-3123.
- 605 Royan, A. et al. 2013. Avian community responses to variability in river hydrology.

   PLoS ONE 8: e83221.
- Royan, A. et al. 2014. River birds' response to hydrological extremes: New vulnerability index and conservation implications. Biological Conservation 177: 64-73.

- Sebastian-Gonzalez, E. et al. 2010. Testing the heterospecific attraction hypothesis with time-series data on species co-occurrence. Proceedings of the Royal Society B-Biological Sciences 277: 2983-2990.
- 613 Ulrich, W. and Gotelli, N. J. 2010. Null model analysis of species associations using 614 abundance data. Ecology 91: 3384-3397.
- Vandewalle, M. et al. 2010. Functional traits as indicators of biodiversity response to land use changes across ecosystems and organisms. Biodiversity and Conservation 19: 2921-2947.
- Vaughan, I. P. et al. 2007. Combining surveys of river habitats and river birds to appraise riverine hydromorphology. Freshwater Biology 52: 2270-2284.
- Veech, J. A. 2013. A probabilistic model for analysing species co-occurrence. Global Ecology and Biogeography 22: 252-260.
- Veech, J. A. 2014. The pairwise approach to analysing species co-occurrence. Journal of Biogeography 41: 1029-1035.

625

626

- Villarreal-Barajas, T. and Martorell, C. 2009. Species-specific disturbance tolerance, competition and positive interactions along an anthropogenic disturbance gradient. Journal of Vegetation Science 20: 1027-1040.
- Ward, J. V. et al. 2002. Riverine landscape diversity. Freshwater Biology 47: 517-628 539.
- Ward, P. and Zahavi, A. 1973. Importance of certain assemblages of birds as information-centers for food-finding. Ibis 115: 517-534.
- Weiher, E. et al. 1998. Community assembly rules, morphological dispersion, and the coexistence of plant species. Oikos 81: 309-322.
- Weiher, E. and Keddy, P. A. 1995. The assembly of experimental wetland plantcommunities. — Oikos 73: 323-335.

#### **Tables**

Land use variables

**Table 1.** Description of the four environmental variables used in the joint species distribution model (JSDM) as predictors of the occurrence and co-occurrence patterns of 19 river bird species across mainland Great Britain. River flow variables were defined *a priori* according to Royan, A. et al. (2013) and were calculated using long-term averages between 1998 and 2011 (inclusive) with data obtained from the National River Flow Archive (NRFA). Land use data were produced by the Centre for Ecology and Hydrology (CEH) as part of the Countryside Survey 2000 and were derived from the satellite-generated Land Cover Map 2000 (LCM2000) with a resolution of 50 m. Climate data were obtained from the UKCP09 gridded observation datasets at 5 × 5 km resolution at monthly timescales and climate variables were calculated using data between January 1998 and December 2011 (inclusive).

Environmental variables	Calculation	Description
Climate variables		
Total Breeding Season	Sum of April to July	During birds' breeding
Rainfall (mm)	(inclusive) rainfall	season.
	between 1998 and 2011	
	(inclusive)	
Mean Breeding Season	Mean value of April to	During birds' breeding
Temperature (°C)	July (inclusive)	season.
	temperature between 1998	
	and 2011 (inclusive)	

Area of woodland (%)	Broadleaved/mixed	Catchment that is
	woodland	woodland habitat
	Coniferous woodland	
Area of arable land (%)	Arable cereals	Catchment that is arable
	Arable horticulture	habitat
	Arable non-rotational	
Area of grassland (%)	Improved grassland	Catchment that is
	Set-aside grass	grassland habitat
	Neutral grass	
	Calcareous grass	
	Acid grassland	
	Bracken	
	Fen, marsh, swamp	
Area of heathland (%)	Montane habitats	Catchment that is
	Dense dwarf shrub heath	heathland habitat
	Open dwarf shrub heath	
	Bog (deep peat)	
Area of urban land (%)	A composite index based	Urban extent within the
	on a refined version of the	catchment boundary.
	data for the LCM2000	
	classes Suburban, Urban	
	and Inland bare ground	
	(Bayliss, A. C. and Davies,	
	H. N. 2003)	

River flow variables

Flow Variation (m <sup>3</sup> /sec)	Standard deviation of daily	Variability around average
	discharge.	flow conditions.
Mean Daily Flow	Mean value of daily	Average flow magnitude,
(m <sup>3</sup> /sec)	discharge divided by	correcting for river size.
	median of daily discharge.	
Three Day Maximum	Average 3-day maximum	Variability around high
(m <sup>3</sup> /sec)	divided by median	flows and the deviation of
	discharge.	high flows from the
		median.
Three Day Minimum	Average 3-day minimum	Variability around low
(m <sup>3</sup> /sec)	divided by median	flows and the deviation of
	discharge.	low flows from the
		median.

#### Figures headings

**Figure 1.** Map showing the distribution of the 100 Waterways Breeding Bird Survey (WBBS) survey locations used to derive presence/absence data for 19 river bird species.

**Figure 2.** Heatmap visualisation showing pairwise associations between 19 river bird species calculated according to the probability model of species co-occurrence (Griffith, D. M., Veech, J.A. & Marsh, C.J. In press). Significant positive (negative) associations are displayed where species co-occurred more (less) frequently than by chance, with an alpha threshold of 0.05.

**Figure 3.** Modelled environmental and residual correlations between pairs of all 19 river bird species (i.e. 171 pairs) averaged across all 100 WBBS locations. The error bars display 95% confidence intervals for the mean modelled environmental and residual correlations. The points are coloured blue where both environmental and residual correlation are positive, and environmental correlation is stronger (i.e. closer to 1), and coloured orange where both environmental and residual correlation are negative, and environmental correlation is stronger (i.e. closer to -1).

**Figure 4.** Network diagrams showing modelled environmental correlation (a & c) and residual correlation (b & d) between 19 river bird species, where the black lines represent positive correlations at Rs > 0.5 and Rs > 0.7.

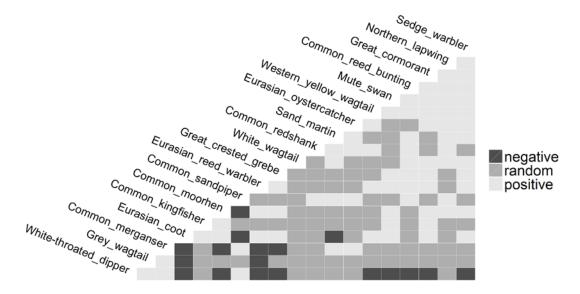
Figure 5. A dendrogram, created by converting the matrix of species occurrences across the 100 river locations into a binary dissimilarity matrix and by then using hierarchical cluster analysis, to show how species occurrences cluster in a manner consistent with three guilds based on foraging habitat and resource acquisition: slow-flowing species, fast-flowing species, and riparian species.

Figure 6. Box-and-whisker plots of the mean residual correlation between species within three guilds based on foraging habitat and resource acquisition: fast-flowing, riparian, and slow-flowing (see Results section for the species composition of guilds). In each box the thick black line represents the median average residual correlation between species and the limits illustrate the interquantile range from the first quartile (i.e. 25<sup>th</sup> quantile) to the third quartile (i.e. 75<sup>th</sup> quantile). The whiskers extend to the minimum and maximum values, excluding outliers.

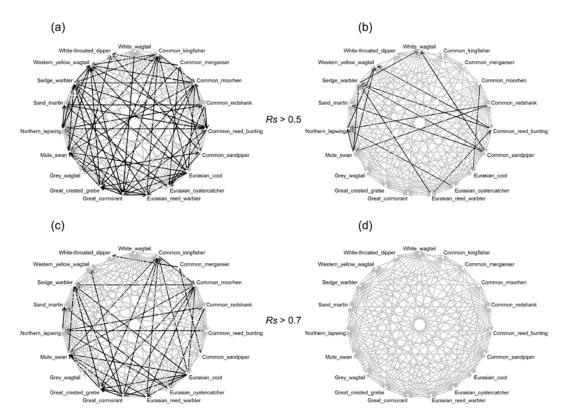


688 Figure 1

689 Figure 2 

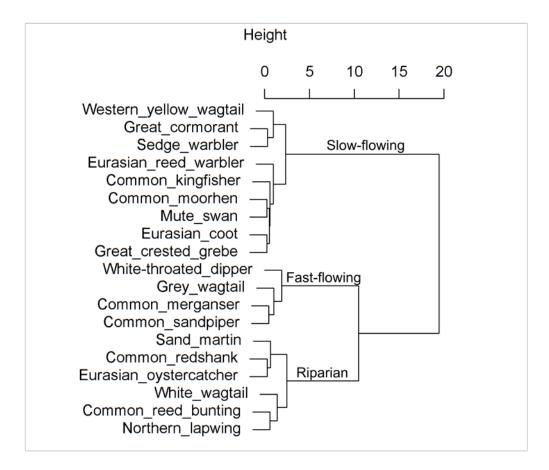


694 695 Figure 3



697 Figure 4

698 Figure 5 



701 Figure 6 

