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Identifying a species is a complex task that relies upon a combination of visual and auditory cues. Observers must be able to process impressions of shape, size, and color under variable conditions. As this process takes place, the observer must compare these impressions against a list of species most likely to occur at that species and constantly recalibrate until the two agree, and the species is correctly identified. Humans can make this difficult computation of classifying organisms. And for birds, tens of thousands of people do this every day for fun.

For more than two hundred years the public has contributed significantly to our understanding of bird identification, distribution, and abundance [1]. In this tradition, eBird (<http://ebird.org/>) is a citizen science project that takes numerous information technologies to engage a global network of bird observers to submit observations of birds via the Internet or through a variety of mobile devices. These amassed observations provide scientists, researchers, and conservationists with data about bird distribution and abundance across varying species and regions. All data are free and readily accessible through the Avian Knowledge Center. eBird data have been used in a wide variety of applications, from highlighting the importance of public lands in conservation [5] to studies on evolutionary biogeography [7].

eBird is part of the growing field of human computation, which focuses on using human intelligence to solve computational problems that are beyond the capabilities of existing artificial intelligence algorithms [8]. Many of these services include game-based approaches such as the ESP Game, which has been used to generate image labels [9]; FoldIt, which attempts to predict the structure of a protein using the advantage of humans' puzzle solving abilities [10]; Galaxy Zoo, which

than 200,000 participants to classify more than 100 million galaxies Sloan Digital Sky Survey [11]; and reCAPTCHA, which provides security on the web while transcribing old print material one word at a time [12]. In our experiences in developing the eBird network of volunteers, who provide an open data resource containing the most current and complete information on bird distribution, migratory pathways, population trends, and use.

Know Your Community

eBird uses data collection protocols that match the way birders go about collecting fundamental data gathered with each observation are: species, date, and whether all species detected are included on the checklist. Most observations include counts of individuals for each species, and basic information that identifies the count and describes how the count was conducted (start time, duration, and location). We chose this relatively simple approach to survey design in order to maximize the number of participants, as increasing complexity of protocols in citizen science tends to decrease the number of participants in those projects [13]. The eBird protocol captures the majority of eBird observations, but birders also have the option to provide additional detailed information for each observation, such as breeding behavior, and additional comments.

A significant aspect of the data that birders contribute to eBird is that each observation has an exact date and is linked to a point on the map. This provides eBird data with a variety of covariate data that potentially influence bird distribution, such as weather, climate, habitat, and human population density. This allows the community to focus on what it does best: finding, identifying, and counting birds.

While we encourage all eBird contributors to conform to the protocols that have been developed, we also allow users to enter data in a variety of other ways. These observations may be of reduced analytical value. For example, users can gather detailed location and temporal specificity, historic data often collected at the county or even state level. Other counts may be entered that do not include dates or that simply report incidental or random observations of birds. Flexibility in data collection increases initial involvement, and once involved with eBird, we can encourage users toward improved data collection techniques. We have found that by providing incentives and training on the eBird web site, we can convert most casual observers into higher quality data collectors who use the effort-based protocols (see below).

Maximizing Participation

Initial reaction to eBird from the birding community was lukewarm, but there was a steady growth in participation. Recognizing this, we modified our development strategy on building tools that appeal to, and provide a service for, birders. For example, we allowed birdwatchers to keep track of their observations, view their own data, and compare their observations with others, we built upon the esta

birding community. The result was rapid and sustained growth. For six years, eBird's usage has increased dramatically, with the number of submissions to exceed 1.7 million from more than 210 countries in 2012.

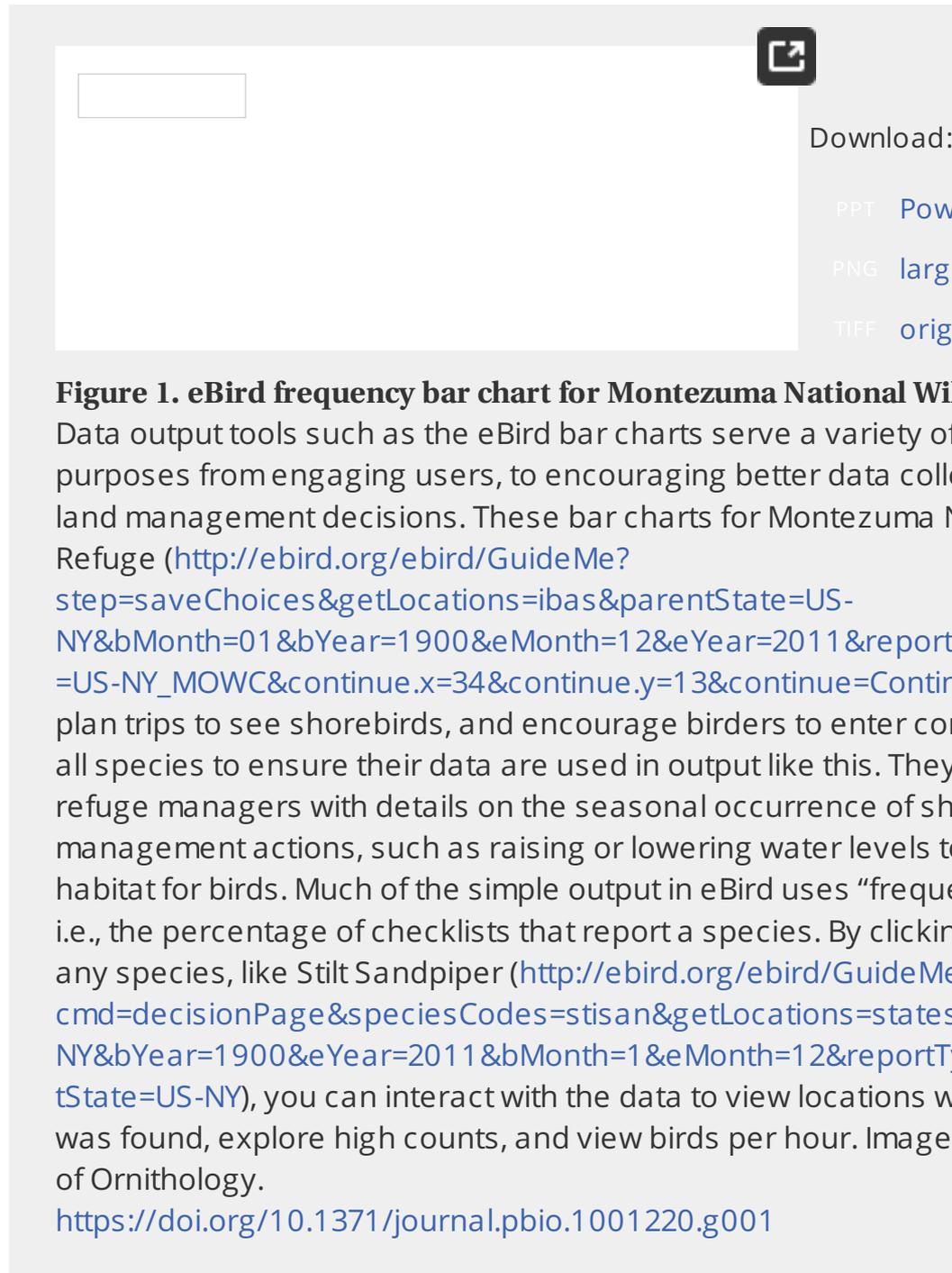
eBird participation generally follows Pareto's Law (the 80/20 law), with 80% of the data submitted by a very active subset of "power users". For example, 80% of the data submitted to eBird have come from the most active 10% of users. Using the power users allowed us to develop eBird features focused on these users, based on the belief that this would in turn increase broader participation. Over the years, we have been successful in greatly increasing participation. Having the majority of data submitted by committed, repeat users helps maintain high data quality, since these users have a strong commitment to the project, and an understanding of its best practices. This is a significant investment in the community. One should not confuse eBird power users with "power birders". While many eBird power users are indeed experts in terms of birding, detecting, and identification, many other birders with more modest participation are also power users who may only enter data from their backyard or local area. These users are often able to detect very rare species in their region on the basis of sound alone. They are often very proficient with a subset of regularly occurring birds, and they provide valuable samples of the birdlife in a defined area over time. Moreover, through their participation they develop a higher level of eBird expertise.

Changing Behavior through Tools and Visualizations

Once we are able to engage a birder in eBird, we focus on modifying their behavior so that the checklists entered are more useful for research and conservation. We use a variety of techniques to encourage higher quality submissions: education about the importance of the data, making small changes to the way you go birding can improve the value of the data, and incentives that reward participants for collecting data following more rigorous protocols.

Users receive educational training regarding the scientific reasons for using eBird, the methodologies in eBird, and learn improved data collection techniques from project leaders and volunteers. Even more effective in transforming behavior are incentives in the form of visualizations and data output tools that demonstrate how the data collection techniques benefit our users on a personal level. We use bar charts (Figure 1) to tackle one of our biggest challenges: encouraging users to submit complete checklists of all species seen and heard from a series of observations throughout the day. Bar charts are intuitive data visualizations that have been traditionally used to understand the seasonal timing of birds within a region. Finding guides often include these charts to give a visual representation of the species expected to be present in a region (usually a state). eBird has automated the creation of bar charts and similar visualizations that depict frequency of occurrence of birds. Birders could understand the seasonal patterns of bird movements and their own interest. The key is that in order for these visualizations to work best, users must submit *complete lists of all species recorded*. Birdwatchers now have a tool to help them improve their data collection techniques. Equally important, they now have a tool that encourages them to regularly enter records that help develop these bar charts. We also use other types of tabular outputs (eBird's "Top 100") that give recognition to the user.

the most checklists or observed the largest number of species. This important tool helps maintain high levels of participation by allowing users to engage with each other in competitions to submit more complete checklists and species. This combination of educational features and data output tools, along with personal reward are very powerful motivators for generating increased data quality.



Ensuring Data Quality

High data quality is critical for achieving scientific goals and for engaging users. One approach to data quality is to develop tools that allow experts to detect and identify outlier records. Editors establish a maximum number of records to be entered for every species and each month for a given region. They maintain the same kind of records that amateur and professional ornithologists keep for their regional records of bird occurrence. The tools in eBird provide our regional editors to isolate and follow-up on unusual records with

observers. These volunteers provide an enormous service to eBird greatly improves the quality of eBird data. To date, our network of experts has reviewed more than 3.5 million records.

Adding Value with Statistical Modeling

Despite the growing amount of species occurrence data collected observations are, at best, sparsely distributed in space and through motivated us to develop spatially and temporally explicit models of relating environmental features that are important to a species (e.g. elevation) to observational data. Once related, statistical models can unsampled locations and times. To facilitate this process, eBird observations are linked through observation location and time to a large number of landscape descriptors, such as remotely sensed habitat information from the National Land Cover Database, and vegetation phenology from MODIS [3]. We use this data to model species distributions with the SpatioTemporal Exploratory Model (STEM). We utilize both the broad extent and fine resolution information collected. STEM has been used to estimate year-round distributions for over 100 temperate species with a wide variety of distinct migration pathways and local habitat requirements [5]. With this information, ecologists will be better able to identify, predict, and plan conservation actions across broad landscapes.

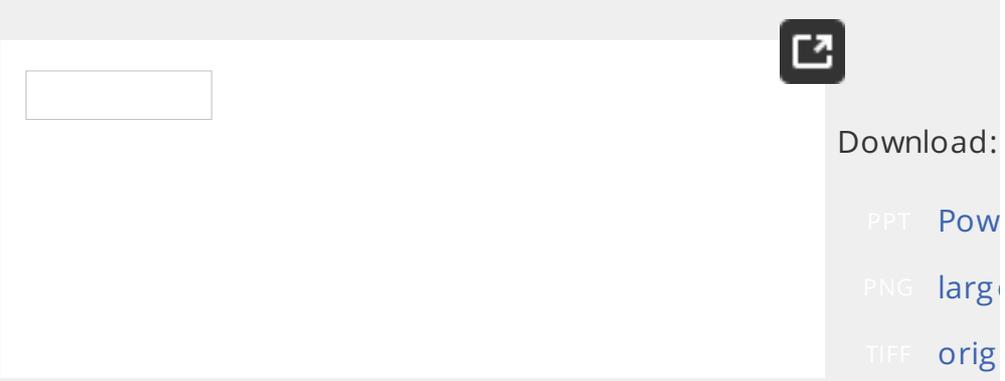


Figure 2. White-throated Sparrow distribution.

Figure 2 illustrates a STEM distribution estimate for White-throated Sparrow (*Zonotrichia albicollis*), a migratory songbird that winters in the southern US and eastern Canada and breeds in the northeastern US and eastern Canada. This occurrence map shows the probability of encountering the species (maximum 50% probability) during a birding walk starting at 7:00 A.M. for 5 January, 3 May, and 7 June. The figure is a temporal sequence of daily continental-scale distribution estimates. This type of assessment of such things as the rate of arrival and departure from breeding grounds, migratory corridors, and regions of particular importance for wintering, or migratory stopover. Over time, these same models can be used to predict and quantify changes. Image credit: Cornell Lab of Ornithology. For an animation of White-throated Sparrow distribution throughout the year, visit <http://ebird.org/content/ebird/about/occurrence-maps/white-throated-sparrow> or <https://doi.org/10.1371/journal.pbio.1001220.g002>

Building a Global Network

Many birds migrate throughout much of the Western Hemisphere and always without regard to geopolitical boundaries. While the Lab's research is focused on the Western Hemisphere, our audience actively encouraged us to expand our reach globally, thereby allowing them to keep track of their bird observations from anywhere on the planet. To accomplish this, we had to develop a broad and continuous network of global partnerships.

The goal of gathering observations of birds to further science and conservation is shared by hundreds of organizations that have local and regional expertise. By partnering with these groups, we could further our mutual goals. For example, the Mexican Comisión Nacional para el Conocimiento y Uso de la Biodiversidad (CONABIO) is interested in gathering observations of birds in Mexico. Together, we developed a version of eBird translated to Spanish for Mexico called aVerAves. All data from aVerAves and eBird are shared in the same database, but CONABIO has news and features directly relevant to a Mexican audience. CONABIO encourages birders to enter data through local promotion and engagement, and we support a community that understands and values their work.

Building upon this model partnership in Mexico, we continue to develop similar partnerships around the world. These are tailored to the individual needs of a specific region. For instance, in the United Arab Emirates, we have focused on improving data quality filters and uploading over 195,000 records from existing data sources, which provided a better way to visualize, store, and access these records than what previously existed. Other collaborations have led to the development of specific projects, ranging from grassland bird monitoring in the Chicago Wilderness (Chicago Wilderness Network eBird), to winter and summer bird atlases. For example, our partnership with the Red de Observadores de Aves y Vida Silvestre de Chile (ROC), focused on using eBird as a tool for Chilean birdwatchers and researchers. In doing so, we made eBird the primary data repository for all bird data in Chile, including data from projects conducted by the Coporacion Nacional Forestal (CONAF). We worked to bring breeding behavior codes into eBird, which has allowed the development of a nationwide breeding bird atlas for a South American country. In building on Chilean goals, we were able to deploy this functionality for the eBird project. The flexibility of eBird provides a database as well as a forum for engaging birders, and researchers and conservationists can develop to achieve local goals.

Conclusion

Identifying ecological patterns across broad spatial and temporal scales requires diverse approaches and methods for acquiring, integrating, and analyzing observational data. Engaging volunteers to collect the required data at local scales has tremendous potential to advance scientific understanding and to improve species conservation and management. Here we have developed a global network of volunteers that collects massive quantities of species observations, one of the largest citizen science projects in existence.

eBird has found a balance between data quantity and quality that has allowed us to gather a sufficient volume of useful data to provide a resource for many bird populations at spatial and temporal scales heretofore unattainable. It has also been shown that an uncomplicated protocol and appropriate rewards for participation are very important for the recruitment and retention of volunteers. Even after the data are collected and passed through a quality control process, we have found that existing methods for analysis still may not be suitable for use with such broad-scale observational data [15]. It is only through close collaboration between ecologists, statisticians, and computer scientists that the development of novel methods for extracting biological insights from large-scale data is enabled.

eBird currently provides a resource on bird occurrence that has allowed for the study of patterns of occurrence of many North American species throughout their entire life history. As the growth in participation in eBird continues, we will extend these analyses globally, and in so doing enable land managers and conservation biologists to better coordinate national and international conservation efforts using citizen science data.

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