Genetic monitoring as a promising tool for conservation and management

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In response to ever-increasing anthropogenic changes to natural ecosystems, regional, national and international organizations have established guidelines for monitoring biological diversity. Most monitoring programs, however, do not take full advantage of the potential afforded by molecular genetic markers, which can provide information relevant to both ecological and evolutionary time frames, while costing less and being more sensitive and reliable than traditional monitoring approaches. As several molecular and computational approaches are relatively new, many technical and theoretical issues remain to be resolved. Here, we illustrate how DNA and population genetic data can provide valuable information, often unattainable via other approaches, for monitoring species of management, conservation and ecological interest.

Introduction

It is becoming increasingly important to monitor unintended consequences of anthropogenic changes on natural populations. Although many national and international organizations have established principles and strategies for monitoring biological diversity [1–3], little use is made of the benefits afforded by molecular genetic markers. Meanwhile, new laboratory and statistical techniques now enable the use of molecular markers for genetic monitoring of wild populations [4–7]. Although the term ‘genetic monitoring’ is becoming widely used throughout the ecological, evolutionary, wildlife and conservation fields, there is no consensus, and some confusion, regarding its definition. To some, genetic monitoring is simply the use of genetic data to study demography or more complex evolutionary and ecological processes, whereas to others it implies systematic measurements of population genetic parameters over time.

Here, we define genetic monitoring as quantifying temporal changes in population genetic metrics or other population data generated using molecular markers. We distinguish monitoring, which must have a temporal dimension, from assessment, which reflects a snapshot of population characteristics at a single point in time. For instance, the term ‘molecular genetic monitoring’ has been used to refer to estimates of the proportion of whales from protected stocks that appeared in Asian markets [8]. We consider this an assessment because data from a seven-year period were pooled to provide a single point-estimate. A monitoring study of whale-meat markets would evaluate whether the fraction of protected species changes over time.

We separate genetic monitoring into two categories: Category I includes the use of diagnostic molecular markers for traditional population monitoring through the identification of individuals, populations, and species, whereas Category II includes the use of genetic markers to monitor population genetic parameters (Figure 1).

Category I. Diagnostic molecular markers for traditional population monitoring

Diagnostic assays that use molecular markers can identify individuals, populations, species and other taxonomic levels. We distinguish methods using molecular markers to identify individuals (Figure 1, Category Ia) from those identifying species or other groups (genera, subspecies or populations; Category Ib).

Category Ia. Identifying individuals

Molecular tags can be used in lieu of physical tags or natural markings to identify individuals. Data generated from such tags are often used in traditional population ecology models estimating abundance or vital rates. Below we describe several metrics commonly monitored in wild populations that have profited from DNA-based methods.

Abundance

The most common metric used for monitoring animal populations is abundance (Box 1). With rare species, abundance indices, such as the number of animals detected, have proven relatively inexpensive to obtain through non-invasive genetic sampling. DNA-based abundance indices are often adjusted to remove biases associated with limited sample size or imperfect detection rates. These adjustments can be accomplished by applying asymptotic analyses or using models that account for covariates associated with capture, such as season, time of day, or habitat [9,10].

Abundance can also be estimated through capture–mark–recapture (CMR) analyses (see Glossary). Traditionally, this has required physical capture and marking of a sample, followed by release and subsequent recapture of a proportion of marked animals (reviewed in Ref. [11]). Because each individual has a unique, naturally occurring genetic code (typically revealed with
Approximate Bayesian computation (ABC): a statistical framework using simulation modeling to approximate the Bayesian posterior distribution of parameters of interest (e.g. \( N_e, N_m \)), often by using multiple summary statistics (e.g. \( H_o, A, F_{ST} \)). It is computationally quicker than full Bayesian approaches, with minimal loss of accuracy and precision.

Assignment method: any of several related statistical methods that use genetic information to ascertain population membership of individuals.

Barcoding: the use of short, standardized DNA sequences, typically from a mitochondrial gene, to identify and discover species quickly and easily.

Capture-mark-recapture (CMR): field sampling method used to estimate population size or vital rates (i.e., survival, recruitment, movement).

Effective population size (\( N_{eff} \)): the size of an ideal population that would have the same rate of genetic change as the population under consideration. \( N_{eff} \) influences the rate of loss of genetic variation, the efficiency of natural selection and the accumulation of mutations. As a rough guideline, \( N_{eff} \) approximates the number of breeding individuals producing offspring that live to reproductive age.

\( F_{ST} \): standardized index of the distribution of genetic variation between populations on a scale between 0 (identical allele frequencies among populations) and 1 (populations fixed for different alleles).

\( F_{temporal} \): standardized measure of change in allele frequency between two or more samples collected at different times from the same population. \( F_{temporal} \) is commonly used to estimate \( N_{eff} \).

Gametic (linkage) disequilibrium (LD): non-random association of alleles at two or more loci.

Genome typing: simultaneous genotyping of tens or hundreds of loci from across the genome, which ideally includes mapped loci and different classes of loci, such as neutral and candidate adaptive loci.

Mixture: group of \( F_2 \) individuals originating in different populations.

Non-invasive genetic sampling: collecting biological samples, such as hair, scat, feathers, or urine, from an animal without having to physically restrain, capture, or even see the animal.

Open mark-recapture model: an open population model assumes that the population can change through processes such as births, deaths, emigration and immigration during the course of the study, whereas closed population models assume the population does not change in such ways while under investigation.

Retrospective monitoring: monitoring changes from historical conditions by examining DNA extracted from previously archived samples.

**Glossary**

Admixture: composite gene pool of \( F_1, F_{ST}, F_{ST}-F_{ST} \) individuals resulting from hybridization between two or more populations.

Vital rates: Vital rates, such as survival and recruitment, drive population change over time and are of fundamental importance to population ecologists. Traditionally, survival has been difficult to estimate, much less monitor, for rare and elusive species. The ability to use molecular data to track known individuals can either replace or augment traditional efforts to monitor survival [16]. For instance, naturally shed feathers of Eastern imperial eagles *Aquila heliaca* have been used in a microsatellite analysis to identify individuals, monitor population turnover and estimate annual survivorship over a four-year period [17]. The results showed that turnover rates varied annually and that cumulative survival was lower than expected; such information can not be obtained easily using traditional approaches.

We expect that genetic monitoring of vital rates will become prevalent in reintroduction efforts, where post-release genetic monitoring is badly needed, but currently underused. This approach will be most compelling when genotypes of reintroduced animals can be determined before release [18] and the fate of each individual tracked by the presence or absence of its genotype in subsequent, non-invasive samples. These data can also be used for parentage analysis to monitor reproductive output.

**Category Ib: Identifying species and other groups**

Diagnostic genetic markers can be used to identify species or other groups (e.g. genetically differentiated populations, subspecies, or genera) in addition to individuals. The combination of non-invasive genetic sampling and novel species identification tools enables the monitoring of (i) changes in site occupancy or the geographical range of a taxon; (ii) the presence of hybrids; and (iii) the emergence of disease or invasive species (Figure 1). This category differs from similar approaches covered in Category II, where gene frequencies and assignment methods are used to assign individuals statistically to populations or species; here molecular markers are diagnostic for the taxon of interest, and probabilistic analyses based on gene frequencies are not required.

**Site occupancy and geographical range** Changes in the proportion and distribution of sites where a species is observed is commonly monitored to elucidate effects of environmental change [19,20]. These occupancy statistics can be hindered by low probability of detection owing to the elusive nature of some species, and by false positives that occur when visual identification is cryptic or a species leaves signs that are easily confused with signs of other species. Because species identification through DNA is reliable and assays are inexpensive, we expect that the use of non-invasive collection devices will soon form the basis of many large-scale surveys monitoring changes in occupancy and geographical range. Several large-scale, DNA-based surveys have already been established to monitor the distribution of species over time [21,22]. For example, the presence of Canada lynx *Lynx canadensis* has traditionally been monitored using unreliable metrics, such as pelt returns, snowtracks (which have high false positive rate) and incidental sightings. In 1998, a five-year monitoring effort was initiated using non-invasive genetic sampling for lynx presence on ~44 million ha of US national forest land [21,23]. Sampling was conducted using bait stations that attracted lynx to the site and elicited a rubbing response on a hair collection device. Lynx were detected in multiple locations from the hair samples using mtDNA. The success of this work provides a baseline to monitor long-term changes in the geographical range of the lynx and can already be compared with historical records [21].

**Hybridization** Regular monitoring with molecular markers can provide early detection of hybridization, thus giving managers a wider range of options than is available if introgression is already extensive before detection. In recognition of this reality, genetic monitoring of hybridization has been proposed for many species, including Canada lynx–bobcat *Lynx rufus* [24], coyote *Canis latrans*–red wolf *Canis rufus* [25], and spotted owl *Strix occidentalis*–barred...
owl *Strix varia* hybridizations [26]. Genetic monitoring programs have already been initiated for several species; for example, native westslope cutthroat trout *Oncorhynchus clarki lewisi* in the Rocky Mountains are threatened with genomic extinction owing to widespread hybridization with introduced rainbow trout *O. mykiss* [27,28]. Diagnostic molecular assays applied to trout from the Flathead River system of Northwest Montana have demonstrated an increase in the proportion of streams with westslope cutthroat–rainbow trout hybrids between 1984 and 2001 [27].

**Pathogens and parasites** Disease is one of the greatest threats to population persistence. Molecular genetic markers, including DNA barcoding markers [7], have enormous potential to help monitor the presence, prevalence and transmission of pathogenic organisms. Whole genomes and gene sequences are available for increasing numbers of pathogens, making marker development relatively easy. For example, brucellosis and TB, both caused by bacterial pathogens, are among the most problematic diseases in ungulates and some carnivores [29,30]; they cost millions of dollars through vaccinations, ‘test-and-slaughter programs’ and reduced numbers of economically valuable wildlife and livestock. Genomes were recently sequenced for both pathogens (*Brucella abortus* and *Mycobacterium tuberculosis*, respectively), facilitating development of molecular markers to diagnose the presence of different strains and to track their spread [31].

Similarly, genetic monitoring provides new insights into West Nile virus (WNV) epidemics in humans. Over a five-month period, molecular markers were used to detect the presence of WNV-carrying *Culex pipiens* mosquitoes in the Washington DC and Baltimore, MD area of the USA; whether the mosquito was engorged with avian or mammalian blood; and the presence of WNV RNA infection in the mosquito. Temporal patterns in the molecular results, combined with temporal abundance data for mosquitoes and birds, suggests that the increase in human WNV infections during late summer is due to mosquitoes shifting their diet from their preferred avian host to humans [32].

**Category II. Monitoring population genetic parameters**

Monitoring population genetic metrics (Figure 1) can provide insights into demographic and evolutionary processes in natural and captive (Box 2) populations that are difficult or impossible to obtain using traditional methods (e.g. temporal dynamics of a seed bank [33]). This type of monitoring can evaluate population characteristics (e.g. effective population size, *N_e*, or connectivity) before a monitoring program begins, as technical advances provide increasingly reliable DNA recovery from archived material (e.g. museum skins, fish scales or trophy collections). This enables ‘retrospective monitoring’ to assess historical conditions [34,35]. Here, we discuss the most commonly monitored population genetic signals.
Population size  Population genetic monitoring often evaluates changes in abundance inferred through changes in genetic diversity (e.g. expected heterozygosity, $H_e$; allelic diversity, $A$), allele frequencies (e.g. temporal changes in allele frequency, $F_{\text{temporal}}$) or $N_e$ (Table 1). Empirical studies [36,37] and simulations [38,39] both demonstrate that $A$ is more sensitive than $H_e$ for detecting a reduction in population size, because $H_e$ is insensitive to the loss of rare alleles. Consequently, $A$ is often the target of monitoring efforts. Monitoring $F_{\text{temporal}}$ is an even more sensitive indicator of population decline than is loss of $A$ [37], because changes in allele frequency can be substantial without any loss of alleles [38]. For example, studies of brown trout *Salmo trutta* in Demark have monitored

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**Box 1. Monitoring for trend in abundance: Owikeno Lake grizzly bears**

As genetic monitoring is a relatively recent phenomenon, most studies using molecular tags to generate trends in abundance have asked whether the population has changed in size over only one or two time intervals. However, long-term data sets with multiple abundance estimates are generally required to distinguish meaningful changes in abundance from sampling error and environmental variability [64,70]. One of the best examples of estimating trends using molecular markers and Category I monitoring is the study by Boulanger et al. [14] on the influence of local salmon availability on the trend in grizzly bear *Ursus arctos horribilis* abundance in the Owikeno Lake area of British Columbia, Canada.

Traditionally, estimating bear abundance trend was limited by the short sampling seasons, cost, the inability to capture and follow or recapture sufficient numbers of animals, and the risk in conducting such work [71]. Boulanger et al. used DNA obtained from hair snares to mark and track individual bears [13] (Figure I), while simultaneously using traditional methods to monitor salmon availability in multiple parts of their study area. Although using DNA-based CMR to estimate single year abundance is becoming commonplace with bears [13,72], Boulanger et al.’s approach was unique in that they had a five-year data set and could apply an open mark–recapture model to the entire data set [73] to estimate apparent survival, rates of recapture, additions (immigration and births) and population size changes. The authors found that, in all three sampling areas, bears had significant negative population growth in the first two years of the study, followed by positive growth in the next two. The combined experimental design enabled Boulanger et al. to elucidate the underlying mechanisms behind this trend; apparent survival and rates of addition were both related to salmon availability.

This study demonstrates that DNA-based abundance estimates can be computed in a formal modeling approach over time to make inferences regarding trends, the basis for a genetic monitoring approach.

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**Box 2. Genetic monitoring of captive populations**

Although we focus here on monitoring wild populations, artificial propagation is being used increasingly as a conservation and management tool. Many captive programs release individuals into the wild, and informed management requires monitoring of the genetic consequences of these releases on natural populations. Here, we discuss two ways in which such monitoring can assist recovery of wild populations.

**Monitoring changes within captive populations**

It is important to monitor genetic changes within captive populations, as the loss of genetic variation from the use of too few founders or faulty breeding protocols can compromise recovery efforts [74]. Dowling et al. [75] monitored gene frequencies for seven years in samples of repatriated (wild-produced larvae reared in captivity and re-released as juveniles) razorback sucker *Xyrauchen texanus* in Arizona to determine whether the captive-rearing program was transmitting sufficient genetic variation to the endangered wild population. Their results suggest that, to date, the program has been successful in avoiding use of progeny from only a small fraction of wild spawners. By contrast, monitoring of the hatchery population of the endangered Rio Grande silvery minnow *Hybognathus amarus* between 2001 and 2003 has consistently shown heterozygosity levels equal to, but allelic diversity much lower than, that of the wild population [76], a classic sign of a population bottleneck.

**Monitoring recruitment in wild populations**

Genetic monitoring can be an efficient way of determining whether captive individuals are recruited into wild populations [77], and it is the most reliable method to determine whether they are making a reproductive contribution to subsequent generations. Hansen [78] examined long-term impacts of intense stocking on brown trout *Salmo trutta* in two Danish populations by comparing genetic profiles from contemporary samples with those from archived fish scales; one population changed dramatically over time owing to stocking, whereas the other showed little genetic contribution of stocked fish.

Molecular markers can also be used in parentage analysis to assess reproductive success of captive individuals in the wild. Araki et al. [79] monitored a supplementation program for steelhead (*Oncorhynchus mykiss*; Figure I) in Oregon and found that the relative reproductive success of hatchery fish improved when the program phased out the non-native hatchery stock and began using local native fish for brood stock. Monitoring the genetic contribution of captive individuals to natural populations is necessary before one can assess how the captive program affects the $N_e$ of the combined captive–wild system [50,80].

Genetic monitoring of captive propagation is common in fishes, especially salmonids [61], but is increasingly being applied to other taxa. Recent studies include evaluations of the effects of oysters transplanted from Louisiana on local Chesapeake Bay populations; spread of ectomycorrhizal fungi used to enhance forest tree growth after out-planting; diversity in translocated populations of wallabies in Australia; and the success of protocols to introduce an endangered annual plant to restored vernal pools in southern California [81–84].
leading to the moderately high $F_{\text{temporal}}$. The method requires samples from at least three time periods. National Park by analyzing samples from the 1910s, disequilibrium, requiring only one sample for each estimate of $N_e$. As two samples are needed for a single temporal and allele frequency distributions.

Difficult to provide general guidelines for sample sizes because these often depend on numerous factors, such as type of loci (SNPs versus microsatellites), population history, and allele frequency distributions. $F_{\text{temporal}}$ for five time periods from 1944 to 1997 [40]. It was found that $H_e$ and $A$ were high and stable over time, but that $F_{\text{temporal}}$ varied substantially among periods. The authors concluded that high genetic diversity was maintained by gene flow in the face of small, local $N_e$, leading to the moderately high $F_{\text{temporal}}$. $N_e$ monitoring usually is based on change in allele frequencies, and is similar to tests described above using $F_{\text{temporal}}$. However, although the above tests are sensitive to relative changes in population size, they do not quantify it. Estimating $N_e$ enables direct tests for changes in population size. As two samples are needed for a single temporal estimate of $N_e$, monitoring changes in $N_e$ via the temporal method requires samples from at least three time periods. $N_e$ has been estimated for brown bears in Yellowstone National Park by analyzing samples from the 1910s, 1960s and 1990s [41]. $N_e$ estimates were $\approx 85$ for both time periods (1910–1960s and 1960s–1990s), providing no evidence of a recent population decline. Other studies have also used a combination of contemporary and historical samples to obtain multiple temporal estimates of $N_e$. Palm et al. [44] monitored allele frequencies for 20 years in two Swedish brown trout populations and used a modified temporal method to obtain annual estimates of $N_e$. They found consistently small estimates, but no temporal trend in $N_e$.

$N_e$ monitoring can also be based on gametic disequilibrium, requiring only one sample for each $N_e$ estimate. Unbiased gametic disequilibrium $N_e$ estimators have been available only recently [45,46]. The precision and reliability of this method have not yet been thoroughly quantified, and further research is needed. However, the ability to estimate $N_e$ based on a single sample should improve the power to estimate any trends.

Gene flow Effective conservation often depends on the identification of management units and timely information regarding the effects of natural and anthropogenic factors on movement and gene flow between these units [47]. Although several genetic methods yield point estimates of gene flow, a monitoring program that produces a temporal series of samples can provide richer insights [48,49]. Furthermore, monitoring changes in gene flow indices, such as $F_{\text{ST}}$, can detect changes in differentiation among populations [50]. For example, genetic monitoring of leopard frog Rana pipiens populations revealed that genetic structure was stable over 11–15 generations [51]. A study of cod Gadus morhua populations from the Baltic and North Sea using both historical and contemporary samples also found high temporal stability in $F_{\text{ST}}$ over 47 and 89 years, respectively [35]. However, other studies have found the opposite result. Genetic monitoring of Scottish red deer Cervus elaphus demonstrated that fine-scale genetic structure in females declined at a steady rate over a 24-year period [52]. Based on demographic data collected over the same time period, the authors concluded that this decline was due to a combination of increasing population size and decreases in polygyny.

One of the most fruitful uses of genetic monitoring is likely to be the quantification of changes to movement patterns in response to events that disrupt historical patterns of connectivity, such as habitat fragmentation. Genetic monitoring is already a useful tool for evaluating the cumulative effects of habitat fragmentation. An allozyme and microsatellite-based study of California valley oak Quercus lobata pollen movement suggests that there was a decline in the effective number of fathers contributing pollen to the next generation between 1944 and 1999 [53]. The authors propose that this was the result of progressive stand thinning and that it might lead to future reproductive failure owing to genetic isolation. In a study of Pyne’s ground plum Astragalus bifolius, the genetic diversity of allozymes in multiple stratigraphic deposits

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*aSample sizes of individuals and loci that are necessary to achieve high power or precision to detect genetic changes are discussed in, for example, Refs [38,63,67–69]. It is difficult to provide general guidelines for sample sizes because these often depend on numerous factors, such as type of loci (SNPs versus microsatellites), population history and allele frequency distributions.

*bAbbreviations: A, allele richness; Cl, confidence intervals (or support limits when using likelihood based or Bayesian approaches); $F_e$, an index of inbreeding (mating between close relatives); $F_{\text{temporal}}$, temporal variance in allele frequencies; $H_e$, expected heterozygosity; LD, linkage (or gametic) disequilibrium; $N_e$, effective population size. *bBottleneck tests, which also use allele frequency distributions, can help detect a recent population decline, but are not listed here because they have been used for assessments rather than for monitoring as defined above.

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was examined to discern the effects of cedar glade fragmentation [54]. Seeds from the top soil layer had higher levels of differentiation among sites than did the two lower layers, a result consistent with recent habitat loss.

Population mixtures Mixed-stock analysis estimates the proportions of individuals in a mixture that originate from each of two or more genetically differentiated breeding groups. Although mixed stock analysis (and detection of hybrids) can be based on diagnostic assays (Category Ib), it is often necessary to use probabilistic approaches based on allele or genotype frequencies. Genetic tools have been used for many years to assess the composition of mixed-stock fisheries [55,56], but recent laboratory and statistical advances provide increased power for real-time genetic monitoring. Fraser River sockeye salmon Oncorhynchus nerka are the most valuable commercial salmon fishery in British Columbia, but efficient harvest of the summer run is complicated by conservation concerns for late-returning populations. In 2002, a large monitoring program (up to 600 individuals several times a week for two months) based on microsatellites and major histocompatibility (MHC) loci provided stock composition estimates within 9–30 h [57], enabling managers to avoid overharvest of the late run.

Population admixtures also can be quantified and monitored with methods that estimate the degree of mixed ancestry of each individual. However, the power and resolution are generally much lower when the source populations are characterized by allele frequency differences, rather than by fixed differences in diagnostic markers ([56] as in Category Ib).

Caveats to genetic monitoring
Genetic monitoring offers some of the best opportunities to track populations over time and to evaluate when populations reach critical thresholds that demand management action. Category I genetic monitoring has many of the benefits of traditional abundance, distribution and vital rate monitoring, with the added benefit of larger and, in some cases, more representative samples owing to the relative ease of non-invasive genetic sampling. Category II monitoring measures genetic variation, which is the raw material of evolution and which affects the long-term status of a population. However, there are limitations to recognize and cautions to heed when implementing genetic monitoring programs.

First, DNA markers represent an added expense, particularly with non-invasive samples, owing to the need to repeat genetic analyses to reduce genotyping errors [58]. This expense is often offset by reductions in field costs, such as those incurred by the handling of live animals. Second, genotyping errors, if not adequately controlled, impede accurate estimates of many metrics. Multifold overestimates in abundance (Category I) can occur if genotyping error rates are high [59]; by contrast, \( N_e \) estimated from the temporal method (Category II) often will be downwardly biased if genotyping errors are common, as errors will be interpreted as drift. Given the desire for retrospective monitoring using archived samples, which often yield low-quality DNA, genotyping error must be aggressively guarded against. Third, genetic monitoring efforts that rely on non-invasive samples are more vulnerable than are traditional monitoring to fraud (e.g. it is easier to plant a spurious genetic sample than a live individual); conflict has already arisen over planting of hair samples in a non-invasive survey designed to monitor the occurrence of a threatened species [23]. Fourth, sufficient molecular, statistical, and computational tools are not always available to detect the phenomenon of interest with the desired power (Box 3). For instance, in most cases, detection of hybridization is limited to \( F_1 \) and occasionally \( F_2 \) generations, largely owing to the lack of diagnostic markers [60]. Lastly, perhaps the main limitation is that genetic changes associated with population declines or fragmentation are most easily detected following severe disturbances. This limitation can be overcome to some extent by using more sensitive methods.

### Box 3. Designing a genetic monitoring program

Most regular monitoring efforts will yield useful information, even though the most valuable aspects might not be apparent until decades later. To maximize prospects of obtaining useful results, numerous factors should be considered in monitoring program design:

**Identify objectives**
Long-term monitoring programs might begin with rather general objectives that are likely to evolve over time. Shorter-term efforts generally have more focused objectives that should be as specific as possible. For example: (i) estimate abundance or \( N_e \) of a target population with a coefficient of variation \( < 0.5 \); or (ii) have a 95% probability of detecting invasive species (or hybrids or immigrants) in the study area.

Alternatively, the primary objective might be to draw general conclusions about ecological impacts that will be applied to other populations (thus increasing the scope of inference); for example, the evaluation of whether dams alter migration patterns in an aquatic species over time.

**Evaluate potential sampling and analytical methods**
Potential sampling and analytical methods should be evaluated with respect to: (i) relevance to objectives. Can the method provide the right type of information? In general, long-term projects should begin with a relatively simple sampling design to provide flexibility for future changes; (ii) logistics. Is it feasible to collect the necessary data?; (iii) power. Given a feasible sampling regime, can the analytical method deliver the necessary statistical power to accomplish the objective?; and (iv) robustness. Unexpected events will occur; how sensitive are the methods to unforeseen developments and model assumptions?

**Include an experimental design**
Appropriate controls should be established whenever possible; if this is not feasible, changes in key population parameters can still be monitored, but it will be more difficult to establish cause-and-effect relationships. If the objectives involve broad inferences about ecological processes, a more complex experimental design is called for. For example, the traditional BACI (before–after, control–impact) design can be modified to enable more temporal and spatial replication of monitoring efforts [85].

**Manage adaptively**
The program should allow for maximum flexibility to respond to new information and unanticipated factors, including changes in the natural system (e.g. flood, drought or cyclic climate regimes); anthropogenic changes (uncontrolled modifications to system); and new analytical methods. If the program extends for any period of time, it is likely that important new methods will emerge. Can the data and sampling regime accommodate new methods? Can voucher specimens be archived for subsequent verification or to provide material for new (as yet unforeseen) analyses?
and larger samples or more loci, which increases statistical power to detect effects of smaller size.

**Perspectives**

Although genetic monitoring has been discussed for some time [61], many techniques, statistics and models are relatively new. Consequently, limited time series exist to evaluate the true promise of genetic monitoring. Given the demonstrated power and flexibility of genetic monitoring, we believe that applications to answer key ecological, conservation and evolutionary questions will continue to expand rapidly.

Such developments are expected in the number and classes of molecular markers used for monitoring, demonstrated by the recent developments in population genomics and genome typing [4]. For instance, monitoring loci related to a stress event (e.g. MHC), along with many neutral markers for background information, could help detect the onset of selection events such as emerging diseases, anthropogenic landscape change, competition from invasive species, or pollution. Population genomics will also provide increased statistical power to monitor for genetic change. Advances will also occur in the monitoring of gene expression. For example monitoring changes in gene expression using microarrays [62] could signal a novel stressor (e.g. water quality change) or provide an early warning of detrimental stressors and challenges to organisms (e.g. emerging disease).

Other areas where progress is likely is in the development of sampling strategies, experimental designs (Box 3) and statistical approaches. New statistical frameworks such as approximate Bayesian computation, which are less computationally demanding, will be used more widely to estimate population parameters such as migration or $N_e$ [4,6]. Furthermore, a greater emphasis will probably be placed on estimating power to detect effects using the aforementioned metrics at a single point in time (e.g. power to detect substructure [63]) and power to detect trend in these metrics given the variability inherent in natural systems (Box 3; [64]). This will, in turn, force researchers to focus more on sampling strategies and experimental design and turn away from ad hoc study designs.

Finally, for Category II, we expect more monitoring of signals embedded in the genetic data, such as changes in microsatellite allele size and frequency distributions, and more monitoring of changes in individual-based gene flow models from the emerging field of landscape genetics (i.e. the use of assignment, and clustering algorithms to look at contemporary movement [5,65]). These new approaches have considerable promise for studying non-equilibrium systems, but the range of practical applications for which they can provide useful information has not yet been established [56].

**Future research**

We hope that our review will inspire additional research. First, simulation studies are needed to evaluate the relative power of monitoring genetic or traditional population metrics (e.g. to determine when it is more effective to monitor abundance or $N_e$ to detect changes of given magnitudes). Second, cross-fertilization of ideas between researchers studying different taxa is desirable, as we have noticed that monitoring with diagnostic molecular markers (Category Ia) is primarily used with terrestrial vertebrates, whereas monitoring population genetic parameters (Category II) is more common with other taxa. We see no biological imperative for this dichotomy. That cross-fertilization is already beginning to occur is illustrated by the use of non-invasive genetic monitoring to estimate abundance and harvest rates in coral reef fish [66].

Finally, we observe that much has been learned recently by analyzing historical samples that are only available through the foresight of previous collectors. These empirical results, together with the recognition that we cannot anticipate future technological advances, argue cogently for aggressive archiving contemporary samples. We never know what creative new approaches will be developed in the future for monitoring ecological change.

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