Flexible CRISPR library construction using parallel oligonucleotide retrieval

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Received November 18, 2016; Revised March 06, 2017; Editorial Decision March 07, 2017; Accepted March 14, 2017

ABSTRACT

CRISPR/Cas9-based gene knockout libraries have emerged as a powerful tool for functional screens. We present here a set of pre-designed human and mouse sgRNA sequences that are optimized for both high on-target potency and low off-target effect. To maximize the chance of target gene inactivation, sgRNAs were curated to target both 5′ constitutive exons and exons that encode conserved protein domains. We describe here a robust and cost-effective method to construct multiple small sized CRISPR library from a single oligo pool generated by array synthesis using parallel oligonucleotide retrieval. Together, these resources provide a convenient means for individual labs to generate customized CRISPR libraries of variable size and coverage depth for functional genomics application.

INTRODUCTION

The discovery of the bacterial CRISPR/Cas9 endonuclease system and its subsequent adaptation for genome editing in mammalian cells has created new opportunities for high throughput, functional genomics screens (1). For loss-of-function CRISPR screens, the most widely used CRISPR/Cas9 system is the Streptococcus pyogenes Cas9 (SpCas9) enzyme. SpCas9 can utilize a single guide RNA (sgRNA) that contains a 20-nt targeting sequence to introduce double-stranded DNA breaks in the genome in a sequence-specific manner. The SpCas9 target site must be upstream of a ‘NGG’ protospacer adjacent motif (PAM). Repair of the double-stranded DNA break generated by SpCas9 through non-homologous end joining introduces small indel mutations at the break site. If the break site is within the coding region of a gene, these indel mutations frequently result in frame-shift mutations that knock out gene expression (2). Thus, virtually any gene can be knocked out by co-expressing SpCas9 and a sgRNA in cells.

Large pooled lentiviral CRISPR libraries have been recently generated for loss of function screens with SpCas9. To construct these libraries, sgRNA sequences are first synthesized as a pool of DNA oligos using array-based oligo synthesis and then PCR cloned into lentiviral expression vectors (1). Currently, most ready-to-use CRISPR libraries are genome-wide libraries that contain tens of thousands of distinct sgRNAs in a single pool (3–8). Although useful for genome-wide screens, these libraries are technically more challenging to generate, maintain and screen due to their large size. In addition, sgRNA representation in these libraries can be adversely affected by their high degree of pool complexity. In many experimental settings, smaller, customized CRISPR libraries are desirable. For example, one might wish to screen a CRISPR library targeting only a specific gene family (e.g. kinases) or a subset of genes that constitute a gene expression or mutational signature. Small libraries might also be more desirable for their utility in in vivo screens. To carry out these ‘focused’ screens, a cost-effective, easy-to-implement method is necessary to enable individual labs to generate custom CRISPR libraries on demand.

In this study, we describe a flexible CRISPR library construction method to enable the generation of sub-genome scale, focused CRISPR sgRNA libraries of variable sizes. Our method uses parallel PCR retrieval to recover subsets of oligos from a master oligo pool generated by array-based synthesis. We demonstrate that sgRNA representations in these smaller libraries are highly consistent. We also generated a pre-designed database of sgRNA sequences targeting most human and mouse protein coding genes that can be used for selecting sgRNAs for custom library construction. These resources will enable easy creation of customized CRISPR libraries for genetic screens.
MATERIALS AND METHODS

Genome-wide sgRNA libraries targeting coding exons and protein domains for the human and mouse genomes were constructed following the steps described below. The approach is the same for both the human and mouse libraries. For the first step, we curated sgRNA sequences from all common exons and exon regions in protein coding genes. For each gene, all coding exons were identified using the NCBI Consensus Coding Sequence (CCDS) database (release# 106 for human, and release# 104 for mouse). For genes with multiple alternative transcripts, only exons or parts of exons (if the alternative splicing site is inside the exon) that existed in all transcripts were selected. For these common exons and exon regions, all S. pyogenes Cas9 sgRNA sequences in the form of (N)$_{20}$NGG in both strands DNA were selected. This results in a database of $4.3 \times 10^6$ sgRNA sequences for human and $4.5 \times 10^6$ sgRNA sequences for mouse.

In the second step, we generated scores that predicts the potency and on-target effect of each sgRNA. The on-target potency score (range 0–1) of each sgRNA was calculated using a previously published algorithm (9). To minimize off-target effects, each sgRNA sequence was mapped to its respective reference genome (hg38 for human and mm10 for mouse) using the Bowtie short read aligner (10), allowing up to three base mismatches. An off-target score for each potential off-target site identified was calculated using a modified version of a previously published algorithm (11), where we only used the empirically-determined mismatch weight table to score the sgRNA's off-target effects. Summing the number of potential off-target sites and each site's off-target scores, we calculated separate guide scores (range 0–100) for off-target sites within only the coding region (guide_c score), for off-target sites within only the non-coding region (guide_nc score), and for off-target sites in the genome (guide_g score) for each candidate sgRNA to measure its uniqueness within the genome. A higher guide score predicts the candidate sgRNA to have less off-target activity (12).

To curate sgRNAs for the exon library, we started by only considering sgRNAs that have no perfect match off-target sites in the genome. Among these, we initially only considered those with potency scores $\geq 0.4$, guide_c scores $\geq 50$ and guide_nc scores $\geq 10$. For each gene we chose up to six sgRNAs with the highest potency scores that also satisfy the following criteria: if a gene has only a single coding exon, all sgRNAs were chosen from that exon, if a gene has two coding exons, up to three sgRNAs were selected from each exon; and if a gene has $\geq 3$ coding exons, we selected sgRNAs from the first five coding exons with no more than two sgRNAs from the same exon. In addition, no two sgRNAs in the library were allowed to have $\geq 15$ nucleotides overlap with each other, and sgRNA sequences that contain the restriction site sequence for BsmBI were excluded. For the small number of fusion genes (read-throughs) and for families of genes that have the same annotated exon genomic locations in the CCDS database, only one sgRNA was chosen for each genomic location. For the majority of genes, we were able to curate 6 sgRNAs per gene. For some genes which are mostly small genes, fewer than 6 sgRNAs satisfied the above selection criteria. In these cases, we allowed the potency score cut-off to be first relaxed to $\geq 0.3$ and then relaxed to $\geq 0.2$ in order to curate additional sgRNAs. If after these two rounds of score relaxation we were still unable to curate six sgRNAs, we did not attempt to further relax the selection criteria. Thus, for a minor fraction of gene there were less than six sgRNAs/gene in the library.

To curate sgRNAs for the domain library, we first used the Pfam protein domains database (version 30) to define the regions in each protein that mapped to conserved domains. The genomic location of each domain was extracted from UCSC genomic databases (hg38 for human and mm10 for mouse) by using the ucscGenePfam and pfasDecs databases to map amino acid sequences to their corresponding exon regions. Intersection of these protein domain exon regions with the total sgRNA database from step 1 above allowed us to extract all sgRNAs mapped to Pfam protein domains in all genes. For each gene, we chose up to six sgRNAs from the set of domain-targeted sgRNAs that satisfied the following criteria. We only considered sgRNAs that have no perfect match off-target sites in the genome, and they must have guide_c scores $\geq 50$ and guide_nc scores $\geq 10$. Unlike the exon sgRNA library, we did not impose a potency score cut-off for the domain sgRNA library. We chose six sgRNAs with the highest potency score without considering from which domains they came or from which constitutive exon they came from. Other selection criteria were the same as the exon library above. Any sgRNAs that were already selected for the exon sgRNA library were excluded from the domain sgRNA library.

To curate a set of PCR primers that could work with the human and mouse sgRNAs for library construction, we first generated 1000 random GC-balanced 24-mers. These primer sequences were filtered to remove those with repetitive sequences, those with BsmBI sites, those with significant sequence similarity with any members of the sgRNA database, those with extreme $T_m$ and those with intramolecular hairpin propensity. In total, we curated 59 primers and 40 primers that are compatible with the human and mouse CRISPR libraries, respectively. These will enable the parallel retrieval of 29 and 20 sub-pools of human and mouse CRISPR library from a single oligo array, respectively.

To compare sgRNA identify among the published genome-wide human and mouse CRISPR KO libraries, we first removed any duplication sgRNA sequences within each library and carried out pair-wise gene and sgRNA overlap analysis among them. Because each library contains a slightly different set of genes, only sgRNAs from the shared genes were used for the overlap analysis. The sgRNA oligo design for array synthesis was illustrated in Supplementary Figure S2. Each oligo contains a 5′ and a 3′ PCR primer sequence. Inside the primer sequences were adaptor sequences containing BsmBI sites that are compatible for cloning into the LentiGuide-puro and the LentiCRISPR v2 expression vector (12). Inside the BsmBI adaptor sequences was the sgRNA 20-mer sequence. For parallel sub-pool retrieval, the sgRNAs were divided into groups of ~1000 sgRNAs per group and each group was assigned a unique primer pair chosen from Supplementary Table S3. For each array experiment, 12.4k sgRNA oligos were synthesized on a single array (CustomArray Inc.) and
the oligos were cleaved from the array as a single master oligo pool.

Individual groups of sgRNAs were PCR recovered from the master oligo pool using their corresponding primer pairs. The PCR components were as follows. Chip oligos 10 ng, 5′ primer 2 μM, 3′ primer 2 μM, Accuprime Pfx supermix (Life Technologies) 45 μl, total reaction volume 10 μl. The PCR condition was as follows. Hot start: 95°C for 5 min; PCR reaction: 95°C (15 s) – 55°C (15 s) – 72°C (15 s) cycles with 15 s per step for 15 cycles; final extension: 72°C for 2 min. PCR product was gel extracted (Qiagen) and quantified using PicoGreen (Life Technologies). To clone PCR products into the lentiviral sgRNA expression vector LentiGuide-puro (12), the following Golden Gate reaction mixture was used. PCR Amplicon of sgRNA insert 4.3 ng, linearized LentiGuide-puro plasmid 150 ng, Esp3I 10 U (ThermoFisher), T7 DNA ligase 3000 U (New England Biolabs), ATP 10 mM, total reaction volume was 20 μl. Esp3I is an isoschizomer of BsmBI that has optimal activity at 37°C. The reaction condition was as follows. Pre-incubation: incubate the reaction mixture without T7 ligase at 37°C for 2 h; Golden Gate reaction: Add T7 ligase and incubate the mixture with 20°C (20 min) – 37°C (5 min) cycle for 10 cycles. For bacterial transformation, ~30 ng of the Golden Gate reaction mixture was electroporated into 20 μl of ElectroMAX Stbl4 competent bacteria per manufacturer’s instructions (Life Technologies). The electroporated bacterial was recovered in 500 μl of SPC media at 30°C for 90 min. A small amount of the bacterial culture was plated on 10 cm LB-agarose plates with carbenicillin in serial dilutions to estimate transformation efficiency. The rest of the bacterial culture was plated out on 24 cm square LB-agarose plates with carbenicillin and incubated at 30°C for ~36 h. Bacteria were scraped off the plate and transferred to 100 ml of pre-warmed liquid LB media containing carbenicillin. After 2 h of growth at 30°C with shaking, plasmids were extracted as maxi-preps (Qiagen). A detailed, step-by-step cloning protocol is included in the Supplemental Materials section.

To verify the cloning efficiency of sgRNAs, LentiGuide-puro plasmid sub-pools were combined and the sgRNA PCR was recovered using the following PCR primers. Forward primer with Illumina adaptor: aatgtacagacacgagatctgatcctgagctttatcttgga aaggacgaaacaga, reverse primer with Illumina adaptor: caaagcagaagacggcatacgagatnntnntgtactgtgacgctacagc tgaatctgcttgatctttttgtgatgtttgtgatgtagtaactatttcccc (where ‘n’ denotes Illumina indexing nucleotides).

The PCR mixture was as follows. Plasmid templates 100 ng, forward and reverse primers 0.2 μM each, hot start Taq 5 U (Takara), dNTP 250 μM each, total reaction volume 50 μl. The PCR condition was as follows. Hot start step: 95°C 5 min; PCR 1: 95°C (15 s) – 57°C (30 s) – 72°C (30 s) for 5 cycles; PCR 2: 95°C (15 s) – 62°C (30 s) – 72°C (30 s) for 5 cycles; final extension: 72°C for 10 min. PCR product was gel purified (Qiagen) and sequenced on the MiSeq platform (Illumina) and reads with perfect match to sgRNA 20-mers were counted. A detailed, step-by-step plasmid sequencing protocol is included in the Supplemental Materials section.
Figure 1. Human CRISPR library characteristics. (A) Potency score distribution of the human (Hs) exon library. The potency score cut-off was 0.2. (B) Guide score distribution of the human exon library. Library sgRNAs were ranked separately for their guide score for coding regions (guide_c) and guide score for non-coding regions (guide_nc). The cut-off for guide_c score was 50 and for guide_nc was 10. (C) Distribution of sgRNA counts for all genes in the human exon library. (D) Potency score distribution of the human domain library (no cut-off was imposed for this library). (E) Guide score distribution of the human domain library. Library sgRNAs were ranked separately for their guide_c scores (cut-off at 50) and guide_nc scores (cut-off at 10). (F) Distribution of sgRNA counts for all genes in the human domain library. (G) Distribution of sgRNA counts for all genes in the combined human exon and domain library. (H) Distribution of off-target site counts for sgRNAs in the combined human exon and domain library. C1, C2 and C3 represent off-target sites in coding regions with up to one, two or three mis-matches, respectively.
library 80.1% genes have six sgRNAs /gene, and 85.6% of the genes have at least four sgRNAs /gene (Figure 1C).

Recent studies suggest that sgRNAs targeting conserved domain regions of a protein could be particularly effective at disrupting the protein’s function (5,17). We therefore further curated domain-targeted sgRNAs from our database to compile a library that we termed the human domain library. To do so, we first used the Pfam database to map regions in a human protein that belong to conserved domains. We then back-mapped the domains’ amino acid sequences to their corresponding exon regions in the genome and only selected sgRNAs that were located within these exon regions. We selected up to six sgRNAs per gene that target domains with the highest potency scores. Because 51% of sgRNAs in the exon library happened to map to protein domains fortuitously, any sgRNAs that were already present in the exon library were not selected again into the domain library to avoid sgRNA redundancy between the two libraries. The resulting domain library consists of 94,057 sgRNAs targeting 16,806 human genes. Among these sgRNAs, 84.8%, 59.6% and 28.5% have potency scores that are ≥0.2, ≥0.4 and ≥0.6, respectively (Figure 1D); and all sgRNAs have guideC scores ≥50 and guideNC scores ≥10 (Figure 1E). In the domain library, 86.2% of genes have six sgRNAs /gene and 92.0% of genes have at least four sgRNAs /gene (Figure 1F).

Because we designed the exon and domain libraries to have no overlapping sgRNA sequences, together the combined human exon and domain library consists of 195,007 sgRNAs targeting 18,236 genes. In the combined library, 81.5% of the genes having ≥10 sgRNAs /gene and 93.9% of the genes having ≥6 sgRNAs /gene, and a total of 5,968 Pfam domains were targeted (Figure 1G). A complete list of the human exon library and domain library sgRNAs is shown in Supplementary Table S1.

One of our goals in designing these CRISPR libraries is to balance the potency and off-target effects of a sgRNA. None of the sgRNAs in the library were allowed to have any perfectly matched off-target sites elsewhere in the genome. The majority of the sgRNAs also have very few predicted off-target sites in coding regions: in the combined mouse exon and domain libraries, 97.5%, 91.2% and 53.4% of sgRNAs have zero off-target sites in coding regions with up to one, two and three mismatches, respectively; and 99.9%, 99.5% and 92.8% of sgRNAs have ≤2 off-target sites in coding regions with up to one, two and three mismatches, respectively (Figure 2H).

Several genome-wide CRISPR KO libraries have been generated using distinct guide RNA picking algorithms. We therefore compared the sgRNA composition of our human libraries with six published human CRISPR KO libraries: Zhang human GeCKOv2 KO library (12), Sabatini human KO library (18), Broad human Brnello KO library (16), Toronto human KO library (8), Wu human KO library (19), and Yusa human KO library (20). Each of these libraries target most of the protein coding genes in the human genome, although they vary in library coverage between ~4 sgRNAs /gene to ~11 sgRNAs /gene (Supplementary Table S4A). To illustrate the difference among these libraries, we mapped sgRNAs from each library to two representative genes. We chose the kinase RAF1, which has a single splice variant and the bromodomain protein BRD4, which has two splice variants that differ in their C-terminus (Figure 3). For each gene, some sgRNA sequences are present in multiple libraries, although the majority of sgRNAs are unique to each library. At the library level, the sgRNA overlap (i.e. number of identical sgRNAs) between any two libraries were relatively small, at only 5–20% of the sgRNAs (Supplementary Table S4B). This is not entirely surprising because each library only represents <5% of all 4.3 × 106 possible sgRNA sequences. Hence, minor differences in selection criteria could result in different sgRNAs being selected. Similarly, we carried out sgRNA overlap analysis with three published mouse CRISPR KO libraries: Zhang mouse GeCKOv2 KO library (12), Broad mouse Kie KO library (16), and Yusa mouse KO library (20). We observed similar degree of sgRNA overlap (Supplementary Tables S5A and S5B).

Parallel retrieval of library oligos for custom sgRNA pool construction

Pooled CRISPR library is a powerful tool for functional genomics screens. Typically, to construct a pooled CRISPR library, sgRNA sequences together with flanking PCR primers are synthesized on a high density oligo array, cleaved off, and PCR cloned into expression vectors as a mixture (3,4). Because oligo arrays have a pre-determined capacity format with a fixed number of oligo features (e.g. 12k, 90k or 1000k features per array), the size of a pooled CRISPR library often has to conform to the feature number on the oligo array. On the other hand, if smaller pools are desired, synthesizing each pool on a dedicated oligo array becomes costly as each oligo sequence will need to be duplicated to occupy multiple features on the array until all the features are filled. To circumvent this problem and enable the economical construction of CRISPR library pools of any size, we developed a parallel retrieval method to allow
Figure 2. Mouse CRISPR library characteristics. (A) Potency score distribution of the mouse (Mm) exon library. The potency score cut-off was 0.2. (B) Guide score distribution of the mouse exon library. Library sgRNAs were ranked separately for their guide score for coding regions (guide_c) and guide score for non-coding regions (guide_nc). The cut-off for guide_c score was 50 and for guide_nc was 10. (C) Distribution of sgRNA counts for all genes in the mouse exon library. (D) Potency score distribution of the mouse domain library (no cut-off was imposed for this library). (E) Guide score distribution of the mouse domain library. Library sgRNAs were ranked separately for their guide_c scores (cut-off at 50) and guide_nc scores (cut-off at 10). (F) Distribution of sgRNA counts for all genes in the mouse domain library. (G) Distribution of sgRNA counts for all genes in the combined mouse exon and domain library. (H) Distribution of off-target site counts for sgRNAs in the combined mouse exon and domain library. C1, C2 and C3 represent off-target sites in coding regions with up to one, two or three mismatches, respectively.
the construction of multiple CRISPR library pools from a single, high-capacity oligo array.

We designed our CRISPR library sequences for array synthesis as follows. Each sgRNA 20-mer was flanked by primer sequences and BsmBI sites for subsequent PCR cloning into expression vectors (Figure 4A). However, instead of using the same PCR primer pair for the entire array, we designed a set of PCR primers that can be used in multiplex format with our human and mouse sgRNA library sequences (Supplementary Table S3). These primers are GC-balanced and they do not have significant sequence similarity to library sgRNAs. Thus, the fixed number of oligo features on one array can be sub-divided into may sub-pools by using a different primer pair for each sub-pool (Figure 4B).

To demonstrate that sub-pools can be reliably retrieved separately and cloned into plasmid vectors from a master oligo pool generated by array synthesis, we synthesized 12,251 sgRNA sequences using a 12k oligo array. These sgRNAs target 2,144 human kinase, phosphatase and transcription factor genes. The sgRNAs were bioinformatically divided into 13 sub-pools of ~1,000 sgRNAs/sub-pool that map to ~170 genes/sub-pool. Each sub-pool was given a unique pair of flanking primers for their selective retrieval from the master oligo pool. We used 13 parallel PCR reactions with low cycle numbers to retrieve each sub-pool separately from the master oligo pool, and we subsequently cloned each sub-pool into lentiviral expression plasmid using Golden Gate ligation (Supplementary Figure S2).

Illumina sequencing of the 13 plasmid pools showed that each sub-pool was retrieved from the master pool with comparable efficiency and consistency. Most of the designed sgRNA sequences were present in the plasmid pool after cloning, with only 0.2–1.2% of the sgRNAs failed to attain reads in the plasmid sub-pool at a sequencing coverage of ~150. More importantly, the relative representation of each sgRNA within each plasmid sub-pool was highly consistent, with few sgRNAs that were either under- or over-
Figure 4. Human CRISPR library construction and sgRNA read distribution. (A) Schematics of sgRNA oligo design. The pre-designed sgRNA 20-mers are flanked by 5’- and 3’-PCR primers and BsmBI sites. (B) Schematics of parallel retrieval of oligo pools. Library oligos with different flanking primer sequences were synthesized together as a single 12k oligo pool. Specific primer pairs were used to PCR amplify and retrieve subsets of oligos from the master pool. The PCR amplicons were cloned into lentiviral vectors using Golden Gate assembly. (C) Distribution of normalized sgRNA read frequency from 13 distinct plasmid sub-pools generated from a single oligo pool. (D) Percentage of sgRNA reads that are within an 8-fold range from 24 plasmid sub-pools generated from two master oligo pools.
represented in a given sub-pool (Figure 4C). We repeated this strategy with a second oligo array of 11,282 sgRNAs designed to target 1970 human metabolic enzyme genes. We bioinformatically divided the sgRNAs into 11 sub-pools for parallel PCR retrieval. Sequencing analysis of these two independent experiments, which involved the generation of 24 sub-pools from two independent oligo arrays, showed that our parallel retrieval strategy was highly consistent. In all 24 plasmid sub-pools, the within-pool raw sequencing reads for most sgRNAs fall in a narrow range (Supplementary Figure S3A and S3B), and > 90% of the sgRNAs were within an 8-fold range of representation to each other (Figure 4D).

We analyzed whether sgRNA representation in each plasmid sub-pool is influenced by the cloning representation during the bacterial transformation stage. The cloning representation (i.e. the fold-coverage of total number of bacterial transformants over pool complexity) for plasmid sub-pools varied between 62 and 410 in our experiments (Supplementary Figure S4). The variance of sgRNA sequence reads within each plasmid sub-pool was consistently between 0.05 and 0.15 for all but one sub-pool. Interestingly, the variance remained largely uncorrelated with the cloning representation of these sub-pools (Supplementary Figure S4). Thus, a cloning representation as low as ~60 bacterial transformants per sgRNA might be sufficient to yield a highly uniform plasmid library in this context. Taken together, these data indicate that our parallel retrieval method is highly robust at generating multiple customized CRISPR library pools using a single oligo pool from array synthesis.

DISCUSSION

In this study, we presented bioinformatics and technical resources that enable individual academic labs to create cost-effective custom CRISPR libraries for functional genomics screening. We generated SpCas9-compatible sgRNA databases that target most human and mouse protein coding genes. In designing the sgRNA sequences, we took several factors into consideration to balance sgRNAs features and achieve an optimal library design. The exon library and the domain library each has a coverage of ~5.5 sgRNAs/gene and the combined library has a coverage of ~11 sgRNAs/gene on average. This degree library coverage should be sufficiently deep for most screening purposes (7,18,21). Depending on the application and gene coverage depth desired, some or all of the sgRNAs for each gene can be chosen as part of a custom CRISPR library. In curating sgRNA sequences, we took particular care to balance both potency and off-target properties of sgRNAs. The majority of sgRNAs in our libraries have good scores with both of these measures. To further maximize the likelihood that these sgRNAs will effectively inactivate their target genes, we took two curation approaches such that for a given gene, half of its sgRNAs will target multiple 5′ constitutive exons whereas the other half will target constitutive exons that encode conserved protein domains. The resulting human and mouse library, which constitute ~200,000 sgRNAs each, represent ~4.5% of the total candidate sgRNA sequences that are in the ‘NGG’ PAM context. Thus, we believe these curated sgRNA sequences will be useful for both custom library construction and for knocking out the expression of individual genes.

Prior to our library design, a number of genome-wide CRISPR KO libraries have been designed for the human and mouse genome (8,12,16,18–20). The relatively small sgRNA sequence overlap among these libraries, including ours, might be attributable to the fact that each library uses a distinct set of rules to pick <5% of all possible sgRNAs. Because sgRNA selection scores with any given algorithm is a near-continuous variable, a small difference in selection criteria could result in a fairly large difference in the sequence of the selected sgRNAs. Our library is the only one that balances sgRNA potency, off-target effects and domain targeting in a single design.

Our human and mouse CRISPR library design serve as a starting point for the construction of custom CRISPR libraries of virtually any size. Although a number of genome-wide CRISPR libraries have been made available (8,12,16,18–20), there is an increasing need among individual investigators to screen smaller, customized libraries that target a specific set of genes. This could be a functional category of genes such as kinases and metabolic enzymes, or a group of genes that represent either an over-expression signature or a mutational signature. Furthermore, with CRISPR screen moving towards in vivo applications (22), smaller libraries are likely to have better signal-to-noise ratio than genome-wide libraries in these settings because cell number, and hence library representation, could become a significant limiting factor in animal models (23,24). Our strategy thus enables the easy and cost-effective construction of sub-genome sized libraries to meet these special screening needs. We demonstrated that a dozen custom CRISPR library plasmid pools can be generated with a single master oligo pool from array synthesis. Although we chose to generate library pools of ~1,000 sgRNAs, our method is scalable and should enable the generation of sub-genome scale CRISPR libraries of virtually any size.

SUPPLEMENTARY DATA

Supplementary Data are available at NAR Online.

FUNDING

US National Cancer Institute Intramural Grant [ZIA BC 011304 to J.L.]. Funding for open access charge: NIH Intramural Funding.

Conflict of interest statement. None declared.

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