which should henceforth be encouraged. Indeed, the US congressional investigation on shadow pricing was spurred largely by coordinated action by patients made aware of pricing discrepancies through research published in a reputable medical journal. Along these lines, we would argue that technological advancement in analytical tools and pricing analytics will be increasingly relevant under the current legal frameworks to establish direct communication and negotiation between payers and drug companies and to induce further transparency in the pricing process—as a major technology company sought to do last year*. Anurag S Rathore** and Faheem Shaheef

Indian Institute of Technology Delhi, Department of Chemical Engineering, Hauz Khas, New Delhi, India.
*e-mail: asrathore@biotechcmz.com

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References
1. Pharmaceutical Research and Manufacturers of America (PhRMA). Prescription medicines: costs in context. Advocacy Cost & Value of Medicine https://www.phrma.org/report-prescription-medicines-costs-in-context (2016).
2. Greenwood, J. U.S. drug costs must be weighed against benefits. Bloomberg News http://www.bloomberg.com/news/articles/2013-12-28/u-s-drug-costs-must-be-weighed-against-benefits (2013).
3. Scannell, J. W., Blanckley, A., Boldon, H. & Warrington, B. Nat. Rev. Drug Discov. 11, 191–200 (2012).
4. Brennan, Z. Congressmen investigate 'shadow pricing' of MS drugs. Regulatory Focus https://www.raps.org/regulatory-focus%E2%84%A2/news-articles/201709/congressmen-investigate-shadow-pricing-of-ms-drugs (2017).
5. Kelley, T. U.S. insulin prices rise as Sanofi, Novo await rivals. Bloomberg News https://www.bloomberg.com/news/articles/2013-08-15/u-s-insulin-prices-rise-as-sanofi-novo-await-rivals (2013).
6. Langreth, R. Hot drugs show sharp price hikes in shadow market. Bloomberg News https://www.bloomberg.com/news/articles/2015-05-06/diabetes-drugs-compete-with-prices-that-rise-in-lockstep (2015).
7. Barrett, P. & Langreth, R. The crazy math behind drug prices. Bloomberg Businessweek https://www.bloomberg.com/news/articles/2017-06-29/the-crazy-math-behind-drug-prices (2017).
8. Persistence Market Research. Biosimilar Market: Pricing Analysis 2006–2017 (Persistence Market Research, New York, 2017).
9. Hartung, D. M., Bourdette, D. N., Ahmed, S. M. & Whitham, R. H. Neurology 84, 2185–2193 (2015).
10. Bennetth, Y. FDA deals blow to Teva’s defense plan with ruling on Bendeka. Bloomberg News http://www.bloomberg.com/news/articles/2016-03-29/nda-deals-blow-to-teva-s-defense-plan-with-ruling-on-bendeka (2016).
11. Serebrov, M. PBMs: The ‘shadow’ players in the drug pricing skirmish? Bio World http://www.biovworld.com/content/pbms-shadow-players-drug-pricing-shadowb (2016).
12. Tufts Center for the Study of Drug Development (CSDD). Cost of Developing a New Drug (Tufts CSMID, Boston, 2014).
13. Loftus, P. U.S. investigates drugmaker contracts with pharmacy-benefit managers. The Wall Street Journal https://www.wsj.com/articles/u-s-investigates-drugmaker-contracts-with-pharmacy-benefit-managers-1462895700 (2016).
14. Wingfield, N., Thomas, K. R. & Abelson, R. Amazon, Berkshire Hathaway and JPMorgan team up to try to disrupt health care. The New York Times https://www.nytimes.com/2018/01/30/technology/amazon-berkshire-hathaway-jpmorgan-health-care.html (2018).

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Federated discovery and sharing of genomic data using Beacons

To the Editor — The Beacon Project (https://github.com/ga4gh-beacon/) is a Global Alliance for Genomics & Health (GA4GH) initiative that enables genomic and clinical data sharing across federated networks. The project is working toward developing regulatory, ethics and security guidance to ensure proportionate safeguards for distribution of data according to the GA4GH-developed “Framework for Responsible Sharing of Genomic and Health-Related Data”. Here we describe the Beacon protocol and how it can be used as a model for the federated discovery and sharing of genomic data.

A Beacon is defined as a web-accessible service that can be queried for information about a specific allele. A user of a Beacon can pose queries of the form “Have you observed this nucleotide (e.g., C) at this genomic location (e.g., position 32,936,732 on chromosome 13)?” to which the Beacon responds with either “yes” or “no.” In this way, a Beacon allows allelic information of interest to be discovered by a remote searcher with no reference to a specific sample or patient, thereby mitigating privacy risks.

In principle, allelic information from any source (or species) can be distributed through a Beacon. For example, a Beacon may serve data from case-level observations, such as genetic variants identified from sequenced samples, or from annotation resources such as variant–disease associations curated from scientific literature. Along with a “yes” response, a Beacon may optionally disclose metadata, including allele frequencies, pathogenicity scores and associated phenotypes, associated with the queried allele. Access to Beacons is secure through institutional systems for authentication and authorization (for example, ELIXIR AAI), allowing hosts to enforce proportionate safeguards for datasets that may be sensitive and consented for use only by trusted individuals and/or for specific purposes.

The Beacon Project is demonstrating the willingness of international organizations to work together to define standards for, and actively engage in, genomic data sharing. Several organizations have ‘lit’ (i.e., implemented) a Beacon, and these have been assembled into a single searchable network. In the years since the project’s inception, over 100 Beacons have been lit by 40 organizations serving over 200 datasets. The datasets served through Beacons are searchable individually or in aggregate—for instance, via the Beacon Network (https://beacon-network.org/), a federated search engine across the world’s Beacons.

Beacons are a general-purpose protocol for genomics data discovery and have been lit by both large and small organizations, as well as by individuals. This has made available datasets collected from large-scale population sequencing efforts (for example, 1000 Genomes), clinical diagnostic settings, in silico predictions (for example, PolyPhen-2), expertly curated or crowd-sourced databases, scientific literature (for example, the Human Genome Mutation Database) and variant curation efforts (for example, ClinVar). The International Cancer Genome Consortium’s Beacon shares case-level somatic variant observations from over 60 cancer subtypes; the PhenomeCentral’s Beacon shares observations from hundreds of clinical cases of undiagnosed and rare genetic diseases; and the BRCA Exchange (https://brcaexchange.org/) Beacon distributes consensus classifications for variants in BRCA1 and BRCA2 cataloged by the ENIGMA Consortium, as well as variants collected from other resources as part of the GA4GH BRCA Exchange (https://brcaexchange.org/). The ELIXIR hub (https://elixir-europe.org/) is also integrating Beacon to connect geographically distributed data centers and unify their data access methodologies. This will enable aggregate sharing of allelic observations between sites, a feature that is not yet available through its services.
With continued adoption, Beacons will produce a large network of globally searchable genomics datasets that have the potential to unlock new genomics-derived discoveries and applications in medicine.

**Beacon protocol**

Many former systems for genomic data sharing have followed a centralized model, wherein data generators deposit information into a single repository, such as the Sequence Read Archive (SRA)\(^2\). This model requires data generators to transfer whole copies of datasets over the internet, which will become inefficient and expensive as the rate of genomic data acquisition increases. An alternative, federated model for data sharing\(^3\) requires organizations to host data independently and to interoperate via an agreed-upon technical language. This model removes the inefficiencies of large data transfers and gives host organizations more control over data privacy, security and representation.

For maximal interoperability, a Beacon is designed to be a communication layer that is compatible with any underlying representation of alleles or their annotations. For example, the GA4GH develops a data representation format for genomic variants and annotations, but in practice these data types may be stored in other formats as well (for example, VCF files or relational databases).

Sharing through Beacon is notably different from sharing fully describe data representations for genomic variants (for example, VCF) or annotations (for example, GFF). The Beacon protocol considers levels of data aggregation and obfuscation that can be added onto raw data representations (such as VCF) to convey useful information without explicitly referring to specific samples or individuals.

With these features in mind, the Beacon protocol was designed to be:

- **Simple:** Beacons can be implemented on top of any underlying variant or variant annotation data store.
- **Federated:** Beacons can be lit and maintained by individual organizations and assembled into a distributed network.
- **General purpose:** Beacons can be used to distribute any allelic dataset, including case-level observations or other annotations.
- **Aggregative:** Beacons provide a boolean answer to whether an allele was observed, possibly aggregated across an entire population, and therefore support deidentification in a way that sharing via VCF files does not.
- **Securable:** Beacon access can be restricted using institutional security protocols, and authorization schemes can be implemented to respect conditions consented to by patients and/or data owners.

The Beacon API (represented as a RESTful web application) provides a technical specification that a Beacon server must implement. The specification is open-source and available online at https://github.com/ga4gh-beacon/specification.

A Beacon has two available functions: the first lists information about the Beacon, including descriptions of the host organization and specific datasets that it serves; the second queries for the existence of information about specific alleles. Alleles are specified with chromosomal coordinates in addition to reference and alternate bases. Much as in their use in VCF, reference and alternative bases can be used together to specify exact matches for single nucleotide variants (SNVs) and small insertions or deletions. A Beacon responds either “yes” or “no” to signal whether the dataset(s) it serves have information about the queried allele. In the affirmative, a Beacon may optionally disclose metadata describing the observations or annotations associated with the queried allele. An example query and response is shown in Supplementary Fig. 1.

**Reference implementation**

To simplify the process of lighting a Beacon, a free, open-source ‘reference implementation’ of the latest specification has been developed.

This implementation can create a public Beacon from a set of VCF files. It may be deployed locally or in a cloud-based environment maintained by a third-party provider (for example, Amazon, Google or Microsoft). Documentation and links to download and run the Beacon reference implementation are available (https://github.com/ga4gh-beacon/). Third-party organizations, such as Cafe Variome, DNastack and the European Genome-phenome Archive (EGA), also support the ability to light Beacons from genetic variation datasets stored in those systems.

**Beacon security design**

In principle, access to Beacons can be secured through any system of authentication or authorization, at the discretion of the host organization. The GA4GH is promoting different levels of data access (open, registered, and controlled) for convenience and for compatibility across its projects. Each so-called ‘access tier’ has distinct visibility and requirements for authorization. For example, ‘open access’ Beacons are accessible to anonymous users of the internet, whereas ‘registered access’ Beacons are accessible to registered users (for example, bona fide researchers and clinicians) who have agreed to a set of conditions of data use\(^1\).

A Beacon may support one or more access tiers to provide progressive disclosure of increasingly sensitive information (for example, patient phenotypes and clinical information) as users pass through more stringent authentication and authorization checks. For example, tiered access makes it possible for organizations to allow anonymous users to discover the existence of an allelic observation, without the Beacon disclosing more information about it until users identify themselves. The ability for organizations to offer minimal data discovery up front can save substantial time and effort in data access applications when data might not contain relevant data points.

Beacon’s ability to reveal different information at specific access tiers affords genomic data stewards options for distributing allelic information, ranging from fully public to private. Access can be controlled using established authentication and authorization protocols (for example, OpenID Connect and OAuth2.0) to enforce proportionate safeguards for datasets that may be sensitive and/or consented for use only by trusted individuals for specific purposes.

**Attribute disclosure attacks and reidentification**

The “yes” response from a Beacon signals the presence of an allele in a dataset comprising possibly many individuals’ genotypes, thereby mitigating risks associated with reidentifying specific individuals. Independent of their technical implementation, Beacon reidentification attempts require prior knowledge of genomic sequence data from the individual (or that of a close relative); they are arguably preceded by more harmful compromises to privacy. However, reidentification can pose additional risks if sensitive attributes about the individual can be inferred from Beacons (for example, HIV status or mental health condition). Such attacks have been characterized as “attribute disclosure attacks using DNA” (ADAD)\(^3\).

Querying a Beacon for many variants known to exist in a person's genome could lead to confirmation of that person’s inclusion in a given database, potentially revealing sensitive information about that individual. The ability to reidentify individuals has been examined previously\(^1\) and recently in the context of Beacons\(^4\).

The power to reidentify an individual
whose genotypes are reflected through a Beacon depends on the number of individuals whose data is served, the allele frequency distribution of the pool, the scope of allowed queries (for example, exome versus genome), the type of DNA source (for example, normal tissue versus cancer sample) and the number of times a Beacon is queried. Models for population allele frequencies can be leveraged to reduce the number of queries required in such an attempt, but reidentification is still possible without using allele frequencies if a Beacon can be queried a large number of (for example, 10,000) times.

**Risk mitigation schemes**

User agreements, data use policies and technical enforcement of usage quotas can be established to limit the possibility of reidentification and ADAD through Beacons. Organizations are advised to specify terms of use that explicitly prohibit reidentification attempts through the service. When the risk of ADAD is considered too high for data to be distributed publicly, data stewards are encouraged to implement secured access. Compared with public-access tiers, secured-access tiers (either registered or controlled) impose extra social and/or legal disincentives that can help prevent service misuse.

Beacon operators may further specify consent-based data use conditions from a structured set of Consent Codes to impose restrictions indicated by consent of research participants. These Consent Codes, which are general purpose and can be used by genomics data stewards, were designed with the purpose of supporting maximum data use and integration while respecting consent permissions. The current set of Consent Codes is provided in Supplementary Table 1.

The ethical, legal and social status of health-related data that are typically considered sensitive in international policy and laws is being examined to provide guidance in aggregating Beacons and in implementing tiered protection of Beacon attributes based on sensitivity. This guidance aims to enable consistent and proportionate provision of data protection for data that are considered more sensitive by individuals and society. Data stewards should consider the sensitivity of attributes used in describing their Beacons, as well as those in the data itself.

Technical provisions can also be used to reduce the statistical power of reidentification attempts. Individual Beacons can be combined to form a single, aggregate Beacon, and direct access to participating Beacons can be blocked. Aggregate Beacons contain more data points than any of the individual Beacons while obscuring the origin of the data. As an example, a publicly accessible Beacon named Conglomerate has been lit as an aggregate of multiple independent Beacons.

An information budgeting approach can also be used to thwart reidentification attempts, which rely on accumulating evidence from many queries for alleles carried by a specific individual. The power to reidentify an individual using this technique varies inversely with the frequency of the alleles being queried (i.e., very rare alleles are more revealing than common alleles). By metering the cumulative information disclosure for individuals, Beacons can be configured to restrict access before reidentification is possible within a desired level of statistical confidence.

Beacon is a general-purpose protocol for genomics data discovery, and as such can be used to distribute allelic information from various origins, including observational studies from patients with known (for example, the International Cancer Genome Consortium) or unknown (e.g., PhenomeCentral) diseases, population studies (for example, 1000 Genomes), in silico predictions (for example, PolyPhen-2), expertly curated or crowd-sourced databases (for example, BRCA Exchange and ClinVar), and scientific literature (for example, the Human Genome Mutation Database). Additional Beacon implementations are ongoing in Europe, mainly through the ELIXIR Beacon project. The deployment of Beacons for select use cases is described below.

**Matchmaking**

A major obstacle to discovering the causes of rare diseases is sample size. A single affected family can be enough to identify one or more compelling candidate variants, but pinpointing causal genetic variants frequently requires examining unrelated cases with a variant in the same gene and similar phenotypic presentations. Recently, patient matchmaking has been formalized through efforts such as the Matchmaker Exchange (MME), in which users who contribute a case to a database within the federated network can find similar cases in other databases within the network.

MME is a secured-access system, requiring that only authorized databases and users can contribute and exchange patient profiles for matching. However, this inherently limits the discoverability of
the data, which may dissuade some users having candidate genes or variants they want to match. In addition to implementing the MME API19 for patient matchmaking, several organizations within the MME have lit Beacons to serve aggregate views of their clinical datasets more publicly. This allows clinicians with candidate variants to quickly search for existing matches within the MME.

Sequencing initiatives and archives
Large-scale sequencing initiatives, such as the 100,000 Genomes Project17 conducted by Genomics England and the Precision Medicine Initiative18, promise to generate vast volumes of genotypic and associated health information. Data from these projects, once shared, help researchers make inferences on the genetic determinants of disease by way of comparative analysis and association studies.

The 1000 Genomes Project16, NHLBI Grand Opportunity Exome Sequence Project (https://esp.gs.washington.edu/drupal/), and Exome Aggregation Consortium19 are exemplar large-scale initiatives that have shared genotypes from diverse populations through Beacons. As the number and scale of population sequencing efforts expand, a more accurate depiction of global sequence diversity will be available in aggregate through Beacons and the Beacon Network.

In addition, many of the largest genomic archives, such as dbGaP20, the European Genome-phenome Archive (https://www.ebi.ac.uk/ega/home) and the European Variation Archive (http://www.ebi.ac.uk/eva), have provided access to variation data through Beacons for some or all of their datasets. These Beacons collectively provide widespread discoverability across a large amount of data. Many of these resources are continually growing with new submissions and thus provide added value for data depositors by simplifying data distribution and unifying their consumption.

Beacon Network
Beacon represents a simple protocol that, like internet protocols such as HTTP, describes a method for data discovery and exchange between distributed, collaborative systems. Toward developing an ‘internet for genomics’, it is useful to establish a network of protocol adopters and an efficient mechanism for searching across it.

The Beacon Network is a directory and search engine for Beacons. Although individual Beacons answer the question “Have you observed this allele?”, the Beacon Network answers the question “Who has observed this allele?”. The Beacon Network serves as a powerful, convenient and real-time genomic data distribution channel through which users can discover the existence of alleles of interest and be directed to host organizations who have observed them. A schematic of the Beacon Network as a global federated network for genomic information discovery is shown in Fig. 1.

The Beacon Network is accessible either through its website or programmatically through an API, and enables fast, simultaneous search of hundreds of datasets from hundreds of thousands of individuals already served through Beacons worldwide. Beacons can be freely registered to the Beacon Network and can be searched independently or in aggregate with other connected Beacons. The Beacon Network has received over 1.5 million queries in the three years since its launch. The value of datasets connected to the Beacon Network increases as more Beacons join, particularly for comparative applications like rare disease and donor matching.

Conclusions and perspectives
The first version of the Beacon Project has validated the feasibility of a globally federated system for genomic data sharing. The conceptual and technical simplicity of the discovery question, “Have you observed this allele?”, enabled rapid and widespread adoption, and this has served to provide practical feedback for the GA4GH to continue to advance its best practices by holistically addressing regulatory, security and technical aspects of global genomics data sharing. However, the narrow focus of the initial Beacon question limits its utility to support other closely related use cases, and successive iterations of the protocol are planned to enable coverage of these.

Future extensions to the Beacon protocol may include the following:

- Support for discovering complex genomic alterations, including copy number variations (CNVs) and somatic copy number alterations (CNAs), which are major contributors to both inter-individual variation and disease susceptibility and prominent features of the oncogenic mutation landscape;
- Integration of non-genomics data in queries, including the ability to discover similar cases on the basis of associated metadata;
- Support for quantitative attributes in responses (for example, allele frequencies) to facilitate statistical analyses that combine information disclosed through multiple Beacons;
- Handoff to services by which users may access additional information about a queried variant.

The development of data-rich extensions to the Beacon protocol will leverage the expertise of GA4GH members and stakeholders to iteratively design and evaluate the technical, privacy and security considerations in evolving Beacons to enable unprecedented access to genomics and clinical datasets through a global, federated ecosystem.

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Marc Fiume*, Miroslav Cupak1, Stephen Keenan2,3, Jordi Rambla4, Sabela de la Torre5, Stephanie O. M. Dyke5, Anthony J. Brookes6,7, Knox Carey8, David Lloyd9, Peter Goodhand10, Maximilian Haueussler11, Michael Baudis12,13, Heinz Stockinger12, Lena Dolman2,9, Ilkka Lappalainen11,12, Juha Törmönen13, Mikael Linden14, J. Dylan Spalding15, Salfi Ur-Rehman16, Angela Page11,16, Paul Flicek17,18, Stephen Sherry19, David Haussler8 and Serena Scollen20

DNAnexus, Toronto, Ontario, Canada. 2Global Alliance for Genomics and Health, Toronto, Ontario, Canada. 3European Molecular Biology Laboratory, European Bioinformatics Institute, Wellcome Genome Campus, Hinxton, Cambridge, UK. 4Centre de Regulació Genòmica, Barcelona, Spain. 5Centre of Genomics and Policy, Department of Human Genetics, McGill University, Montreal, Quebec, Canada. 6Department of Genetics, University of Leicester, Leicester, UK. 7Genecloud, Sunnyvale, CA, USA. 8ELIXIR Hub, Wellcome Genome Campus, Hinxton, Cambridge, UK. 9Ontario Institute for Cancer Research, Toronto, Ontario, Canada. 10Genomics Institute, University of California at Santa Cruz, Santa Cruz, CA, USA. 11Department of Molecular Life Sciences, University of Zurich, Zurich, Switzerland. 12SIB Swiss Institute of Bioinformatics, Lausanne, Switzerland. 13CSC – IT Center for Science Ltd, Espoo, Finland. 14Broad Institute of MIT and Harvard, Cambridge, MA, USA. 15Wellcome Trust Sanger Institute, Wellcome Genome Campus, Hinxton, Cambridge, UK. 16National Center for Biotechnology Information, US National Library of Medicine, Bethesda, MD, USA. *e-mail: marc@dnastack.com

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References
1. Global Alliance for Genomics and Health. Science 352, 1278–1280 (2016).
2. Knoppers, B. M. Nat. Genet. 46, 88–89 (2014).
3. 1000 Genomes Project Consortium. et al. Nature 526, 68–74 (2015).
4. Adzhubei, I. A. et al. Nat. Methods 7, 248–249 (2010).
5. Stenson, P. D. et al. Hum. Genet. 133, 1–9 (2014).
6. Landrum, M. J. et al. Nucleic Acids Res. 42, D980–D985 (2014).
7. International Cancer Genome Consortium. et al. Nature 464, 993–998 (2010).
8. Buske, O. J. et al. Hum. Mutat. 36, 931–940 (2015).
9. Spuré, A. B. et al. Hum. Mutat. 33, 2–7 (2012).
CRISPResso2 provides accurate and rapid genome editing sequence analysis

To the Editor — The field of genome editing is advancing rapidly, most recently exemplified by the advent of base editors that enable changing single nucleotides in a predictable manner. For the validation exemplified by the advent of base editors, we present a substantially updated version of our CRISPResso tool to facilitate the genome editing sequence analysis.

CRISPResso2 introduces five key innovations: first, comprehensive analysis of sequencing data from base editors; second, a batch mode for analyzing and comparing multiple editing experiments; third, allele-specific quantification of heterozygous or polymorphic references; fourth, a biologically informed alignment algorithm; and fifth, ultrafast processing time. We discuss each of these in turn below.

Our updated software allows users to readily quantify and visualize amplicon sequencing data from base-editing experiments. It takes as input raw FASTQ sequencing files and outputs reports describing frequencies and efficiencies of base editing activity, plots showing base substitutions across the entire amplicon region (Fig. 1a), and nucleotide substitution frequencies for a region specified by the user (Fig. 1b). Users can also specify the nucleotide substitution (for example, C→T or A→G) that is relevant for the base editor used, and the software produces publication-quality plots for nucleotides of interest with heat maps showing conversion efficiency.

We also improved processing time and memory usage of CRISPResso2 to enable users to analyze, visualize and compare results from hundreds of genome editing experiments using batch functionality. This is particularly useful when many input FASTQ files must be aligned to the same reference genome. In such cases, it is reasonable to assume that indels aligning individual reads to each allele are not confounded with genome editing.

CRISPResso2 generates intuitive plots to show the nucleotide frequencies and indel rates at each position in each sample. This allows users to easily visualize the results and extent of editing in their experiments for different enzymes (Fig. 1c).

In cases where the genome editing target contains more than one allele (for example, when heterozygous single nucleotide polymorphisms (SNPs) are present), genome editing on each allele must be quantified separately, even though reads from both alleles are amplified and mixed in the same input FASTQ file. Current strategies are not capable of analyzing multiple reference alleles and may lead to incorrect quantification. CRISPResso2 enables allele-specific quantification by aligning individual reads to each allelic variant and assigning each read to the most closely aligned allele. Downstream processing is performed separately for each allele so that insertions, deletions or substitutions that distinguish each allele are not confounded with genome editing.

To demonstrate the utility of our approach, we reanalyzed amplicon sequencing data from a mouse with a heterozygous SNP at the Rho gene in which an engineered SaCas9-KKH nuclease was directed to the P23H mutant allele, CRISPResso2 deconvoluted reads, quantified insertions and deletions from each allele, and produced intuitive visualizations of experimental outcomes (Fig. 1d).

Existing amplicon sequencing analysis toolkits ignore the biological understanding of genome editing and instead optimize the alignment on the basis of sequence identity only. However, this can lead to incorrect quantification of indel events, especially in sequences with short repetitive subsequences where the location of indels may be ambiguous as a result of multiple alignments with the same best score. In such cases, it is reasonable to assume that indels should overlap with the predicted nuclease cleavage site. Our improved alignment algorithm extends the Needleman–Wunsch algorithm with a mechanism to incentivize the assignment of insertions or deletions to

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M.F., S. Scollen, G.S., S.V.K., D.L., P.G., S. Sherry, M.B., I.L. and D.H. provided project leadership and management; M.C., J.R., S.d.I.T., J.T., K.C., A.I.B., M.H., M.B., H.S., M.L., J.D.S. and S.U.R. designed and developed software; S.O.M.D. developed ethics and policy research; M.F. and M.C. designed and developed the Beacon Network; P.F. provided security review; M.F. and I.D. wrote the manuscript with contributions from all other authors.

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The authors declare no competing interests.

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References
10. Leinonen, R., Sugawara, H. & Shumway, M. & International Nucleotide Sequence Database Collaboration. Nucleic Acids Res. 39, D19-D21 (2011).
11. Dyke, S. O. M. et al. Eur. J. Hum. Genet. 24, 1676–1680 (2016).
12. Erdlich, Y. & Narayanan, A. Nat. Rev. Genet. 15, 409–421 (2014).
13. Homer, N. et al. PLoS Genet. 4, e1000167 (2008).
14. Shringarpure, S. S. & Bustamante, C. D. Nat. Am. J. Hum. Genet. 97, 631–646 (2015).
15. Dyke, S. O. M. et al. PLoS Genet. 12, e1005772 (2016).
16. Dyke, S. O. M., Dove, E. S. & Knoppers, B. M. Genomic Med. 1, 16024 (2016).
17. Raisaro, J. L. et al. J. Am. Med. Inform. Assoc. 24, 799–805 (2017).
18. Philippakis, A. A. et al. Hum. Mutat. 36, 913–921 (2015).
19. Buske, O. J. et al. Hum. Mutat. 36, 922–927 (2015).
20. Peplow, M. & Med J. 353, i757 (2016).
21. Ashley, E. A. J. Am. Med. Assoc. 313, 2119–2120 (2015).
22. Mallman, M.D. et al. Nat. Genet. 39, 1181–1186 (2007).