SUPPLEMENTARY INFORMATION

Theodorou et al. Urban areas as hotspots for bees and pollination but not a panacea for all insects

Supplementary Figures

Mean nearest taxon distance (MNTD)

Supplementary Figure 1. Mean nearest taxon distance (MNTD) at N=9 flower-rich rural *versus* N=9 paired flower-rich urban sites. means \pm SE are shown; LMM: ns, not significant.

Supplementary Figure 2. The proportion of green land (Forest, semi-natural vegetation, parks and allotments) and edge density (ecotones) in the 1 km surrounding the N=9 flower-rich rural *versus* N=9 paired flower-rich urban sites (green land-uses, LMM, t=-0.080, P=0.938; edge density, LMM, $t=0.487$, P=0.632). ns, not significant.

Supplementary Figure 3. Non-metric multidimensional scaling (NMDS) ordination of (a) overall insect communities and (b) Hymenoptera communities. Stress levels are reported in the top right of the figures. Results of the *adonis* analyses of differences in community composition are reported in the top left of the figure.

Supplementary Figure 4. Main families sampled (>3% of OTUs per order) for each insect order: (a) Diptera n = 342 OTUs, (b) Coleoptera n = 53 OTUs, (c) Hymenoptera n = 116 OTUs, and (d) Lepidoptera $n = 81$ OUTs. Highlighted in yellow are the hoverflies (family: Syrphidae) within the Diptera (a), and five families of bee (members of the Anthophila) within the Hymenoptera (c).

Supplementary Figure 5. Results from a previous study^{1,2} conducted in 2013 and demonstrating the positive correlations across plant species in seed set of (a) *Borago officinalis*, (b) *Sinapis alba*, (c) *Trifolium pratense* and (d) *Trifolium repens* across an agricultural-urban landscape gradient around the city of Halle, Germany. The statistical results of the correlations are reported in Supplementary Table 5. Means \pm SE as well as individual (overlapping) data points are shown.

Supplementary Figure 6. Rarefaction curves showing the level of saturation of OTU richness of insects in pan trap samples.

Supplementary Tables

Supplementary Table 1. Total patch flower richness (number of co-flowering plant species), abundance of flowers in 10 x 1 $m²$ quadrats, and number of inflorescences of co-flowering *Trifolium pratense* plants within a 200 m buffer at urban and rural sites.

Supplementary Table 2. Best generalised linear model (GLM) explaining OTU richness of Hymenoptera, bee (Anthophila, a subset of Hymenoptera), Coleoptera, Diptera, hoverflies (Syrphidae, a subset of Diptera) and Lepidoptera for N=9 urban and N=9 paired rural sites at the local (patch) and landscape scales (1,000 m radius for all Hymenoptera and Coleoptera, 250 m for all Diptera and Lepidoptera). We used AICc for model selection. Due to low sample size (mean=1.44±1.01 SD), Lepidoptera OTU richness was not modelled within urban sites. We did not find any significant predictors of Syrphidae OTU richness in rural or urban ecosystems.

10

*, P<0.05; **, P<0.01; R^2 _{adj}= Proportion of the variance in the dependent variable that is predicted from the independent variable(s).

S.E. = standard error; P = statistical significance; Hymenoptera PSV = Hymenoptera phylogenetic species variability

Supplementary Table 5a. Table of path coefficients from the best-fit piecewise SEM of the relationships between local flower richness, *Trifolium pratense* seed set, *T. pretense* flower visitation rate, local (patch) and landscape factors and flying insect OTU richness across the nine rural sites. The SEM showed stable fit to our data (Fisher's C=7.096, d.f.=6, P=0.312).

S.E. = standard error; P = statistical significance; Hymenoptera PSV = Hymenoptera phylogenetic species variability

S.E. = standard error; P = statistical significance; Hymenoptera PSV = Hymenoptera phylogenetic species variability

Supplementary Table 6. Proportion of the main land cover classes at urban sites.

Supplementary Table 7. Proportion of the main land cover classes at rural sites.

Supplementary Table 9. Results from a previous study^{1,2} conducted in 2013 around the city of Halle (Germany) on the pollination of four plant species at nine independent sites across an agricultural-urban landscape gradient. Pearson correlation coefficients (*r*, below diagonal) of the relationship between seed set of *Borago officinalis*, *Sinapis alba*, *Trifolium pratense* and *Trifolium repens* and significance (uncorrected *P* values, above diagonal). There was a positive correlation in seed set for each pair of plant species across the nine sites except *B. officinalis*-*T. repens*, for which the relationship was non-significant (r=0.688, P=0.119).

Supplementary Table 10. Land-cover types provided by land cover data obtained from Geofabrik GmbH and average % cover across all sites (rural and urban) at 1,000 m radius from the site centre. We split the Geofabrik feature class 'park' into two for our analyses based on ground truthing: botanical park and public park; we also combined four Geofabrik feature classes into one for our analyses: meadow, nature reserve, grass and scrub were combined into seminatural.

Supplementary Table 11. Pearson correlation coefficients (r) of the relationship between insect species (OTU) richness for the orders Diptera, Lepidoptera, Coleoptera and Hymenoptera, and landscape diversity (measured as Shannon-Weiner diversity of land-uses) at increasing area (given as radius in metres) from the centre of a site. The largest absolute correlation coefficient is given in bold.

Site	Number of reads	OTU richness
Rural Halle	6475	76
Urban Halle	6732	80
Rural Leipzig	4139	122
Urban Leipzig	5282	93
Rural Jena	3877	105
Urban Jena	3417	68
Rural Dresden	4344	90
Urban Dresden	3528	53
Rural Chemnitz	6729	77
Urban Chemnitz	4569	57
Rural Braunschweig	4392	81
Urban Braunschweig	5133	73
Rural Berlin	3561	74
Urban Berlin	5080	61
Rural Potsdam	4670	71
Urban Potsdam	4183	71
Rural Göttingen	4723	83
Urban Göttingen	4077	62
Total	8 4 9 1 1	

Supplementary Table 12. Number of reads and OTU richness per site.

Supplementary Table 13. Pearson correlation coefficients (*r*, below diagonal) of the relationship between detected OTU richness, rarefied OTU richness and extrapolated total OTU richness (Chao 1) and significance (uncorrected *P* values, above diagonal).

Supplementary Table 14. Number of Sanger sequence-generated OTUs (Sanger OTUs) for each mock community and 454-generated OTUs successfully blasted to Sanger OTUs at 97% similarity, using both the original pipeline of Yu et al.⁴ and our pipeline, as well as OTUs we detected with our pipeline that did not match those of Yu et al.¹⁰.

Supplementary Methods:

Metabarcoding: sample preparation, PCR amplification and 454-pyrosequencing for OTU and species assignment

To generate mitochondrial DNA (*cytochrome oxidase I*) sequences for the identification of flying insect OTUs (our proxy for species) whilst avoiding biases in DNA extraction and PCRs, we used standard methods recommended for insect barcoding (http://ccdb.ca/resources/)³, as adapted for metabarcoding⁴⁻⁶. Insect samples from each site were washed, dried and weighed and, per 10 g of biological material, we added 5 ml of sterile ddH2O. The entire insect sample per site was then homogenized using a semi-automated Homex6 homogenizer (Bioreba AG, Reinach, Switzerland), after which 15% of the solution was used for genomic DNA extraction with a tissue DNA maxi kit (Omega Bio-tek, Norcross, USA). This gave us one DNA extract (DNA soup) per study site.

The quantity and quality of purified DNA was assessed using an Epoch microplate spectrophotometer (BioTek, Winooski, USA); all samples contained > 100 ng/ μ L DNA of high purity $(A_{260}/A_{280} = 1.7$ -2.0 for all 18 samples). Each sample was PCR amplified with universal primers LCO-1490 (5'-GGTCAACAAATCATAAAGATATTGG-3') and HCO-2198 (5'- TAAACTTCAGGG-TGACCAAAAAATCA-3')⁷ that target the approximately 650 bp 'barcode' region of the mitochondrial cytochrome *c* oxidase subunit I (COI) and that have been successfully used to amplify the barcode region of German insects⁸. The standard Roche Aadaptor and a unique 10 bp MID (Multiplex IDentifier) tag for each sample were attached to the LCO primer. Each sample was amplified in three independent reactions to reduce PCR bias;

PCR products were then pooled per site. PCR reactions were performed in 20 µL volumes consisting of 2X Promega PCR buffer with 3.0 mM MgCl₂ (Promega, Fitchburg, USA), 0.4 μ M of each primer, 0.2 mM dNTP, 0.5 U GoTaq Polymerase (Promega, Fitchburg, USA) and 60 ng of template DNA. We performed PCRs for each site separately to avoid cross contamination. Within each PCR reaction, a negative control lacking DNA template was always included to detect contamination from extraneous sources such as PCR reagents; contamination was never detected. PCRs were performed with a Biometra TProfessional basic gradient thermocycler (Biometra, Göttingen, Germany) using the following thermal cycling program: initial denaturation at 94 \degree C for 3 min, followed by 35 cycles of 94 \degree C for 30 s, annealing temperature at 51^oC for 45 s and 72^oC for 1 min, plus a final extension step of 72^oC for 8 min. We have found these PCR criteria (master mixes of PCR components, cycling temperatures and durations) to allow successful amplification of the German bee fauna⁹. PCR products were quantified and visualized through a QIAxcel automated capillary electrophoresis system (Qiagen, Hilden, Germany). No signal was visible in the negative (no DNA template) control for the PCR reactions whereas a single, clear product of ca. 650 bp was visualised in all samples.

For 454-pyrosequencing, all PCR products were purified using a QIAquick PCR purification kit (Qiagen, Hilden, Germany), quantified using the Quant-iT PicoGreen dsDNA Assay kit (Invitrogen, Grand Island, USA) and diluted down to $1x10⁹$ molecules/ μ l. Pooled and labelled PCR products were sequenced on a 454/Roche GS-FLX Plus System (Macrogen, Seoul, Korea) next generation sequencing (NGS) machine.

DNA concentrations at each measurement step (pre- and during NGS library preparation) were consistent across samples. For example, we used 60 ng of template DNA in all PCR reactions to amplify the COI region and all purified PCR products were diluted down to $1x10^9$ molecules/ μ l prior to NGS sequencing. Potential taxon-specific PCR amplification biases should therefore have been consistent across all samples and not have influenced downstream bioinformatics or statistical analyses.

Bioinformatics analysis of metabarcoding data

We employed well-established and robust software and pipelines for OTU assignment using metabarcode data 10^{-15} that we independently verified and that gave consistent OTU and species assignments when altering parameter settings of bioinformatics algorithms.

Step 1: *quality filtering.* Low quality reads were removed using a strict quality filter applied with FlowClus 1.1¹⁰ and cutadapt 1.14¹¹. Reads were retained if they (i) matched one of the MID tags with one mismatch allowed, (ii) contained the forward primer with 4 mismatches allowed, (iii) were at least 395 bp long, (iv) had at least a mean Phred score of 30 on the trimmed length, (v) did not hold any ambiguous nucleotides, (vi) had homopolymers no longer than 12 bp, and (vii) had a flowgram length of at least 360, as previously advised for 454 GS FLX reads¹². The reads were subsequently denoised using FlowClus, which has been shown to recover sequences with a lower error rate and be more easily applicable to large sequence datasets than the original 454 denoising algorithm AmpliconNoise^{10,12}. The denoised and quality filtered reads were trimmed to their first 395 bp and potential chimeras were removed using UCHIME $4.2.40^{13}$, as implemented in MOTHUR 14 .

Step 2: de novo *clustering and taxonomic assignment.* Quality filtered reads were de-replicated, sorted in decreasing order of abundance and clustered into OTUs with a global threshold of 97% similarity using the cd-hit-est program¹⁵. The most abundant read in each OTU was selected as the representative sequence. Representative sequences were again used to detect and discard putative chimeric OTUs using UCHIME.

In order to assign taxonomically the reads, we created a database of COI reference sequences by first selecting all GenBank entries (release 211), accessed on $06.01.2016^{16}$; with the query "COI AND 500:10000[Sequence Length]". Reference sequences were kept if they had taxonomic information at family, genus and species ranks and if they did not contain any ambiguous nucleotide. Ambiguous species annotations (e.g. sp., cf., aff., nr., n.sp., pr.) were all normalized to "sp.". Only one sequence was conserved for each unique taxonomic path (including 16 labelled ranks from superkingdom to species). The final reference database contained 425,824 sequences, of which 217,544 were Insecta (763 families, 14,857 genera). Then, all de-replicated reads were taxonomically assigned using the naïve Bayesian classifier¹⁷ at a consensus threshold of 60%. The OTUs were finally assigned to the longest taxonomic path shared by at least 60% of their reads. As most GenBank COI sequences were produced with the same primer pair as used in our study and as end-gaps have no effect on the naïve Bayesian classifier assignments, it was not necessary to cut our reference database to the amplified region.

Step 3: *clean-up.* Singleton OTUs (N=395), which have a high probability of originating from sequencing errors, were removed from the data set. Non-target taxa OTUs (N=38; bacteria, fungi, unclassified eukaryotes, Mollusca, Nematoda, Arachnida) were also excluded. The

remaining 592 OTUs, representing insects (orders: Diptera, Lepidoptera, Coleoptera and Hymenoptera; Supplementary Figure 4, Supplementary Dataset), and accounting for a total of 84,911 sequence reads, were used for further statistical analysis. The number of total reads per site is shown in Supplementary table 12. The number of reads was not correlated with OTU richness across our dataset (Pearson's product-moment correlation *r*=0.033, N=18, P>0.05), suggesting we had sufficient reads to saturate our OTU assessment per site (Supplementary Figure 6). Detected number of OTUs, rarefied OTU richness and extrapolated total OTU richness (Chao 1) were highly correlated (P<0.001, Supplementary Table 13)

To verify independently our *de novo* clustering using cd-hit, we also used VSEARCH¹⁸, a 56-bit reimplementation of the well know 32-bit USEARCH-UCLUST 19 , to cluster and assign quality filtered and de-replicated reads. VSEARCH generated a larger number of OTUs (655 compared to 592 with UCHIME). However, there was a high correlation between the number of OTUs generated by the two methods across our 18 sites (Pearson's r=0.99; P<0.001). This suggests that our pipeline for *de novo* clustering and assignment of insect OTUs was robust. Pipeline scrips are available in a figshare Digital Repository [https://doi.org/10.6084/m9.figshare.10304795.v1].

Most of the 592 flying insect OTUs could be assigned to a species; Coleoptera: 53 OTUs, 12 of which were not assigned to species; Diptera: 342 OTUs, 225 of which were not assigned to species; Hymenoptera: 116 OTUs, 31 of which were not assigned to species; Lepidoptera: 81 OTUs, 16 of which were not assigned to species. For the well sampled bee species of Germany, we could assign species names to 40 of 46 OTUs and assign the remaining 6 OTUs to a unique taxon within a genus, which is consistent with public databases (e.g. NCBI) comprising 503 fully

'compliant' barcoded bee species of the 571 currently recognised German species 20 . Species names were checked against faunal lists to confirm their presence in the study region (Coleoptera: expert opinion, Matthias Seidel; Diptera: expert opinion, Martin Musche; Hymenoptera: expert opinion, co-author Paxton; Lepidoptera: expert opinion, co-author Settele). An OTU table with species names is available in a figshare Digital Repository [https://doi.org/10.6084/m9.figshare.10304795.v1].

To test the robustness of our pipeline, we used the raw sequence data (Sanger and *454*) from seven mock arthropod community datasets provided by Yu et al.⁴. With our pipeline applied to Yu et al.'s $4\,$ 454 pyrosequencing dataset, we were able to recover a large proportion (88.5%) of the original Sanger sequenced taxonomic information (Supplementary Table 14). This suggests that our bioinformatics pipeline was well able to capture the diversity of flying insect species in a 454-NGS dataset and was a good compromise between over- and under-splitting.

Supplementary References

- 1. Theodorou, P. *et al.* The structure of flower visitor networks in relation to pollination across an agricultural to urban gradient. *Funct. Ecol.* **31**, 838–847 (2017).
- 2. Theodorou, P. *et al.* Pollination services enhanced with urbanization despite increasing pollinator parasitism. *Proc. Biol. Sci.* **283**, 20160561 (2016).
- 3. Hebert, P. D. N., Cywinska, A., Ball, S. L. & deWaard, J. R. Biological identifications through DNA barcodes. *Proc. R. Soc. London B Biol. Sci.* **270**, (2003).
- 4. Yu, D. W. *et al.* Biodiversity soup: metabarcoding of arthropods for rapid biodiversity assessment and biomonitoring. *Methods Ecol. Evol.* **3**, 613–623 (2012).
- 5. Ji, Y. *et al.* Reliable, verifiable and efficient monitoring of biodiversity via metabarcoding. *Ecol. Lett.* **16**, 1245–1257 (2013).
- 6. Shokralla, S. *et al.* Massively parallel multiplex DNA sequencing for specimen identification using an Illumina MiSeq platform. *Sci. Rep.* **5**, 9687 (2015).
- 7. Folmer, O., Black, M., Hoeh, W., Lutz, R. & Vrijenhoek, R. DNA primers for

amplification of mitochondrial cytochrome c oxidase subunit I from diverse metazoan invertebrates. *Mol. Mar. Biol. Biotechnol.* **3**, 294–9 (1994).

- 8. Geiger, M. *et al.* Testing the Global Malaise Trap Program How well does the current barcode reference library identify flying insects in Germany? *Biodivers. Data J.* (2016). doi:10.3897/BDJ.4.e10671
- 9. Radzevičiūtė, R. *et al.* Replication of honey bee-associated RNA viruses across multiple bee species in apple orchards of Georgia, Germany and Kyrgyzstan. *J. Invertebr. Pathol.* **146**, 14–23 (2017).
- 10. Gaspar, J. M. & Thomas, W. K. FlowClus: efficiently filtering and denoising pyrosequenced amplicons. *BMC Bioinformatics* **16**, (2015).
- 11. Martin, M. Cutadapt removes adapter sequences from high-throughput sequencing reads. *EMBnet.journal* **17**, 10 (2011).
- 12. Quince, C., Lanzen, A., Davenport, R. J. & Turnbaugh, P. J. Removing noise from pyrosequenced amplicons. *BMC Bioinformatics* **12**, (2011).
- 13. Edgar, R. C., Haas, B. J., Clemente, J. C., Quince, C. & Knight, R. UCHIME improves sensitivity and speed of chimera detection. *Bioinformatics* **27**, 2194–200 (2011).
- 14. Schloss, P. D. *et al.* Introducing mothur: open-source, platform-independent, communitysupported software for describing and comparing microbial communities. *Appl. Environ. Microbiol.* **75**, 7537–41 (2009).
- 15. Fu, L., Niu, B., Zhu, Z., Wu, S. & Li, W. CD-HIT: accelerated for clustering the nextgeneration sequencing data. *Bioinformatics* **28**, 3150–2 (2012).
- 16. Benson, D. A. *et al.* GenBank. *Nucleic Acids Res.* **41**, D36-42 (2013).
- 17. Wang, Q., Garrity, G. M., Tiedje, J. M. & Cole, J. R. Naive Bayesian classifier for rapid assignment of rRNA sequences into the new bacterial taxonomy. *Appl. Environ. Microbiol.* **73**, 5261–7 (2007).
- 18. Rognes, T., Flouri, T., Nichols, B., Quince, C. & Mahé, F. VSEARCH: a versatile open source tool for metagenomics. *PeerJ* (2016). doi:10.7717/peerj.2584
- 19. Edgar, R. C. Search and clustering orders of magnitude faster than BLAST. *Bioinformatics* (2010). doi:10.1093/bioinformatics/btq461
- 20. Schmidt, S., Schmid-Egger, C., Morinière, J., Haszprunar, G. & Hebert, P. D. N. DNA barcoding largely supports 250 years of classical taxonomy: identifications for Central European bees (Hymenoptera, Apoidea). *Mol. Ecol. Resour.* **15**, 985–1000 (2015).