**Methods for Optimizing CRISPR-Cas9 Genome Editing Specificity**

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Advances in the development of delivery, repair, and specificity strategies for the CRISPR-Cas9 genome engineering toolbox are helping researchers understand gene function with unprecedented precision and sensitivity. CRISPR-Cas9 also holds enormous therapeutic potential for the treatment of genetic disorders by directly correcting disease-causing mutations. Although the Cas9 protein has been shown to bind and cleave DNA at off-target sites, the field of Cas9 specificity is rapidly progressing, with marked improvements in guide RNA selection, protein and guide engineering, novel enzymes, and off-target detection methods. We review important challenges and breakthroughs in the field as a comprehensive practical guide to interested users of genome editing technologies, highlighting key tools and strategies for optimizing specificity. The genome editing community should now strive to standardize such methods for measuring and reporting off-target activity, while keeping in mind that the goal for specificity should be continued improvement and vigilance.

The clustered regularly interspaced short palindromic repeats (CRISPR)-Cas9 genome engineering technology has been established as a powerful molecular tool for numerous areas of biological study in which it is useful to target or modify specific DNA sequences (Hsu et al., 2014; Sander and Joung, 2014; Cox et al., 2015; Wright et al., 2016). Originally derived from bacterial adaptive immune systems known as CRISPR-Cas (Mojica et al., 2000, 2005; Jansen et al., 2002; Pourcel et al., 2005; Barrangou et al., 2007), the Cas9 nuclease can localize and cleave a target DNA via a guide RNA (Brouns et al., 2008; Deltcheva et al., 2011; Gasius et al., 2012; Jinek et al., 2012). The Cas9 nuclease has been engineered to edit target DNA at desired loci in eukaryotic cells (Cong et al., 2013; Fu et al., 2013; Hsu et al., 2013) like its genome editing predecessors zinc finger nucleases (Bibikova et al., 2001; Urnov et al., 2005; Miller et al., 2007) and transcriptional activator-like effector nucleases (Boch et al., 2009; Moscou and Bogdanove, 2009; Christian et al., 2010; Miller et al., 2011). Efforts to measure, understand, and improve Cas9 specificity have accordingly been of great interest in the field. While prior review articles should be referred to for foundational background and insight into Cas9 genome editing development (Makarova and Koonin, 2015) and applications (Mali et al., 2013b; Hsu et al., 2014; Sander and Joung, 2014; Porteus, 2016; Wright et al., 2016), in this article we focus on the sub-field of Cas9 specificity by comparing relevant methods for off-target improvement and detection, highlighting useful experimental and computational tools, and emphasizing further research directions for continued improvement.

Recently, several improvements have been devised for the wild-type Cas9 enzyme derived from Streptococcus pyogenes, currently the nuclease most widely used for genome editing applications. Initially, bioinformatic solutions for guide RNA selection recommended guide sequences that recognized cognate genomic DNA targets with minimal sequence homology across that given genome. Although such approaches proved reasonably effective, subsequent systematic profiles of Cas9 off-target activity both in vitro (Pattanayak et al., 2013) and in vivo (Fu et al., 2013; Hsu et al., 2013; Mali et al., 2013a) led to the first set of data-driven guidelines for target site selection. Given a particular locus and species of interest, computational models and associated web tools typically rank guide RNAs by predicted specificity and suggest likely off-target sites that can be checked for undesired mutagenesis via targeted sequencing or enzymatic assays (see Biased Off-Target Detection). However, the appropriate number of potential off-target sites to experimentally assay remains unclear, as the accuracy of in silico prediction can vary. Thus, improving methods of off-target detection and quantification has become a focus of a number of research groups in the CRISPR-Cas9 field.

Several protocols for the unbiased detection of genome-wide Cas9 off-target activity in cells have been recently described (Frock et al., 2015; Kim et al., 2015; Ran et al., 2015; Tsai et al., 2015; Wang et al., 2015) and are important complements to biased off-target detection methods based on in silico prediction and targeted sequencing (Fu et al., 2013; Hsu et al., 2013; Mali et al., 2013a; Pattanayak et al., 2013) (see Unbiased Genome-wide Off-Target Detection). The adoption of these protocols marks an important departure from methods for measuring an inherently biased subset of potential off-target sites. In addition to these protocols, the computational tools used for guide RNA design and off-target prediction are expanding to account for a deepening understanding of specificity-governing parameters in the cell, including detailed features of the off-target sequence and the Cas9 orthologs (see Computational Guide RNA Selection and Predictive Models).
Furthermore, multiple methods have been developed to improve specificity by using natural or engineered orthologous Cas9s (see Cas9 Orthologs and New CRISPR Proteins), engineering the Cas9 protein or guide RNA (see Protein Engineering and Bespoke PAMs and Modified Guides), or modulating the kinetics and regulation of the CRISPR components in the cell (see Kinetics and Regulation). Meanwhile, use of nuclease-dead Cas9, or dCas9, for targeted binding applications such as transcriptional regulation have particular considerations given the distinctions between the Cas9 target search and nucleolytic mechanisms (see dCas9 Specificity). CRISPR specificity must also be considered in a broader sense to include cell type and temporal and spatial specificity. Last, the standardization of off-target analysis methods and data reporting in research publications would benefit the field by enabling comparison across studies and better algorithms for guide RNA design (see Broader Implications of Specificity and The Need for Standardization).

**Biased Off-Target Detection**

To date, the predominant approach for identifying Cas9 nuclease off-target activity has been to (1) computationally predict likely off-target sites on the basis of sequence homology and then (2) assess any potential editing activity by enzymatic assays on the basis of mismatch-sensitivity endonucleases (Guschin et al., 2010; Ran et al., 2013b), Sanger sequencing, or targeted deep sequencing (see The Need for Standardization). These a priori predictions are critical, as ~98% of Streptococcus pyogenes Cas9 (SpCas9) guide RNAs in human exons and promoters have at least one off-target site with three or fewer mismatches (Bolukbasi et al., 2016), and foundational Cas9 specificity studies collectively demonstrated that off-target sites with three or fewer mismatches are significantly more likely to be cleaved than more dissimilar sites (Fu et al., 2013; Hsu et al., 2013; Mali et al., 2013a; Pattanayak et al., 2013). Selecting the most unique target site possible is thus a valuable strategy for improving specificity, because many gene editing applications have several possible guide RNAs capable of accomplishing the same experimental outcome (i.e., knocking out a gene by introducing indels in an exon of interest).

Sequence uniqueness is thus an important initial filter for selecting guide RNAs with minimal potential off-target sites with high sequence homology. However, as the number of mismatches considered relative to the on-target site increases, the total number of potential off-target sites dramatically increases as well. For example, a benchmark guide RNA designed to target the EMX1 locus (Ran et al., 2015), with no “off-by-one” mismatched off-target sites in the hg38 reference genome, has 10 “off-by-two” sites, 69 “off-by-three” sites, and 27,480 “off-by-six” sites.

As a result, even targeted sequencing, likely the most scalable of the aforementioned approaches, is biased toward a small number of sites that can reasonably be assayed. However, Cas9-mediated cleavage has been reported at off-target sites with as many as six mismatches to the guide sequence (Tsai et al., 2015), demonstrating the risk of missing detectable events by setting low “off-by” thresholds for the sake of practical, biased off-target screening. Furthermore, most computational off-target prediction tools do not adequately consider off-target sites with gaps, bulges, or alternative protospacer adjacent motif (PAM) sequences (sequence motifs recognized by Cas9 directly 3' of the matching guide sequence in the target DNA; typically 5'-NGG for SpCas9 but also 5'-NAG with lower efficiency) (Table 1).

Looking forward, assays relying on next-generation sequencing of off-target sites may become the standard because of their sensitivity and accuracy, as current methodologies have a reported lower limit of detection of approximately 0.1%, or 1/1,000 (Kleinstiver et al., 2016a). It is suggested that the more accessible Surveyor assay should be used only when quick, first-pass cleavage efficiency measurements are needed. The Surveyor assay is generally quantified by analyzing the band intensity of gel electrophoresis images and is thought to have a limit of detection of ~5% indels (Fu et al., 2013), making it unsuitable for most off-target detection scenarios. Finally, targeted sequencing is perhaps best applied to verify activity at off-target sites discovered by unbiased, genome-wide methods (Kim et al., 2015; Ran et al., 2015; Tsai et al., 2015; Kleinstiver et al., 2016a; Slaymaker et al., 2016).

**Unbiased Genome-wide Off-Target Detection**

To address limitations associated with in silico predictions of potential off-target sites, several unbiased methods have been recently reported for measuring Cas9 off-target activity across the genome (Table 1). We and others have recently reviewed and compared these deep sequencing-based protocols (Gabriel et al., 2015; Gori et al., 2015; Koo et al., 2015; Bolukbasi et al., 2016), including direct in situ breaks labeling, enrichment on streptavidin, and next-generation sequencing (BESS) (Ran et al., 2015), high-throughput, genome-wide translocation sequencing (HTGTS) (Frock et al., 2015), genome-wide, unbiased identification of double-strand breaks (DSBs) evaluated by sequencing (GUIDE-seq) (Tsai et al., 2015), integration-deficient lentiviral vector (IDLV) capture (Wang et al., 2015), and in vitro nuclease-digested whole-genome sequencing (Digenome-seq) (Kim et al., 2015). These unbiased methods collectively demonstrate that sequence homology alone is not fully predictive of off-target sites. Further, although the PAM tends to be the most stringent predicate of target recognition fidelity relative to the guide sequence, off-target mutagenesis was also observed at non-canonical PAMs, confirming that biased methods may miss important Cas9 off-target sites (Frock et al., 2015; Kim et al., 2015; Ran et al., 2015; Tsai et al., 2015).

However, depending on the approach used for enriching Cas9-induced DNA cleavage, unbiased methods tend to be less sensitive and have a lower throughput than biased targeted sequencing, in addition to typically requiring higher sequencing coverage and much more complex protocols. As a result, it is currently difficult to generate off-target data at a scale required to adequately train a predictive model for specificity. Regardless, these approaches remain valuable for researchers interested in validating individual guide RNA sequences in their experimental systems.

GUIDE-seq and Digenome-seq have reported sensitivity to off-target events as low as 0.1%, or 1 in 1,000 events, whereas IDLV capture reported a sensitivity of 1%. The BLESS and HTGTS studies did not explicitly report summary sensitivity metrics. It should be noted that all of these sensitivity rates vary...
depending on the number of cells assayed in the experiment, the 
depth sequencing depth, and the bioinformatics methods used to 
differentiate true Cas9-induced cleavage events from back-
ground noise. As a result, differences in reported detection 
sensitivity between competing protocols do not necessarily 
reflect inherent limitations of the off-target detection method 
but also a need for technical optimization. 

Beyond sensitivity, interested users should also compare 
the protocols’ applicability in their experimental context (Table 
1). For example, GUIDE-seq requires optimized delivery of a 
defined, chemically modified, double-stranded oligonucleotide 
(dsODN), making it potentially toxic to certain primary cell 
types and difficult to apply in vivo on tissue samples. How-
ever, the dsODN-specific PCR enriches for DSBs and thus 
should enable appropriate sequencing coverage with one 
sequencing run of reasonable depth. Although BLESS can 
be applied to both ex vivo and in vivo samples after Cas9 
editing alone, it requires the biochemical ligation of adaptor 
sequences to free, unrepaired DSB ends. As a result, its sen-
sitivity will be affected by the time point after Cas9 editing and 
DNA repair kinetics. Digenome-seq, on the other hand, can 
immediately be applied in any Cas9-editable cell type, yet it 
is an in vitro method that removes genomic structural context. 
It originally required expensive whole-genome sequencing 
(WGS) at high coverage for four sets of genomic DNA per 
assay. An improved version of Digenome-seq lowers the 
cost by multiplexing guide RNAs and reducing the WGS 
requirement from four genomes to one (Kim et al., 2016b). 
This improved protocol was run on ten guides that had 
been analyzed by GUIDE-seq, and the results were compared 
and validated with targeted sequencing of transfected cells. 
Although Digenome-seq generally discovered ~70 more off-
target sites per guide initially, the number of validated off-
target sites was similar for both methods. 

Importantly, as these unbiased methods require manipulation 
of the genome, they will each introduce different and subtle 
artifacts to the set of identified off-target sites and likely will 
not report the exact same sites with identical frequencies. The 
techniques have been compared for some “benchmark” guide 
RNAs, but a comprehensive head-to-head comparison is lacking 
at this time and may be needed (Gori et al., 2015). The improved 
Digenome-seq protocol can be rapidly transferred to new 
laboratories, as it consists of in vitro Cas9 cleavage and WGS. 
BLESS and GUIDE-seq, as currently configured, may be the 
most accessible protocols, as they enrich for DSBs and require 
fewer sequencing reads per sample. Future generations of these 
methods should collectively look to improve accuracy and 
throughput as well as reduce the input cell number required. 

In sum, these biased and unbiased methods can and perhaps 
should be applied in a complementary manner. For example, off-
target sites can be discovered in vitro (e.g., Digenome-seq), in 
easily cultured cell types (e.g., HTGTS, GUIDE-seq), or even 
in vivo (e.g., BLESS) prior to validation in vivo or in cell lines of in-
terest via targeted sequencing. Although combinations of biased 
and unbiased methods have been applied to a small set of guide 
RNAs, larger data sets will be required to determine how well 
in vitro off-target activity corresponds with Cas9 targeting in vivo 
and across different cell types. These types of data sets will, in 
turn, drive continued improvement of bioinformatics tools for 
target site selection. 

Computational Guide RNA Selection and Predictive 
Models 

Many such online tools and applications have been developed 
for the in silico design of guide RNAs and prediction of their 
off-target sites (Table 2). Some, such as Cas-OFFinder, simply 
identify off-target sites via sequence similarity and rank them 
on the basis of relative orthogonality, while others use a speci-
ficity score calculation we have previously described (Hsu 
et al., 2013). This formula scores guides on the basis of a 
weighted sum of the mismatches in their off-target sites, with 
weights experimentally determined to reflect the effect of the 
mismatch position on cleavage efficiency. 

Other studies have similarly established that mismatches at 
the 5’ end of the spacer are more tolerated than PAM-proximal 
mismatches (Cong et al., 2013; Fu et al., 2013; Pattanayak 
et al., 2013). This “seed sequence” effect has been corroborated 
by biochemical experiments demonstrating that Cas9 unwinds 
and binds DNA in a directional manner moving away, or 3’ to 
5’, relative to the PAM (Stemberg et al., 2014), and that proper 
matching of the 5’ end of the guide sequence is important for 
proper nuclease domain positioning and successful Cas9 cleav-
age (Jiang et al., 2015). Beyond weighting for PAM-proximal mis-
matches, the specificity score also reflects the finding that 
consecutive mismatches attenuate Cas9 activity more than an 
equivalent number of interspaced mismatches. Importantly, as 
these target site selection guidelines were derived from studies 
on SpCas9, their general applicability to less commonly used 
Cas9 orthologs remains unclear. 

Overall, these bioinformatic tools generally perform well in 
ranking guide RNAs such that, for a given locus, users can easily 
select guide sequences that are optimized for high genomic 
specificity. However, as our mechanistic and empirical under-
standing of Cas9 activity has progressed, several caveats as 
well as avenues for the improvement of in silico prediction 
have emerged. 

For example, there is evidence suggesting off-target sites with 
small insertions or deletions in the targeted sequence or alterna-
tive PAMs can indeed be cleaved by Cas9, albeit with low fre-
quency (Hsu et al., 2013; Jiang et al., 2013; Kleinstiver et al., 
2015b; Ran et al., 2015). Cas9 specificity is similarly known to 
vary because of other variables such as cell type, species, deliv-
ery modality, and dosage (see Kinetics and Regulation). Although 
combining existing Cas9 binding and cleavage data sets with 
whole-genome chromatin state information has been suggested 
to provide more accurate off-target prediction (Singh et al., 2015), 
generally such data have been collected for a limited number of 
guides. It is difficult to integrate all of these observations in a sys-
tematic, data-driven way, so these parameters are not taken into 
account by the majority of available off-target prediction tools. 

To realize the full potential of in silico prediction, comprehen-
sive data sets addressing such limitations must be generated. 
Large-scale guide RNA efficiency data sets have been used to 
create increasingly accurate computational models that predict 
the knockout efficiency of new guide RNAs (Doench et al., 
2014; Chari et al., 2015; Xu et al., 2015). Most recently, a pooled
Table 1. Improvements to CRISPR-Cas9 Specificity

<table>
<thead>
<tr>
<th>Improvement</th>
<th>Description</th>
<th>Advantage</th>
<th>Disadvantage</th>
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<tbody>
<tr>
<td>Measurement</td>
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<tr>
<td>Targeted deep sequencing (Ran et al., 2013b)</td>
<td>targeted amplicon NGS of putative or known off-target sites, followed by computational analysis to quantify the proportion of reads with indels near the PAM site</td>
<td>more quantitative, sensitive, and scalable than alternative assays such as Surveyor</td>
<td>biased toward a subset of off-target sites that can reasonably be assessed; best used as complement to unbiased, genome-wide off-target detection methods</td>
</tr>
<tr>
<td>GUIDE-seq (Tsai et al., 2015)</td>
<td>captures DSBs with a dsODN, which is then used as a priming site for sequencing</td>
<td>straightforward wet-lab protocol with computational pipelines available online for data processing</td>
<td>requires efficient delivery of the dsODN, which may be toxic to some cell types at some doses, and has not been demonstrated for in vivo models</td>
</tr>
<tr>
<td>BLESS (Ran et al., 2015; Slaymaker et al., 2016)</td>
<td>biochemical ligation of NGS sequencing adapters to exposed gDNA ends; computational filtering separates Cas9-mediated DSBs (aligned near PAM) from naturally occurring DSBs (more randomly distributed)</td>
<td>no exogenous bait is introduced to cells; can be applied to tissue samples from in vivo models</td>
<td>sensitive to time of cell fixation; current configuration requires relatively large number of cells</td>
</tr>
<tr>
<td>Digenome-seq (Kim et al., 2015; Kim et al., 2016b)</td>
<td>cell-free gDNA samples are digested in vitro by Cas9 RNP with multiplexed guide RNAs before WGS; cut sites are identified on the basis of read alignment</td>
<td>can be applied to any cell type as digestion performed on extracted gDNA; more sensitive than GUIDE-seq in a head-to-head</td>
<td>WGS may be costly; must be paired with one of the other methods to validate the sites discovered in vitro are truly mutagenized in the cell</td>
</tr>
<tr>
<td>Computational Guide Selection</td>
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<tr>
<td>Specificity score (Hsu et al., 2013) and CFD (Doench et al., 2016)</td>
<td>an experimentally derived function to score the likelihood of an off-target site being edited, on the basis of the position and nature of its mismatches</td>
<td>a continuous variable to distinguish guide RNAs during design and to rank off-target sites for targeted sequencing follow-up</td>
<td>score is derived from SpCas9 data with 20-mer guides, so it may not generalize to other contexts</td>
</tr>
<tr>
<td>WGS of reference genome (Yang et al., 2014)</td>
<td>WGS of the relevant cell line, animal model, patient, etc. (Box 1)</td>
<td>identify new off-target sites created by genetic variation, which are not present in the reference genomes (i.e., hg38)</td>
<td>remains costly; custom reference genomes only available with some guide design tools (Table 2)</td>
</tr>
<tr>
<td>Protein Engineering</td>
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<tr>
<td>Single/paired nickases (Cong et al., 2013; Mali et al., 2013a; Ran et al., 2013a)</td>
<td>a point mutation at the active site of one of the Cas9 nuclease domains yields a targeted nickase; paired nickases targeting complementary strands create a deletion</td>
<td>nicks are repaired with higher fidelity than DSBs, so off-target edits are less frequent; nickases can also mediate efficient homology directed repair</td>
<td>less efficient on-target editing with some guide RNAs</td>
</tr>
<tr>
<td>SpCas9 PAM Variant D1135E (Kleinstiver et al., 2015b)</td>
<td>a single point mutation increases specificity for the 5'-NGG PAM</td>
<td>significant decrease in editing at 5'-NAG and 5'-NGA PAMs; improved genome-wide specificity for several guides as assessed by GUIDE-seq</td>
<td>on-target efficiency may be affected; it was comparable with wtCas9 for six guides by T7E1 but somewhat lower for three guides by deep sequencing</td>
</tr>
<tr>
<td>eSpCas9 (Slaymaker et al., 2016)</td>
<td>three mutations within the nt-groove weaken Cas9's non-target DNA strand stabilization and therefore increase stringency of guide RNA-DNA complementation for nuclease activation</td>
<td>off-target editing was nearly entirely avoided, as assessed by both BLESS and targeted deep sequencing</td>
<td>on-target efficiency may be affected, although it was comparable with wtCas9 for most of the guide RNAs reported</td>
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(Continued on next page)
screen analysis of 9,914 guide RNA variants covering all single mismatch and single RNA indels for 65 DNA targets was used to derive a cutting frequency determination (CFD) score capable of predicting the likelihood of off-target activity (Doench et al., 2016). Overall, these data underscore the complexity of predicting Cas9 specificity and the importance of considering both position and mismatch identity in metrics like the CFD. For example, 256 of 289 (89%) rG:dT mismatched guides retained high activity, whereas only 37% of rC:dC mismatches were active. Looking just at position 16 in the guide sequence, all 46 purine-purine mismatches tested attenuated Cas9 activity, whereas 97% of rN:dT mismatches were tolerated. Furthermore, in a validation study, modest correlation was reported between the CFD and observed off-target activity at 60 sites with one or two mismatches. Generally, well-designed guide RNAs avoid off-by-one and off-by-two mismatch sites, so these models should be improved to predict activity at off-by-three and off-by-four mismatch sites. Future refinements and insights will likely be derived from large training data sets that consider off-target sites with more than one mismatch or indel.

Last, the online tools compared in Table 2 will be of practical utility for most users, but bioinformaticians may benefit from the performance and flexibility of offline tools, such as CRISP-Rseek (Zhu et al., 2014). A comprehensive comparison of these computational tools can be found with the CRISPR Software Matchmaker tool (MacPherson, 2015).

**Cas9 Orthologs and New CRISPR Proteins**

Recently, a broader array of Cas9 ortholog proteins have been engineered as genome editing tools, such as the *Staphylococcus aureus* Cas9 (SaCas9), which is more than 1 kb smaller than the more commonly used SpCas9 and thus easily packaged into AAV paired with guide RNA expression cassettes for in vivo genome editing (Ran et al., 2015; Nishimasu et al., 2015; Friedland et al., 2015). Interestingly, SaCas9 demonstrates highly efficient nuclease activity with target sequences of 21–24 nt compared with the 20 nt targets used with SpCas9, while using a 5’-NNGRRT PAM as opposed to the 5’-NGG of SpCas9. SaCas9 also demonstrates substantial cleavage efficiency at 5’-NGRR PAM sites. It remains to be determined how a longer target sequence and longer PAM preference

<table>
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<tbody>
<tr>
<td>SpCas9-HF (Kleinstiver et al., 2016a)</td>
<td>four mutations weaken Cas9 binding to the target DNA strand and therefore increase stringency of guide RNA-DNA complementation for nuclease activation</td>
<td>off-target editing was nearly entirely avoided, as assessed by both GUIDE-seq and targeted deep sequencing</td>
<td>on-target efficiency may be affected, although it was comparable with wtCas9 for most of the guide RNAs reported</td>
</tr>
<tr>
<td>tru-guides (Fu et al., 2014b)</td>
<td>tru-guides have spacers of 17–18 nt as opposed to the canonical 20 nt; the shorter region of RNA:DNA complementarity allows fewer mismatches</td>
<td>simple to implement; can decrease activity at genomic off-target sites</td>
<td>less efficient on-target editing with some guide RNAs; some guides may have more off-target sites with high-sequence homology</td>
</tr>
<tr>
<td>5’GGX20 guides (Cho et al., 2014)</td>
<td>two additional mismatched guanines are added to the 5’ end of the guide RNA; these PAM-distal mismatches may adjust binding energetics such that Cas9 more stringently requires complementarity</td>
<td>this simple modification can be rapidly assessed; specificity ratios at off-target sites generally were reported to improve by 10- to 100-fold or more</td>
<td>this modification is compatible only with certain guide RNAs; on-target efficiency is sometimes severely decreased</td>
</tr>
<tr>
<td>SaCas9 (Ran et al., 2015)</td>
<td>the Staphylococcus aureus Cas9 ortholog is 3.2 kb, uses a 20–23 nt guide RNA, and recognizes a 5’-NNGRRT PAM</td>
<td>smaller size allows AAV packaging with one or two sgRNAs in an all-in-one vector; specificity improvements were observed with some guides by BLESS, GUIDE-seq, and targeted sequencing</td>
<td>the ortholog is less characterized than SpCas9 and may be less efficient on-target in some contexts</td>
</tr>
<tr>
<td>RNP (Cho et al., 2013b; Kim et al., 2014; Lin et al., 2014; Zuris et al., 2015)</td>
<td>purified Cas9 protein is complexed with guide RNA before delivery to cells by lipid transfection or electroporation</td>
<td>Burst-like Cas9 expression kinetics mediates efficient on-target editing without long-term risk of accumulating off-target edits</td>
<td>best used ex vivo</td>
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</table>

gDNA, genomic DNA; NGS, next-generation sequencing; tru-guides, truncated guides.
affects the genome-wide off-target activity of SaCas9, although initial results with the unbiased BLESS and GUIDE-seq protocols appear promising (Friedland et al., 2015; Ran et al., 2015). Given identical guides targeting sites with nested 5′-(NGG)RRT PAMs, editing with SaCas9 as opposed to SpCas9 resulted in lower indel frequencies at fewer off-target sites (Ran et al., 2015).

Cas9s derived from Neisseria meningitidis (NmCas9) and Streptococcus thermophilus (StCas9) CRISPR1 and CRISPR3 have also been engineered for specific genome editing in mammalian cells (Mali et al., 2013a; Müller et al., 2016). Similar to SaCas9, NmCas9 requires a more complex 5′-NNNNGATT PAM and uses a longer crRNA spacer (21–24 nt) than SpCas9 (Hou et al., 2013; Mali et al., 2013a; Lee et al., 2016). Importantly, although the longer PAM limits the targeting range, it also limits the number of off-target sites for any given guide. For example, NGG motifs occur roughly every 8 bp, whereas NNNNGATT is found every 128 bp (Lee et al., 2016). A systematic study of NmCas9 mismatch and indel tolerance found an overall improvement over SpCas9 and a promising absence of any off-target editing at all SpCas9 and three or fewer mismatches (Lee et al., 2016). However, the on-target efficiency is consistently below that of SpCas9, possibly because of weaker DNA unwinding activity (Ma et al., 2015). The two StCas9 orthologs similarly might be more specific than the wild-type SpCas9, but they also suffer from weaker efficiency (Müller et al., 2016).

Overall, Cas9 orthologs with improved specificity profiles can likely be discovered and engineered. By screening available genome databases for type II CRISPR loci, more than 56 bacterial species have been identified with predicted novel trans-activating CRISPR RNAs (tracrRNAs) (Chylinski et al., 2013). Furthermore, Cas9 proteins phylogenetically clustered within the same subgroup can interchangeably use their crRNA-tracrRNA (dual-RNA) duplexes to cleave DNA, suggesting that Cas9-guide RNA pairings could be optimized by mining the natural diversity or engineering improved guide architectures (Fontana et al., 2014).

Furthermore, the large natural diversity of bacterial CRISPR systems extends beyond type II systems featuring Cas9 as the sole effector nuclease. Recently, a newly discovered CRISPR protein named Cpf1 has also been used for human genome editing (Zetsche et al., 2015). Also targeted by a single guide RNA (sgRNA), Cpf1 recognizes a 5′-TTN PAM upstream (rather than downstream) of the target site and creates a 5 nt staggered overhang DSB distal to the PAM (Fonfara et al., 2016; Yamano et al., 2016). It is important to note that eight Cpf1 orthologs were screened in human cells, and only two mediated efficient cleavage. Cpf1 genome-wide specificity was recently assayed with Digenome-seq (Kim et al., 2016a) and GUIDE-seq (Kleinstiver et al., 2016b) and can be comparable with or greater than SpCas9 for many guide RNAs. Other putative CRISPR effector proteins, known as c2c1, c2c2, and c2c3, have also been bioinformatically mined from metagenomic data sets (Shmakov et al., 2015). C2c2 has since been characterized as a programmable RNA-guided RNA-targeting nuclease in bacterial cells (Abudayyeh et al., 2016). Interestingly, once c2c2 binds its

<table>
<thead>
<tr>
<th>Name</th>
<th>Website</th>
<th>Specificity Score</th>
<th>Genomes Supported</th>
<th>Single/ Paired Design</th>
<th>Cas9 Orthologs Supported</th>
<th>Custom Guide Length</th>
<th>Lists Off-Targets</th>
<th>Considers Off-Targets with Indels</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchling</td>
<td><a href="http://benchling.com">http://benchling.com</a></td>
<td>Hsu et al. (2013) and CFD (Doench et al., 2016)</td>
<td>22</td>
<td>both</td>
<td>5</td>
<td>yes</td>
<td>yes</td>
<td>no</td>
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<tr>
<td>BLAST (on pre-designed guide)</td>
<td><a href="http://blast.ncbi.nlm.nih.gov/">http://blast.ncbi.nlm.nih.gov/</a> Blast.cgi</td>
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complementary target RNA, it is active as a non-specific RNase and significantly impedes cell growth. It remains largely unclear how these new CRISPR proteins’ different characteristics (i.e., guide RNAs, PAMs, nuclease domains) will affect their specificity, but their discovery reflects great potential for genome engineering beyond SpCas9.

**Protein Engineering and Bespoke PAMs**

Meanwhile, biological engineers have rapidly optimized and successfully matured SpCas9 as a tool, with increasing precision as understanding of its molecular structure and mechanism has deepened (Table 1). The first generation of improvements leveraged knowledge of the two nuclease domains of Cas9, initially identified on the basis of sequence homology (Sapranauskas et al., 2011). For example, a single point mutation (D10A) can inactivate the RuvC domain and change the SpCas9 nuclease into a nickase capable of cleaving only the DNA strand complementary to the guide RNA (Cong et al., 2013). Using paired guide RNAs to create two appropriately offset nicks results in efficient indel formation while improving specificity by as much as 1,500-fold, likely because single DNA nicks are repaired with high fidelity compared with DSBs (Mali et al., 2013a; Ran et al., 2013a; Friedland et al., 2015). Paired nickases have been applied to efficiently generate mouse models with no detectable off-target edits, whereas wild-type Cas9 can generate off-target edits that are inherited by founders’ progeny (Shen et al., 2014). Beyond nickases, a second point mutation in the HNH catalytic domain creates a fully nuclease-inactive “dCas9” (Nishimasu et al., 2014). By fusing dCas9 to the FokI nuclease domain and again expressing paired guide RNAs, one can create a dimerization-dependent system that also improves specificity (Güllinger et al., 2014; Tsai et al., 2014).

Another recent protein engineering approach, known as SpCas9-MT-pDBD fusions, involves linking a mutated Cas9 with attenuated PAM-binding (SpCas9-MT) to a programmable DNA-binding domain (pDBD) such as a ZF or TALE that targets nearby genomic DNA (Bolukbasi et al., 2015). SpCas9-MT-pDBD provided specificity improvements up to 150-fold, likely by leveraging cooperativity between two independent DNA-binding events to improve specificity, much like the aforementioned nickase and FokI strategies.

However, these strategies have the overall disadvantage of requiring additional components and larger transgenes. Recently, a crystal structure-based, rational engineering strategy generated SpCas9 and SaCas9 nuclease with “enhanced specificity” (eSpCas9 and eSaCas9, respectively). These Cas9s differ by only three and four codon substitutions, respectively, within the non-targeted DNA strand groove (nt-groove) of Cas9 (Slaymaker et al., 2016). These mutations are thought to weaken non-target strand binding by the protein, encouraging DNA rehybridization in competition with guide RNA invasion of the target DNA strand (Figure 1). With increased stringency of RNA-DNA matching, these Cas9s demonstrate dramatic reduction in the number of genome-wide off-target sites detectable by BLESS or follow-up targeted sequencing. Moreover, eSpCas9’s on-target efficiency appeared to be comparable with wild-type SpCas9 for the 24 guide RNAs tested, suggesting a surprising de-coupling of efficiency and specificity without discretely modifying the mechanism of action (as accomplished by the cooperativity strategies).

A similar strategy by a different research group resulted in a “high-fidelity” SpCas9-HF (Kleinstiver et al., 2016a). In contrast with eSpCas9, SpCas9-HF contains four alanine substitutions at residues that interact with the phosphate backbone of the targeted DNA strand. Although on-target efficiency appeared to be slightly more variable than eSpCas9 for the guides tested, it remained comparable (70%) with wild-type efficiency for 86% of 37 guide RNAs. Altogether, these engineered Cas9s greatly improve specificity without much sacrifice in efficiency. Interestingly, these improved variants present a new challenge to improve the limit of detection of biased and unbiased assays beyond 0.1%, in order to capture remaining rare off-target events that may be important in clinical applications.

Another successful approach has been to alter the PAM recognition sequence of Cas9 itself. Because the PAM is perhaps the most stringent determinant of Cas9 specificity, Cas9s with alternative PAMs could not only increase the number of targetable sites across a genome of interest but also potentially demonstrate improved genome-wide specificity by requiring a longer PAM or a PAM with less abundance in the target genome. One strategy is to swap the putative PAM-interacting domain (PID) of the protein for the PID from an ortholog recognizing a different PAM. This has been previously demonstrated with SpCas9 and Streptococcus thermophilus CRISPR3 Cas9 (St3Cas9), which share 60% sequence identity and can swap dual guide RNA sequences while retaining function (Fonfara et al., 2014; Nishimasu et al., 2014).

Similarly powerful is the leveraging of directed evolution to alter the PAM specificity, as shown recently for SpCas9 (Kleinstiver et al., 2015b) and SaCas9 (Kleinstiver et al., 2015a). Notably, just four mutations resulted in a modified “VRE” SpCas9 with high specificity for NGCG PAMs, which are roughly 23 times less abundant in the human genome than the canonical NGG PAM for standard 5’-G-N19 guide RNAs. Directed evolution also yielded a single point mutation (D1135E) SpCas9 that increases specificity for NGG PAMs over the alternative NAG PAM (Hsu et al., 2013; Jiang et al., 2013; Kleinstiver et al., 2015b). In the case of SaCas9, directed evolution was applied to modify the PAM from NNGRTT to NNNRRT, in order to decrease the targeting range by 2- to 4-fold (Kleinstiver et al., 2015a).

One advantage of this unbiased directed evolution approach is that it does not require structural data and can be rapidly applied to diverse orthologs. It could prove important for therapeutic targets requiring a very specific edit and lacking nearby canonical PAMs. Finally, bespoke PAMs are likely compatible with the enhanced specificity and high fidelity mutations. Chimeras, directed evolution, and other protein engineering approaches may also be fruitful for generating Cas9s that recognize AT-rich PAMs, which could be useful for targeting intronic regions or AT-rich genomes.

**dCas9 Specificity**

Aside from DNA cleavage, a large collection of Cas9 applications use catalytically inactivated Cas9 (dCas9), which can be fused to...
various effector domains for applications ranging from programmable chromosomal labeling to targeted transcriptional and epigenetic control (Hsu et al., 2014). However, the specificity profile of dCas9 binding versus wild-type (wtCas9) cleavage applications needs to be carefully considered. Recent work studying the target search mechanisms (Figure 1) of the active Cas9 enzyme indicates that Cas9 initially scans the genome for PAM matches by diffusion and undersamples heterochromatin regions under certain conditions (Knight et al., 2015).

(A) Cas9 in complex with guide RNA samples the genome for PAM matches by diffusion and undersamples heterochromatin regions under certain conditions (Knight et al., 2015).
(B) Upon PAM recognition and double-stranded DNA binding, the DNA is directionally unwound as the guide RNA interrogates the target strand for complementary sequences.
(C) The non-target single-stranded DNA (ssDNA) strand is stabilized in a non-sequence-specific manner in Cas9’s nt-groove. Incomplete complementarity or a short 15 nt PAM-proximal region of complementarity is sufficient for Cas9 binding, as demonstrated by dCas9 or “dead guides” with wtCas9, respectively. Near-complete complementarity of >17 nt of RNA:DNA heteroduplex is required for nuclease activation.
(D) After this RNA strand invasion, the HNH domain undergoes a conformational change and communicates through a linker to the RuvC domain. The HNH and RuvC catalytic domains simultaneously cleave the target and non-target ssDNA strand, respectively. Cas9 may remain bound to the cut site for an extended period before returning to the pre-target state (Richardson et al., 2016).

The model shown is adapted from several sources (Anders et al., 2014; Nishimasu et al., 2014; Sternberg et al., 2014, 2015; Jiang et al., 2016; Slaymaker et al., 2016).

Figure 1. Model of Cas9 DNA Targeting and Cleavage
(A) Cas9 in complex with guide RNA samples the genome for PAM matches by diffusion and undersamples heterochromatin regions under certain conditions (Knight et al., 2015).
(B) Upon PAM recognition and double-stranded DNA binding, the DNA is directionally unwound as the guide RNA interrogates the target strand for complementary sequences.
(C) The non-target single-stranded DNA (ssDNA) strand is stabilized in a non-sequence-specific manner in Cas9’s nt-groove. Incomplete complementarity or a short 15 nt PAM-proximal region of complementarity is sufficient for Cas9 binding, as demonstrated by dCas9 or “dead guides” with wtCas9, respectively. Near-complete complementarity of >17 nt of RNA:DNA heteroduplex is required for nuclease activation.
(D) After this RNA strand invasion, the HNH domain undergoes a conformational change and communicates through a linker to the RuvC domain. The HNH and RuvC catalytic domains simultaneously cleave the target and non-target ssDNA strand, respectively. Cas9 may remain bound to the cut site for an extended period before returning to the pre-target state (Richardson et al., 2016).

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In accordance with this model, dCas9 immunoprecipitation studies demonstrate a high promiscuity of Cas9 binding that correlates poorly with Cas9 cleavage (Kuscu et al., 2014; Wu et al., 2014; Ran et al., 2015): across nearly 300 off-target binding loci identified via dCas9 chromatin immunoprecipitation sequencing (ChIP-seq), only 1 of these sites was mutated slightly above background level by wtCas9 (Wu et al., 2014). In a GUIDE-seq
genome-wide study of DNA cleavage, only 2% of off-target sites intersected with a set of 585 dCas9 ChIP-seq off-target binding sites (Tsai et al., 2015). Moreover, dCas9 ChIP-seq identified sites with about seven mismatches on average, whereas cleavage sites harbored about three mismatches on average (Tsai et al., 2015). Although a useful unbiased genome-wide assay for specificity of dCas9 applications, one should not consider dCas9 ChIP-seq as a proxy for genome-wide detection of “true” cleavage events.

Just as for its nuclease counterpart, a combination of biased and unbiased approaches provides the most reliable assessment of dCas9 off-target activity. One method is a mismatch reporter assay in which a library of mismatched target sites is cloned upstream of a promoter driving expression of barcoded RNAs (Mali et al., 2013a). RNA sequencing (RNA-seq) is then used to determine the extent to which that mismatched target site enabled dCas9-mediated activation or repression of the promoter. This method revealed that dCas9-effector fusions are largely insensitive to point mutations and can repress or activate transcription while tolerating up to three mismatches, especially near the 5′ end of the guide RNA. That mismatch tolerance may be less than that measured by ChIP-seq because of the longer duration of promoter binding required to make the transcriptional regulatory function detectable.

Whole-transcriptome RNA-seq also works as an unbiased method for dCas9-effector off-target analysis (Perez-Pinera et al., 2013; Konermann et al., 2015; Polstein et al., 2015). Generally, this assay confirms the highly specific upregulation of the genes targeted by dCas9 activators. This specificity is unsurprising given that dCas9 activators are efficient only within a limited window of the transcription start site: binding events >200 nt upstream of a start site will have little effect on transcription (Gilbert et al., 2014; Konermann et al., 2015). Notably, one RNA-seq study identified a moderate yet significant downregulation of IL32 mRNA, a proinflammatory cytokine, in response to dCas9-VP64 transcriptional activators even without a guide RNA (Perez-Pinera et al., 2013). This result emphasizes the importance of considering off-target effects of Cas9 beyond the first-order risk of binding or cutting activity at a degenerate target site, such as unintended biological activity in response to the introduced genome editing components (see Broader Implications of Specificity).

Modified Guides

Although the aforementioned strategies engineer the Cas9 protein, specificity improvements may also be achieved by modifying the guide RNA (Table 1). Recent characterization of single and dual-guide RNA systems for type II CRISPR systems identified six functional modules within the guide (Briner et al., 2014). These modules can be mutated, extended, and deleted to modulate on-target efficiency. For example, extending the sgRNA duplex region by 5 nt and/or mutating the string of four Ts shortly downstream of the spacer can improve efficiency by changing the RNA’s structure and transcription rate, respectively (Chen et al., 2013; Dang et al., 2015). This approach also improves efficiency for SaCas9 sgRNAs, indicating the general applicability of guide modifications across multiple Cas9 orthologs (Chen et al., 2016; Tabebozbar et al., 2016). Further investigation into guide RNA features will provide a basis for the optimization of guide RNAs for specificity, functionalization, and other applications.

The length of the guide sequence itself is also an important mediator of specificity. Although extending the 20 nt guide sequence of SpCas9 to 30 nt appeared to result in processing back to the natural 20 nt length (Ran et al., 2013a), shortening the guide sequence proved to be a successful strategy. Spacer sequences of 17 or 18 nt have been reported to mediate more precise genome editing, potentially by increasing the sensitivity of Cas9 binding to mismatches within the shorter complementary sequence (Fu et al., 2014b). Although truncated guides can eliminate activity at many off-target sites, they can also create new off-target sites because of their shorter length (Fu et al., 2014b; Tsai et al., 2015; Wyvekens et al., 2015; Slaymaker et al., 2016). Interestingly, truncated guides are not compatible with the “enhanced specificity” nt-groove mutants or SpCas9-HF, as the combination results in very low on-target efficiency (Kleinstiver et al., 2016a; Slaymaker et al., 2016). It remains to be seen if this simple guide design modification could improve the specificity of other Cas9 orthologs while maintaining editing efficiency, which also varies by guide length (Friedland et al., 2015; Ran et al., 2015).

Notably, shortening the guide further to 14 or 15 nt creates a “dead guide” that can mediate efficient wild-type Cas9 binding without nuclease activation, enabling multiplexed applications of DNA cleavage and targeted effector activity (Dahlman et al., 2015; Kiani et al., 2015). This engineered guide approach benefited from insight into the mechanism of Cas9 cleavage, which has been shown to require near complete strand invasion and complementarity (Figure 1). Whole-transcriptome RNA-seq revealed highly similar specificity profiles for both 15 and 20 nt guides applied for transcriptional activation (Dahlman et al., 2015). Although “dead guides” may have more predicted off-target binding sites, they likely exhibit acceptable specificity (fewer than three off-target genes significantly activated) because off-target sites must fall within a limited window of a transcription start site in order to perturb transcription.

Another simple modification, known as “GGX20,” adds two additional mismatched guanine nucleotides to the 5′ end of the guide RNA (Cho et al., 2014). Several studies have demonstrated specificity improvements of 10- to 100-fold or greater, although the on-target efficiency varies from comparable to severely reduced (Cho et al., 2013a, 2014; Sung et al., 2014; Kim et al., 2016b). Despite such variability, “GGX20” and truncated guides are easily generated and assessed, although they should still be compared with standard-length guides on a case-by-case basis to optimize on-target efficiency.

Finally, chemical modification of guide RNAs can also impact Cas9 specificity. Synthesis of guide RNAs with 2′-O-methyl (M), 2′-O-methyl 3′phosphorothioate (MS), or 2′-O-methyl 3′thioPACE (MSP) at the 5′ and 3′ ends significantly improves editing efficiency in human primary T cells and CD34+ hematopoietic stem and progenitor cells (Hendel et al., 2015). However, these RNA modifications increased off-target activity at some sites. In a contrasting study, synthetic guide RNAs with 10 M-modified 5′ bases and interspersed 2′-fluoro modifications toward the 3′ end were more stable and specific
but less efficient on-target (Rahdar et al., 2015). Given these mixed early findings, the ability to modulate RNA stability and mismatch binding kinetics through alternative nucleotide chemistries of guide RNAs warrants further exploration.

**Kinetics and Regulation**

A critical and underexplored determinant of Cas9 specificity is the pharmacokinetic profile of the delivered components across cell and tissue types, especially the form factor of the genome editing components (DNA, RNA, or protein) and the delivery vehicle (viral or non-viral).

Lentiviral-mediated integration of Cas9 and guide RNA into melanoma cell lines has been reported to achieve saturating indels up to 100% at on-target sites (Shalem et al., 2014). Although both on- and off-target activity generally appeared stable between days 7 and 14, indels at one off-target site with three mismatches rose from ~40% to ~50%. Because off-target levels can increase with sustained exposure to Cas9, controlling or limiting the time of Cas9 expression in the cell or nucleus may improve specificity.

As such, the Cas9 delivery modality is an important mediator of specificity (Table 1). For example, cationic lipid-mediated ribonucleoprotein (RNP) delivery improved the specificity of three SpCas9 guides at 11 assessed off-target sites (typically by 10-fold) compared with plasmid DNA transfection, after normalizing for equivalent on-target efficiency (Zuris et al., 2015). In a similar study, RNP-Cas9 improved the on-target/off-target ratio by approximately 3-fold for two guides tested (Kim et al., 2014).

Delivering MS-modified synthetic guide RNAs as RNPs significantly improved on/off-target ratios at all 3 assessed sites compared with co-delivery with Cas9 plasmid or mRNA (Hendel et al., 2015). This specificity benefit is possibly due to the burst-like kinetics of RNP delivery and associated Cas9 exposure as opposed to more stable plasmid-driven expression (Kim et al., 2014).

Cas9 dosage can also affect the kinetics and modulate specificity, both in vitro (Pattanayak et al., 2013; Fu et al., 2014a) and in cells (Hsu et al., 2013). For example, by plasmid transfection, a 5-fold drop in dose led to a 7-fold increase in the specificity ratio at the cost of a ~2-fold decrease in on-target efficiency, emphasizing the importance of titrating delivery conditions to optimize specificity for any given application (Hsu et al., 2013). Using the weaker H1 promoter for guide RNA expression gives a greater decrease in off-target activity than on-target activity (Rangathan et al., 2014). Such non-linearity in the on-target/off-target editing rate relationship with Cas9 or guide RNA concentration reflects the significant, yet incomplete, coupling of the target-finding and nuclease activities of the enzyme. Single-molecule imaging studies performed with very low Cas9 concentrations revealed undersampling of heterochromatin regions, suggesting that DNA site accessibility can also play a role in specificity under certain conditions (Knight et al., 2015).

Small molecules can also modulate the kinetics of CRISPR editing. For example, a small molecule-dependent Cas9 was developed by genetically inserting an evolved ligand-dependent intein, which increased specificity upon induction by 2- to 25-fold at 11 assessed off-target sites for three guide RNAs, while showing very limited nuclease activity in the absence of the small molecule (Davis et al., 2015). However, the on-target efficiencies were also generally less than that of wild-type Cas9, again demonstrating the cost associated with some efforts to engineer improved specificity. Small molecules can also be used to modulate the cellular response to genome editing. For example, the rate of homology-directed repair over non-homologous end-joining has been improved by using a DNA ligase IV inhibitor (Maruyama et al., 2015) and other molecules with poorly understood mechanisms of action (Yu et al., 2015). Additionally, synchronizing the cell cycle has been suggested to increase the homology-directed repair rate with RNP delivery in human cells (Lin et al., 2014).

**Broader Implications of Specificity**

Proper consideration of Cas9 genomic specificity should include not only the aggregate number of potential off-target sites for a given guide RNA but also the physiological impact of individual off-target events (both observed and not). Although tumorigenesis and unwanted editing of oncogenic tumor suppressors is a common concern, editing events that negatively impact the cell’s viability or functional capabilities also need to be considered. Moreover, the ability to measure off-target editing differs significantly by tissue. Although blood can be sampled relatively easily for genomic DNA sequencing, other tissues cannot be so simply biopsied. As a result, linking observed phenotypes with Cas9-induced genome modifications could be difficult in certain organs. The “acceptable” thresholds for off-target rates could also differ in terminally differentiated cells, as opposed to dividing cells. Certainly, “acceptable” specificity thresholds may not be universal across applications, if they are even determinable, as ultimately empirical demonstration of safety will be of paramount importance in the clinic.

Furthermore, the specificity of Cas9 nucleases is not necessarily confined to genomic specificity, especially for in vivo applications. For example, tissue and cell-type specificity are also important considerations when targeting genetic disorders that primarily affect certain organs or cells. However, selective distribution may be mediated by a careful choice of the delivery method. For example, different AAV serotypes present different cell-type tropisms and blood clearance rates when administered systemically (Zincarelli et al., 2008). Alternative routes of administration can also provide localized or tissue-specific expression, as well as immunogenicity benefits (Ge et al., 2001; Limberis and Wilson, 2006). Whereas gene augmentation therapy approaches previously demanded viral serotypes that provided sustained and high transgene expression (Hastie and Samulski, 2015), Cas9 applications may be safer with vectors that deliver short-term expression in a highly specific cell type.

Cell-type specificity may also be achieved with transcriptional regulatory elements. RNA polymerase II tissue-specific and cell type-specific promoters (e.g., TBG, Synapsin, GFAP) capable of driving protein-coding transcripts such as Cas9 could be combined with novel guide RNA promoters and expression strategies beyond the U6 and H1 RNA polymerase III promoters long used in the RNAi field for expressing small interfering RNA (siRNA) and small hairpin RNA transcripts. A Csy4-dependent method allows RNA polymerase II promoters to express guide RNAs (Nissim et al., 2014), permitting simple multiplexing as
The simplest way to avoid most off-target effects is to design guide RNAs that have few off-target sites with high sequence homology (i.e., zero, one, or two mismatches). However, mutations in the genome or the guide RNA can theoretically create off-target sites with greater sequence homology than expected.

**MUTATIONS IN THE GENOME**

It can be important to consider the reference genome used with these design tools. The current standard practice is to design guide RNAs against the species of interest’s standard reference genome (i.e., hg38), yet variation in the particular genome of the edited tissue, cell line, or strain can be an important determinant of guide specificity. In fact, one group recently observed an off-target event of unexpectedly high activity (36.7%) when targeting human induced pluripotent stem cells, because of a SNP that converted a three-mismatch site into a two-mismatch site on one allele (Yang et al., 2014). For CRISPR-Cas9 applications that require exquisite specificity, WGS and an understanding of patient or genotypic variation may become an important initial step of guide design.

**MUTATIONS IN THE GUIDE RNA**

Guide RNAs can be introduced to the cell in several ways with varying fidelity. For example, DNA oligos are prone to small deletions, as the coupling reaction that adds each base is 99% efficient. Digenome-seq reported “RNA-bulge”-type off-targets with guides transcribed from an annealed-oligo template but no such off-target sites with guides transcribed from a clonal plasmid DNA template (Kim et al., 2016b). Effectively, 1 nt deletions in the oligo pool were leading to “false positives” at off-target sites that would not be edited with other delivery methods. When using RNP delivery, note that RNA oligo synthesis is less efficient (~98%) and thus more prone to deletions. As an alternative, one can transcribe RNA in vitro from sequence verified plasmid with T7 polymerase, which has a more acceptable error rate of ~1/20,000 (Huang et al., 2000). The error rate of Pol III transcription is ~1 x 10^-7 in yeast and may be similarly low for guide RNAs transcribed from the human U6 promoter (Alic et al., 2007). Thus, guide RNAs expressed by plasmid and viral delivery likely boast high fidelity, but direct delivery of synthesized oligos should be examined closely for some guide RNAs.

well as expression restricted to certain cell types. Additionally, microRNA binding sites could be added to the 3' end of the Cas9 mRNA to limit expression in certain cells. Reported to be antigen-presenting cell (APC)-specific, mir-142-3p has been exploited in this manner to effectively limit antibody and cellular responses to transgenes by repressing expression and peptide presentation in APCs (Majowicz et al., 2013).

Finally, another means of controlling the specificity of Cas9-mediated activity may be through optogenetic control. As previously reported with TALEs (Konermann et al., 2013), harnessing light-inducible heterodimerizing proteins to dCas9 (CIB1-dCas9) and a transcriptional activator (CRY2-VP64) provides for optogenetically controlled transcriptional activation (Polstein and Gersbach, 2015). Although background activity remains to be optimized, this approach could facilitate effective manipulation of the spatial and temporal specificity of a dCas9 synthetic transcription factor.

**The Need for Standardization**

As demonstrated, the field of Cas9 specificity is developing rapidly and has already accomplished dramatic improvements over the first-generation methods. In thinking about developing medicines, it is important to keep in mind that all methods for manipulating therapeutic targets, ranging from genome editing reagents and siRNAs to antibodies and small molecules, are prone to unintended off-target effects. The expectation for the specificity of biological engineering and Cas9 in particular should be continuous improvement and appropriate vigilance. Striving for ever improving enzymes and measurement technologies with improved lower levels of detection while carefully assessing candidate therapies using state of the art functional models and assays will be important to making appropriate pre-clinical decisions for genome editing therapies.

To rigorously benchmark progress, the genome editing community should strive to standardize the best-performing methods for measuring and reporting off-target activity (Joong, 2015). For example, the commonly used Surveyor assay inherently underreports high levels of indels because of the heteroduplex dependency of the Surveyor enzyme (Vouillot et al., 2015). Moreover, Surveyor quantification by gel image analysis or electrophoresis devices like the Qiagen Qiaxcel or Agilent TapeStation has limited sensitivity compared with targeted deep sequencing. As such, targeted sequencing should become the standard assay for validating off-target events.

Targeted deep sequencing of off-target events is particularly amenable to standardization and reporting of developments in the field, although the method is variably applied across the literature in terms of experimental execution and computational analysis. In our view, it is important to minimize PCR cycle number, use error-correcting DNA polymerases, consider the sequencing coverage per input genome copy, and compare reads with appropriate reference sequences that take into account genetic variation from the standard reference genomes (Box 1). Many groups currently use custom analysis pipelines, so raw data and code should be shared when possible via web repositories such as Github. Ideally, robust analysis tools will
be widely adopted as standard pipelines, as exemplified by the Tuxedo suite (Kim et al., 2013; Trapnell et al., 2013) for RNA-seq, CRISPR-GA (Güell et al., 2014) and CRISPRESSO (http://crispesso.rocks) are two publicly available indel analysis packages that are available as online or command-line tools.

In addition, standard guide RNAs for specificity experiments already help researchers compare methods and should continue to be used as benchmarks. For example, each of the BLESS, HTGTS, GUIDE-seq, IDLV capture, and Digenome-seq papers assesses at least one shared guide RNA (targeting either EMX1 or VEGFA). The different methods identified intersecting but unequal sets of off-targets for the standard guide RNAs, confirming that none of the unbiased methods are yet perfectly sensitive and reliable. However, the off-target sites detected across multiple studies are also generally the most frequently detected and edited, whereas the others are very rare, an encouraging result for the field at large (Gabriel et al., 2015).

We propose that a set of guidelines for reporting CRISPR-Cas9 specificity research should be established, much like the minimum information for publication of quantitative real-time PCR experiments guidelines for qPCR (Bustin et al., 2009), the minimum information about a microarray experiment for microarrays (Brazma et al., 2001), the minimum information about a proteomics experiment for proteomics (Taylor, 2006; Taylor et al., 2007; Martínez-Bartolomé et al., 2014), the minimum information specification for in situ hybridization and immunohistochemistry experiments for fluorescence in situ hybridization experiments (Deutsch et al., 2006, 2008), as well as others (Chervitz et al., 2011). Such guidelines are directed toward ensuring unambiguous communication of experimental designs and results, such that independent scientists can fairly interpret their conclusions in the context of the broader field while reproducing those results without requiring additional information. These “minimum information” requirements should include the number of off-target sites assessed, the off-target prediction constraints, cell type, delivery method, time point, dosage, analysis method, computational tools, sample size, and other key factors known to impact specificity.

The research community could begin to report guide RNA on- and off-target activity to shared databases, such as EENdb (Xiao et al., 2013) or CrisprGE (Kaur et al., 2015). Such standardized information, in combination with continued progress in off-target detection methods and Cas9 optimization, could greatly improve guide selection and enable fair comparison and benchmarking across publications while ultimately enabling precise and safe genome editing. Importantly, such standards must be developed with a focus on maximizing end-user adoption rates, minimizing usability hurdles, and incorporating user feedback, because “a standard that is not widely used is not really a standard” (Chervitz et al., 2011). Historically, working group scientific consortia have led the development of “Omics” standards while scientific journals and funding agencies have been critical in enforcing their broad compliance. Aligning genome editing stakeholders toward these goals would represent a discrete step forward in the maturation of this field.

Conclusions

The scientific landscape around Cas9 specificity is rapidly evolving, with marked improvements recently achieved in guide selection, protein and guide engineering, off-target detection methods, and more. Notably, two recent protein engineering approaches (eSpCas9 and SpCas9-HF) have elegantly increased SpCas9 specificity, most likely by reducing tolerance for mismatched DNA binding, although their overall on-target efficiency relative to the wild-type enzyme remain to be studied by more groups in different applications and organisms. These highly specific variants should encourage development of refined off-target detection methods with limits of detection well below 0.1%. Cas9 orthologs and non-Cas9 CRISPR enzymes are also emerging as genome editing tools, but will ultimately face the same challenges (Shmakov et al., 2015; Zetsche et al., 2015). In order to propagate these advances throughout the broader research community, the genome editing field will greatly benefit from discussion and consensus about best practices for future research (Box 2) as well as data standardization. Beyond specificity, key challenges and opportunities for the field lie in improving delivery and increasing efficiency, especially for precise edits.

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REFERENCES


Intervening sequences of regularly spaced prokaryotic repeats derive from vectors transduce murine alveolar and nasal epithelia and can be readministered by the CRISPR-Cas9 system enables specific genome editing in mammalian cells. Molecular Cell


