High-throughput interrogation of programmed ribosomal frameshifting in human cells

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Summary

Programmed ribosomal frameshifting is the controlled slippage of the translating ribosome to an alternative frame. This tightly regulated process is widely employed by human viruses such as HIV and SARS coronavirus and is critical for their life cycle and virulence. It is also utilized from yeast to human to implement a feedback control mechanism to regulate polyamine levels. Here, we developed a high-throughput, fluorescence-based approach to assess the frameshifting potential of a sequence. We designed and tested >12,000 sequences based on 15 viral and human frameshifting events, allowing us to elucidate the rules governing ribosomal frameshifting in a systematic way and to discover novel regulatory inputs based on amino acid properties and tRNA availability. We assessed the natural variation in HIV gag-pol frameshifting rates by testing >500 clinical isolates and identified subtype-specific differences as well as associations between viral load in patients and the optimality of gag-pol frameshifting rates. We further devised computational models that accurately predict frameshifting potential (up to auROC=0.93) and frameshifting rates (up to Pearson r=0.81) of novel variants, including subtle differences between HIV clinical isolates (r=0.60). Taken together, this systematic approach can contribute to the development of antiviral agents acting on programmed ribosomal frameshifting.

Introduction

Programmed ribosomal frameshifting (PRF), i.e. controlled slippage of the ribosome, is a mechanism by which two proteins with alternative C termini can be generated from the same mRNA. It allows for an expansion of the proteome but also constitutes an additional regulatory layer to fine-tune gene expression (Advani and Dinman, 2016; Dinman, 2012; Ketteler, 2012). This mechanism is widespread and indispensable in viruses, which often utilize controlled slippage of the ribosome to an alternative frame to regulate the production of key enzymes, such as in the case of the gag-pol frameshift in HIV and other retroviruses. Utilizing alternative frames increases the amount of genetic information that can be encoded in a given sequence and constitutes another level of gene regulation, which might explain why especially RNA viruses – with their compact genomes and lack of regulation of their genes on the transcriptional level - utilize PRF for crucial regulatory switches (Firth and Brierley, 2012). The importance of maintaining the stoichiometry between structural proteins encoded by the *gag* gene and enzymes encoded by the *pol* gene for viral replicative success make the gag-pol frameshifting event a promising antiviral drug target (Brakier-Gingras et al., 2012; Hung et al., 1998).

Cases of functionally important programmed frameshifting have also been discovered in humans (Belew et al., 2014; Clark et al., 2007; Ivanov et al., 1998; Matsufuji et al., 1995; Rom and Kahana, 1994; Tosaka et al., 2000; Wills et al., 2006). Discovering PRF events in the human genome has been hampered by the limited amenability of PRF to proteome-wide methods due to the generally low abundance of the frameshifted protein relative to the canonical protein or inherent instability of the frameshifting product. Most human cases known to date were found serendipitously or through homologous genes. A striking example of regulatory conservation is the case of ornithine decarboxylase antizyme (OAZ), which is produced through polyamine-stimulated +1 frameshifting and inhibits polyamine production (Matsufuji et al., 1995; Rom and Kahana, 1994). This negative feedback loop is used by virtually all organisms from yeast to humans to control polyamine levels (Ivanov et al., 2000), attesting to the evolutionary success of PRF as a regulatory mechanism.

Fameshifting is generally believed to happen at defined positions consisting of a slippery sequence and a downstream roadblock, most commonly a stable secondary RNA structure like a pseudoknot or an extensive stem-loop structure (Caliskan et al., 2015; Dinman, 2012). Most presently known -1 slippery sites follow the pattern X XXY YYZ, with the shift happening from codons XXY and YYZ to XXX and YYY. Some known slippery sites in -1 PRF show divergence from this pattern, e.g. in the NSA2 gene of west nile virus with the slippery site sequence UCCUUUU (Firth and Atkins, 2009; Melian et al., 2010). At +1 frameshifting sites like OAZ ribosomal translocation happens at a very distinct motif (e.g. UCCUGA). Many case studies have contributed to an understanding of the molecular events happening during frameshifting (e.g. Belew et al., 2014; Caliskan et al., 2015; Kurian et al., 2011; Ritchie et al., 2017; Tholstrup et al., 2012), but the general, overarching regulatory principles that determine if and to what extent PRF occurs remain largely unknown.

Here, we developed a massively parallel reporter assay that allows for high-throughput quantification

of ribosomal frameshifting in human cells. We designed and tested 13,390 oligonucleotides containing

rationally designed variants of known frameshifting signals. We systematically deciphered

determinants of PRF efficiency across frameshifting events and assayed natural variation in HIV gag-

pol frameshifting, providing the first systematic large-scale investigation of ribosomal frameshifting.

Results

A massively parallel reporter assay for programmed ribosomal frameshifting

To assay PRF in a comprehensive manner we designed a synthetic oligonucleotide library containing

12809 variants with systematic sequence manipulations of previously reported PRF sites

(Supplementary Data 1, "main set") and 581 sequences of gag-pol frameshifting sites in HIV clinical

isolates (collated from http://www.hiv.lanl.gov/, see Methods for selection criteria of HIV variants) (Fig

1A).

The oligonucleotides comprising library-specific common primers, a unique barcode and a 162 nt long

variable region containing a potential frameshifting site (Fig 1A, bottom) were synthesized on an

Agilent microarray, amplified and cloned in between *mcherry* and *gfp* coding sequences such that the

gfp coding frame was shifted by +1 or -1 relative to the original, mCherry-encoding frame (Fig 1B). If

the corresponding frameshift occurs, GFP is made into protein, and GFP fluorescence intensity thus

serves as a measure for frameshifting efficiency. We introduced this construct in the AAVS1 locus in

the human K562 cell line using zinc finger nucleases, such that every cell has one frameshifting

reporter construct from the library and all the variants have the same genomic environment (Methods).

For both, the -1 and +1 reporter libraries, we selected a narrow mCherry-positive population to

minimize effects coming from the influence of the variable region on overall expression levels (Fig

S1AB, Methods). We sorted this population corresponding to a single integration of the reporter

transgene using flow cytometry into 16 bins according to their GFP fluorescence intensity (Fig S1CD)

and sequenced genomic DNA from all the bins to determine the distribution of each variant across

bins. We previously demonstrated that similar approaches are highly accurate and reproducible (Mikl

et al., 2019; Vainberg Slutskin et al., 2018; Weingarten-Gabbay et al., 2016, 2019), and the consistent

bin profiles for many barcode control groups with identical variable region (Fig S1E, Methods)

corroborate the low technical noise we are able to achieve.

The distribution of mean GFP expression of all library variants showed a clear peak corresponding to

background green fluorescence (Fig S2AB). We set the lowest mean GFP fluorescence we observed

to 0% and the highest – coming from variants with GFP in frame with mCherry – to 100%. Accordingly,

we assigned a percentage to every variant that passed filtering for read number, bin profile and

Figure 1. A massively parallel reporter assay for programmed ribosomal frameshifting.

A. Schematics of the library design; for: forward primer, rev: reverse primer, BC: barcode. B. Outline of the experimental pipeline; the plots in the bottom constitute representative bin profiles for variants with no (left), low (middle) or intermediate (right) GFP fluorescence. CD. Mean and 95% CI of -1 (C) and +1 (D) PRF signal (% GFP fluorescence) for barcode control groups corresponding to the indicated wild-type PRF sequences (blue), carrying single nucleotide deletions (green) or having a stop codon inserted upstream of the frameshift sequence (red); n between 1 and 32; the shaded area denotes the range of background fluorescence; asterisks denote significant differences between the wild-type and the other groups (single nucleotide deletion and stop-control) combined (Mann-Whitney-U test). E. Comparison of -1 and +1 PRF reporter readout (% GFP fluorescence) for previously reported PRF sites (cf. Fig 1CD, Fig S3AC).

expression levels (by gating for a narrow range of mCherry fluorescence to minimize effects coming

from the influence of the variable region on overall expression levels, Methods). This percentage value

does not denote the precise rate of frameshifting events, but gives us a meaningful measure of

frameshifting efficiencies across PRF events. We thereby obtained measurements for 8972 (67%)

and 5922 (44%) sequences cloned into the -1 PRF and +1 PRF reporter, respectively.

To test the possibility that the signal in our assay comes from internal translation initiation downstream
of mCherry (which might lead to GFP expression independently of ribosomal frameshifting in between
the *mCherry* and *gfp* coding region) we cloned our library in a modified reporter construct containing
a stop codon after the mCherry coding region (Fig S2C top). GFP fluorescence coming from a
frameshifting event would in this case be lost. We subjected this library to the same experimental
pipeline and obtained green fluorescence measurements for 10561 variants (79% of the library, 82%
and 83% of variants for which we obtained -1 and +1 PRF readouts, resp.; Fig S2C). Green
fluorescence was almost exclusively at background levels, showing that the overwhelming majority of
tested sequences cannot drive GFP expression on their own.

Another potential source of GFP expression is the removal of cryptic introns in the variable region that
would lead to *gfp* being in frame with *mCherry* in the resulting mature mRNA. To test if this is a source
of false positives in our assay, we performed RNA sequencing on the whole library, covering the entire
variable region and its surroundings (Fig S2D). We mapped the full length reads to the library variants
to detect splicing events and obtained an RNA readout (at least 100 reads mapped) for 10666 variants
(80% of the library, 94% of the variants with readout in the -1 frame and 96% of the variants with
readout in the +1 frame). Even without any filtering, only 2.95% of the variants showed gaps in the
mapping of at least 20 nt which could represent events of cryptic splicing (Fig S2EF; 0.95% introducing
a shift to frame +1 and 0.64% to frame -1). Missing stretches can also come from synthesis or cloning
errors (and are therefore present and filtered out on the DNA level. Considering only those cases that
had a potential (degenerate, cryptic) donor or acceptor splice site in the area of the gap further
reduced the fraction of variants for which we cannot rule out a relevant splicing event to 36 (0.55%;
Fig S2EF, red; 0.22% introducing a shift to frame +1 and 0.14% to frame -1). Importantly, even for
these cases there was no correlation between the fraction of potentially spliced reads and GFP
expression (Pearson r=0.01, p=0.21 for -1 PRF and r=0.02, p=0.09 for +1 PRF), indicating that cryptic
splicing events do not constitute a common source for false positive signal in our assay.

During oligonucleotide synthesis errors can occur that can introduce a point mutation or an
insertion/deletion. For assaying ribosomal frameshifting, especially the latter type would be
detrimental, and therefore we perform full-length DNA sequencing in our assay and consider only
reads that exhibit perfect alignment along the length of the variable region covered by sequencing
reads (approximately 20 nt in the center of the variable region cannot be covered due to the read
length (150 nt from either side)). These synthesis errors, however, also further enrich our synthetic
library, as they add additional variants to the collection of sequences tested. Especially single
nucleotide deletions or insertions constitute a great internal control, as they differ only in one position
from the designed variant. Cases where we obtained enough reads (after filtering, Methods) for the
intended variant (Fig S2G top, "wild-type"), a single nucleotide deletion abolishing the signal coming
from -1 PRF (Fig S2G middle) and a single nucleotide insertion leading to *gfp* being in frame (Fig S2G

bottom) constitute a proof of principle. They also provide additional evidence that the signal that we
measure stems from a (reading frame-sensitive) ribosomal frameshifting event. Looking at the
distribution of all single nucleotide deletion and insertion variants from the library reveals the expected
peaks at minimal and maximal GFP expression (Fig Fig S2HI), with a small group of variants being
misclassified due to the fact that our reads cover only approximately 90% of the variable region.

Based on the distribution of all library variants with a stop codon cassette integrated immediately after
the *mcherry* coding sequence we assigned a noise threshold (95th percentile of the stop-vector library
distribution, corresponding to 1.3% of the maximal GFP fluorescence; Fig S2C) imposed by
autofluorescence of the cells. The contribution of autofluorescence coming from cellular components
and metabolites such as NADH and flavins is negligible when methods for signal amplification are
used (e.g. in Luciferase assays commonly used for the quantification of PRF). It does, however,
constitute a limitation for detection of GFP-fusion proteins made from a single gene copy and present
at low copy number in the cell owing to the fact that the PRF event responsible for its expression
happens in only a few percent of translation events. This entails a higher detection limit of our assay
compared to for example Luciferase assays, which is a necessary compromise to enable high-
throughput testing.

Based on the distribution of GFP intensities of variants with a mutation leading to GFP being in frame
with mCherry (Fig S2I) we also excluded variants exhibiting a dominant peak with mean GFP
fluorescence above 2^{12} (25% of maximal GFP fluorescence). This value corresponds to the 5th
percentile of the distribution of variants with a single nucleotide insertion (above threshold). We
thereby aim to rule out biases stemming from DNA frameshifts occurring as synthesis or cloning errors
that our mapping strategy might have missed (Methods).

We tested previously reported frameshifting sites (with or without experimental validation,
Supplementary Data 1, based on Moon et al., 2007 and our own survey of the literature) with multiple
different barcodes in our assay and could reproducibly detect GFP fluorescence in the expected frame
for many of the previously reported PRF events (Fig 1CD, Fig S3A-D). For example, the previously
reported PRF sites in HIV-1 (Jacks et al., 1988a), SARS coronavirus (Thiel et al., 2003), HERV-K10
(Ono et al., 1986; Wang et al., 2002), simian retrovirus (ten Dam et al., 1994), SIVmac239
(Marcheschi et al., 2007; Pollom et al., 2013), human T cell lymphotropic virus (Nam et al., 1993) and
OAZ (Matsufuji et al., 1995; Rom and Kahana, 1994) yielded a fluorescent signal between 1.4% and
18% of maximal GFP fluorescence detected in our assay. These values do not necessarily denote
the percentage of translation events, in which frameshifting happens, in absolute terms, and are not
directly comparable with percentages reported in other studies.

PRF events tested in our assay typically gave a signal over threshold only in either the -1 or +1 PRF
reporter, but not in both (Fig 1E). In cases where the native sequence contained stop codons in any

frame downstream of the frameshifting site, we also tested variants with all stop codons mutated
(Methods). These typically either did not affect or led to a substantially reduced frameshifting signal
(Fig S4A). Moving the slippery site by one to three nucleotides typically abolished the signal, e.g. in
the case of SARS (Fig S4B, Methods).

Some previously reported PRF events failed to yield a signal above threshold in our assay. Several
reasons like sensitivity to the larger context of the reporter construct or a general effect of the tested
sequence on expression could be possible explanations. In the case of the PRF site in Rous sarcoma
virus (Jacks and Varmus, 1985; Fig S4C), a construct containing 124 nt of sequence downstream to
the position of frameshifting was able to induce frameshifting, but shorter versions were not (Jacks et
al., 1988b). A sequence motif (termed PK4) at the end of this downstream region was shown to be
important for maintaining wild-type frameshifting ability, potentially by creating a pseudoknot-like
structure through basepairing with a complementary sequence in the loop of the main stem-loop
structure (Marczinke et al., 1998; Nikolic et al., 2012). Our construct (including 120 nt of downstream
sequence) does contain the PK4 motif, but it is located at the very 3' end of the native sequence
included in the reporter construct. The sequence context of PK4 was shown to be important for proper
folding (Marczinke et al., 1998), and therefore the specific context created by our reporter construct
might explain the lack of signal.

We also failed to detect a frameshifting signal for CCR5 (Belew et al., 2014; Fig S4C). In addition to
potential sensitivity to the larger context of the reporter construct and a specific miRNA that is required
for maximal frameshifting efficiency (Belew et al., 2014), the lack of signal might be due to a potential
effect of the PRF site on overall expression levels on the RNA or protein level, as has been suggested
for cellular frameshifting sites in general (Advani and Dinman, 2016). We select for a narrow range of
mCherry levels and therefore expression levels (Methods, Fig S1AB), which leads to filtering out
variants that strongly affect overall expression levels. In the case of CCR5 only 29% of variants with
the wild-type sequence passed filtering (Fig S4D; as opposed to 45-97% for other PRF events), which
could hint at an effect of the included sequence on expression at the RNA or protein level. Estimating
steady state RNA levels did not show systematic differences between PRF events yielding a signal
above threshold in our assay and those that failed to do so (Fig S4E), but indicated that specifically
RNA from variants containing the PRRSV PRF site was present at reduced levels.

Some of the frameshifting signals tested showed variability between variants with identical variants,
but different barcodes (Fig S4C, right). In the case of west nile virus, the majority of sequences tested
resulted in a fluorescent signal above threshold, leading to it being classified as "frameshifting" in our
assay (p=0.027 for the difference between wild-type and stop control; Fig 1C); in the case of PEG10
(Clark et al., 2007; Manktelow et al., 2005) the majority of sequences failed to yield a signal (after
filtering). This points at an instability of the signal in this context, and therefore our assay might not be

suitable for investigating these specific events. PEG10 is thought to have a relatively extensive
downstream region necessary for frameshifting (~80 nt according to estimates from in vitro testing of
truncated regions (Manktelow et al., 2005)). Our library includes 120 nt of downstream sequence, but
the more extensive the downstream sequence, the higher the sensibility to the context and
experimental conditions might be. This is in line with our observation for SARS and HERV-K10, for
example (see below). In addition, the required context might be larger in the *in vivo* situation compared
to *in vitro*.

This instability of the signal might be affected by small sequence changes: For some library
sequences measurements of a large number of variants with sequence alterations introduced during
synthesis and cloning are available. In the case of PEG10, all the synthesis variants containing a
specific mutation (C-to-G at position 7 after the slippery site) show a -1 PRF signal above background
(Fig S4F). Likewise, a fraction of the designed sequence alterations introduced in the PEG10 context
triggered frameshifting (Fig S4G), especially mutations in the upstream region (Fig S4G left). This
shows that minimal sequence changes can have far-reaching (and consistent) consequences for the
potential of a sequence to yield a frameshifting signal in the context of our high-throughput assay and
probably in any experimental setup.

Slippery sites are optimized for the respective frameshifting event

While previous investigations of frameshifting focused on individual examples, we aimed to identify
commonalities and differences between frameshifting sites. To this end we introduced systematic
sequence alterations (such as point mutations, synonymous substitutions, altering and mimicking
endogenous secondary structure, etc.; Fig 1A; Methods) in all contexts and measured the effect on
frameshifting in large scale using our FACS based assay. Individual PRF events exhibited
characteristic sensitivities to sequence alterations (Fig 2A). The HIV-1 gag-pol PRF site (Jacks et al.,
1988a) was particularly sensitive to changes in or immediately around the slippery site (any sequence
change in this region led to a median PRF signal of 25% of the wild-type), while changes in the
upstream or downstream region led to PRF signal of – on average – 97% (upstream) and 70%
(downstream; Fig 2A, Fig S5A) of the corresponding wild-type. A similar pattern could be observed
for SIVmac239. The HERV-K10 (Ono et al., 1986; Wang et al., 2002), SARS (Thiel et al., 2003) and
SRV1 (ten Dam et al., 1994) PRF sites were highly sensitive to changes in both the slippery site and
the downstream region and in the case of HERV-K10 and SRV1 also showed a significant (p=0.0004
and 0.0023, resp.; Wilcoxon signed-rank test) reduction following changes in the upstream region (Fig
2A, Fig S5B). It is noteworthy that the clustering of effects (although done on percent of wild-type PRF

Figure 2. Slippery sites are optimized for the respective frameshifting event.
A. Clustered heat map of median percent wild-type frameshifting of variants in which the indicated region (upstream, slippery site (including the preceding codon) and downstream) differs from the native one (n=37-325). B. The median percent wild-type PRF signal of variants in which the indicated position is changed is plotted for the entire length of the variable region; gray box: slippery site. C. Percent wild-type -1 PRF signal for variants with a point mutation only in the indicated position of the slippery site (n=6-22). D. Heatmap showing median percent GFP fluorescence conferred by replacing the native slippery site in -1 PRF sites tested positive in the assay (Fig 1C) with the indicated combination of slippery site elements. E. Distribution of percent wild-type GFP fluorescence of variants in which the slippery site was replaced by any possible combination of bases following the pattern XXXYYYZ (n=19-80). F. Distribution of normalized reads across GFP (+1 frame) expression bins for variants of the OAZ1 frameshifting site, in which either the first (right) or the second (left) codon is replaced; blue lines: wild-type; green and magenta lines: bin profile for the variant in which the native codon was replaced by the indicated one; grey lines: bin profiles for other codons at this position. G. Percent GFP fluorescence for variants in which the slippery site was replace with all possible combinations of the slippery site pattern found in west nile virus (XYYZZZZ) (n=16-56).

signal) resembles the wild-type PRF efficiencies (Fig 1C), with the most efficient sites (HERV-K10,

SARS and SRV1) showing the highest and broadest sensitivity to sequence alterations.

Comparing per position the median PRF signal for variants in which this position was changed reveals

a sensitivity profile around the PRF site (Fig 2B, Fig S5C), recapitulating these context specific

characteristics. Across all -1 PRF events tested positive in our assay, not only mutations at the

slippery site (marked in grey) and downstream of it, but also mutations immediately upstream of the

slippery site negatively affect PRF efficiency. Other -1 PRF events reported in the literature, but not

yielding a signal above background in our assay, tend to show substantially increased signal upon

mutations in the areas upstream (PEG10) and downstream (PEG10, CCR5, Rous sarcoma virus) of

the slippery site (Fig S5D). Like the specific examples of frameshifting PEG10 variants described

above (Fig S4EF), this points at a strong sensitivity of these PRF events to the context in which they

are tested.

Most single point mutations within the slippery sequence abrogated frameshifting and resulted in

green fluorescence at background levels, both for -1 and +1 frameshifting sites (Fig 2C, Fig S5E),

demonstrating that we indeed measure frameshifting at the expected site. Replacing the last position

in the canonical slippery site X XXY YYZ had – on average – a less severe effect on PRF efficiency,

due to the fact that it is the wobble position in the original frame and therefore has larger flexibility (Fig

2C). In high efficiency PRF sites also this position showed strong and PRF site-specific preferences:

In the case of SRV1 (GGGAAAC) replacing the Z position with other options than the endogenous C

led to a drastic decrease (U, A) or complete loss (G) of -1 PRF signal (Fig S4E), while in the case of

SARS (UUUAAAU) all other nucleotides apart from the endogenous U at the Z position almost

completely abolished the signal. To determine whether there is a universally optimal slippery site, we

replaced the slippery sequence in our set of -1 PRF events with all possible variations of the pattern

XXXYYYZ (Fig 2D, Fig S4F). This revealed preferences for specific combinations common in known

PRF sites (Fig 2D), like UUUUUUZ (HIV, SIVmac239) and XXXAAAC (HERV-K10, HTLV, PEG10,

SARS), but no universally optimal combination of bases that would lead to maximal frameshifting

efficiency and consequently GFP fluorescence across contexts (Fig S5F). Notably, PRF sites showed

a high degree of optimization when it comes to the identity of the bases, with any substitution (although

preserving the canonical XXXYYYZ pattern) typically leading to a more or less severe drop in PRF

signal (Fig 2E, Fig S5EF).

Replacing the first or second codon of the OAZ slippery site (UCC UGA) revealed a very limited

tolerance to sequence alterations, consistent with earlier results (Matsufuji et al., 1995). Replacing

the second codon (UGA), the only variants leading to a +1 PRF signal over background where the

other two stop codons (UAA and UAG; Fig 2F, right), reaching approximately 50% of wild-type

frameshifting rates, confirming that a translation termination signal is required for frameshifting (Rom

and Kahana, 1994). At the first codon position, UUU was the only codon leading to +1 PRF signal

similar to wild-type (Fig 2F, left), maybe due to the creation of a stretch of four Us and thereby

facilitating re-pairing of the tRNA in the +1 frame. Indeed, UUU is the second most widely used codon

found in that position in the OAZ frameshifting site in other species, e.g. certain funghi and nematodes

(Ivanov and Atkins, 2007).

Some reported -1 PRF sites deviate from the canonical slippery site pattern (XXXYYYZ). To test the

ability of the alternative patterns to induce frameshifting across sequence contexts we replaced the

slippery site in a set of PRF events with all combinations of nucleotides following the pattern found in

west nile virus (XYYZZZZ, Fig 2G; Melian et al., 2014; Moomau et al., 2016). In general, the ability of
alternative slippery site patterns to cause frameshifting was limited to their native sequence contexts.
In this context, however, several different nucleotide combinations following the respective pattern
were tolerated. The only exception to this rule was human T-lymphotropic virus (Fig 2G, endogenous
slippery site UUUAAAC), which gave a robust -1 PRF signal corresponding to wild-type levels across
many variants following the (west nile virus) pattern XYYZZZZ. In summary, our data suggest that in
most cases a slippery site not following the canonical pattern requires a specific sequence context in
order to trigger frameshifting, in accordance with earlier results comparing west nile virus strains with
canonical or non-canonical slippery sites (Moomau et al., 2016).

Downstream sequence and structure affect PRF efficiency in a context-dependent manner

The region downstream of the slippery site is thought to be crucial for PRF, typically because it creates
a roadblock for the translating ribosome by folding into a stable secondary structure like a hairpin or
a pseudoknot (Brierley and Pennell, 2001; Brierley et al., 1989; Parkin et al., 1992; Tu et al., 1992).
To get an estimate of the size of the relevant downstream region in different PRF events, we replaced
native regions with a constant sequence, leaving stretches of different length after the slippery site
unchanged (Fig 3A, top; Methods). While in the case of SARS almost the entire downstream region
included in our reporter construct (120 nt) was crucial for frameshifting, in the HIV and SIVmac239
PRF sites only a comparatively small region (<24 nt) was required to achieve PRF signal >50% of
wild type (Fig 3B). The shortness of the necessary downstream sequence in these cases was striking,
but in line with earlier evidence demonstrating that base pairing at the first 3-4 positions of the HIV
downstream stem-loop structure shows the strongest association with PRF efficiency (Mouzakis et
al., 2013).

To identify specific positions crucial for PRF we performed scanning mutagenesis of regions ranging
from 7 to 36 nt downstream of the slippery site (Fig 3A, bottom). This revealed large differences in
the fraction of relevant downstream positions (Fig 3C), ranging from 6% of single point mutations
reducing -1 PRF efficiency by more than half in the case of SIVmac239 to 53% in the case of SARS
(46% for +1 PRF at the OAZ1 site). The HIV gag-pol PRF site (among others) shows remarkable
resilience to point mutations (Fig 3C, Fig S6A), in line with the high degree of genetic variability in
HIV. Specific mutations affecting PRF typically disrupt base-pairing (e.g. positions 11, 12, 29 and 30
in HIV and positions 8-13 and 18-26 in SRV1; Fig S6A) and therefore change the (predicted)
secondary structure. In the case of HIV only mutations in the middle of the stem are detrimental,
suggesting that a reduced stem is still able to drive PRF at wild-type levels, and only forcing a
complete change in the structure affects frameshifting.

Figure 3. Downstream RNA structure affects PRF efficiency in a context-dependent manner.

A. Schematic of downstream sequence manipulations. B. Percent wild-type frameshifting rates of variants in which the native downstream region has been replaced by constant sequences, leaving the indicated number of nucleotides after the slippery site unchanged; n=1-3 per data point. C. Fraction of point mutations in the 40 nt downstream of the slippery site resulting in <50% of wild-type GFP fluorescence (green) and background fluorescence (<1.3% GFP fluorescence, blue), for -1 (top) and +1 (bottom) PRF events. D. The difference in mean % wild-type frameshifting between variants in which the indicated position is predicted to be paired vs unpaired along the variable region; gray box: slippery site. E. Schematic of downstream structure variations. F. Percent -1 GFP fluorescence of variants in which the downstream region is replaced by a synthetic sequence predicted to fold into the respective native secondary structure (using Vienna RNA (only stem-loop; light blue) or pKiss (including pseudoknots; dark blue; Methods)); the shaded area denotes the range of background fluorescence. G. % -1 GFP fluorescence of variants in which the downstream region is replaced by a synthetic sequence predicted to fold into the HIV (blue) or SRV1 (green) secondary structure (or variations thereof; Methods); the shaded area denotes the range of background fluorescence. H. % GFP fluorescence of variants in which the downstream region carried a single point mutation (left) or was replaced with sequences predicted to fold into variations of PRF inducing structures, plotted against the corresponding wild-type values; p-values (Mann-Whitney-U) for comparisons between groups of low (<4% GFP fluorescence) and high wild-type PRF (>4% GFP fluorescence). IJ. Clustered heatmap of Pearson correlation coefficients between percent -1 GFP fluorescence and minimum free energy of regions after the slippery site (computed using Vienna RNAfold on the 40 nt following the slippery site (I) or pKiss on the 120 nt following the slippery site (J)) for groups of variants of the indicated PRF sites; values denote the p-values of the correlations.

We tested if we can detect these mutational signatures of secondary structure also on the scale of the entire library and compared the median PRF signal between all variants of a PRF event, in which this position is predicted to be paired or unpaired (Fig 3D). Preferences for downstream positions to be paired or unpaired reveal the architecture of structural elements downstream of the PRF site. Groups of frameshifting events show remarkable concordance between these preferences (e.g. HIV and SARS, HERV-K10 and SRV1), but length and position of the optimal downstream secondary structures differ between groups (Fig 3D, upper vs. lower panel).

Replacing the endogenous downstream region with elements that have different primary sequence, but are predicted to fold into the same secondary structure (Fig 3E, top) abolished frameshifting in most cases, regardless of whether pseudoknots were included or not (Vienna RNA vs. pKiss to determine the structure that served as the input for antaRNA (Kleinkauf et al., 2015a, 2015b); Fig 3F). Introducing variants of the SRV1 and the HIV downstream structure (without preserving the original sequence, Fig 3E, bottom; Methods) had the ability to trigger frameshifting in some cases (Fig 3G), including ones where the wild-type sequence did not yield a frameshifting signal in the context of our assay (CCR5 and PEG10). Strikingly, events associated with higher wild-type frameshifting rates could not be "rescued" by introducing a secondary structure variant from a different PRF event (Fig 3H), indicating that more efficient frameshifting sites like SARS seem to be highly optimized, but less tolerant to changes in the type of downstream secondary structure.

Changes in secondary structure (minimum free energy) induced by point mutations downstream of the slippery site showed strong correlation with most, but not all frameshifting events (Fig 3I left column, Fig S7 left panels). Surprisingly, this correlation between predicted minimum free energy of secondary structure variants of the native downstream region and GFP fluorescence of -1 PRF reporters was lost when replacing the downstream region with structural variants (more) different from the native sequence (Fig 3I middle column, Fig S7 middle panels; these structures generally had the potential to induce frameshifting in certain contexts; Fig 3G). Likewise when testing for correlation across all variants of a PRF site (Fig 3I right column, Fig S7 right panels), there was no association of lower minimum free energy (MFE) with higher PRF rates.

Many stimulatory downstream sequences are thought to fold into a pseudoknot (e.g. SRV-1 (ten Dam et al., 1994; Du et al., 1997), west nile virus (Firth and Atkins, 2009) and coronaviruses (Brierley et al., 1989; Plant et al., 2005)), a more complex folding pattern that is not covered by many secondary structure predictions including the Vienna RNA package (Hofacker et al., 2004; Lorenz et al., 2011). We therefore repeated the analyses using pKiss (Janssen and Giegerich, 2015), an algorithm that can detect certain classes of pseudoknots (see Supplementary Data 2 for pKiss-predicted structures of library variants with systematic sequence changes in the downstream region). MFE calculated using Vienna RNAfold and pKiss, respectively, show good correlation (Fig S6B). We repeated the

above analysis and observed the same pattern as with RNAfold, namely widespread association
between secondary structure (low MFE) and higher PRF signal for small deviations from the native
sequence (Fig 3J left column, Fig S8 left panels), but not otherwise (Fig 3J middle and right column,
Fig S8 middle and right panels). This is in line with the fact that we do not see systematic differences
between PRF events whose downstream sequence was shown to fold into a hairpin structure (e.g.
HIV-1 (Parkin et al., 1992; Staple and Butcher, 2003) and SIVmac239 (Marcheschi et al., 2007)) or a
pseudoknot (e.g. SRV-1 (ten Dam et al., 1994; Du et al., 1997) and SARS (Brierley et al., 1989; Plant
et al., 2005)).

This observation does not imply a (counterintuitive) tendency for more open secondary structure to
promote frameshifting. As we do not find evidence for an effect of MFE on overall RNA levels (Fig S7,
S8, S9; e.g. through triggering No-Go decay (Belew et al., 2011)), we suggest that this reflects a
strong preference for downstream secondary structure rigidity to be close to that of the wild-type
sequence (in particular for SARS, SRV1 and HERV-K10, Fig S7, S8, right panels). Although present
secondary structure prediction algorithms might not be sufficiently accurate for this task, especially in
the case of pseudoknots, our results support a view according to which not only the structure, but also
the sequence downstream of a slippery site is critical for triggering frameshifting.

Additional regulatory elements contribute to PRF

The distance between the slippery site and the downstream secondary structure has been shown to
be important for efficient frameshifting (Brierley et al., 1989; Kollmus et al., 1994; Lin et al., 2012),
probably to position the paused ribosome at the slippery site (Brierley and Pennell, 2001). To test
whether this is generally true or if there are PRF event-specific differences, we introduced between
one and nine nucleotides from different constant sequences or deleted between one and six, mutating
downstream stop codons as required (Fig 4A top). Most insertions or deletions led to substantially
reduced frameshifting signal (Fig 4B), no matter what the context and if the original downstream
coding frame was preserved or not, suggesting that the exact position of the downstream regulatory
elements is important across PRF events. This is in contrast to an earlier study reporting unchanged
or even dramatically increased -1 frameshifting efficiency at the HIV site upon reducing spacer
lengths, which the authors attribute to a change in local secondary structure triggered by the specific
deletion introduced in their construct (Mouzakis et al., 2013). Overall, our results are in line with the
regulatory elements being in optimal distance from one another and the downstream secondary
structure creating a roadblock that causes the ribosome to pause exactly at the slippery site (Brierley
and Pennell, 2001; Tu et al., 1992).

While mutations in the slippery site and the downstream region typically have a negative effect on
frameshifting efficiency (Fig 2A), changes in the upstream region have the potential to lead to higher

Figure 4. Additional regulatory elements contribute to PRF.

A. Schematic of sequence manipulations upstream and immediately downstream of the slippery site. B. Boxplots of % wild-type frameshifting rates for variants of "positive" -1 PRF sites in which the indicated number of nucleotides were deleted (negative numbers) or inserted (positive numbers, taken from two different constant sequences) after the slippery site (n=4-54). C. % GFP fluorescence for variants of "negative" -1 PRF sites with the sequence upstream of the slippery site being randomly recoded or replaced with sequences predicted to have the indicated secondary structure; the shaded area denotes the range of background fluorescence. D. % wild-type frameshifting rates (+1 for OAZ, blue, and -1 for HIV, green) of variants in which the native upstream region has been replaced by constant sequences up to the indicated position relative to the PRF site. E. Boxplot of percent of wild-type frameshifting rates for variants in which the -2 or -1 amino acid relative to the slippery site is replaced with an amino acid from the indicated groups, for codons which maintain the slippery site pattern XXXYYYZ (bottom) or not (top); n=10-115. F. Pearson correlation coefficient (blue) and associated p-values (green) between tAI at the indicated position of the original reading frame and % GFP fluorescence. G. Clustered heat map showing all possible combinations of 3 synthetic slippery sites, 34 synthetic downstream variants and up to 5 synthetic upstream regions (minimal value 1.3%).

389   -1 PRF signal (e.g. PEG10, Fig S5D). This suggests that inhibitory signals upstream of the

390   frameshifting site as found for SARS (Su et al., 2005) could be a more widespread property of PRF

391   sites, probably creating a balance to ensure that frameshift promoting signals like a rigid downstream

secondary structure do not lead to a complete inhibition of translation and potentially degradation of the mRNA. To test for effects of upstream secondary structure on PRF, we replaced the region upstream of the slippery site with a recoded version (encoding the same amino acid sequence), a sequence predicted to fold into the same secondary structure as the wild-type, a sequence lacking any strong predictable secondary structure or a hairpin at the 5' or 3' end. Specifically in the case of PEG10, a hairpin at the 5' end of the upstream region or a lack of secondary structure resulted in a frameshifting signal above threshold (2-8% GFP fluorescence; Fig 4C) as opposed to the wild-type sequence (Fig 1C, see our discussion of potential reasons above).

The upstream region of OAZ1 is thought to harbor signals enhancing frameshifting (cf. Ivanov et al., 2006). By testing variants in which increasing portions of the upstream region have been replaced with constant sequences, we find these PRF-promoting signals to be located in the 20 nucleotides before the slippery site (Fig 4D, decrease of PRF signal on average to 31.3% of wild-type, p=0.0077, Wilcoxon signed-rank test, effect on region further upstream not significant). In contrast to the other -1 PRF sites, HIV frameshifting rates show an increase upon mutation of the corresponding upstream region (Fig 4D, mean increase of PRF signal to 128.2% of wild-type, p=0.013, Wilcoxon signed-rank test, effect on region further upstream not significant), indicating the presence of an inhibitory element in this region.

We expanded our search for properties affecting PRF efficiency and examined the effect of the codons preceding the slippery site. We found that the presence of a charged amino acid immediately upstream of the slippery site significantly reduced frameshifting signal, even when the sequence of the slippery site was unchanged (Fig 4E, p<0.0015, Wilcoxon signed-rank test). We examined also the influence of decoding efficiency as measured by the tRNA adaptation index (tAI) on PRF and detected an increasingly negative correlation between tAI and PRF signal in the 0 frame codons leading up to the site of ribosome slippage (Fig 4F). This observation might indicate that progressive slowing down of the ribosome by limited tRNA availability might contribute to stalling at the slippery site and frameshifting.

To not limit our analysis to endogenous frameshifting events, we designed completely synthetic regions corresponding to a slippery site and the upstream and downstream region with different sequence and structural properties. Some fully designed sequences were able to trigger frameshifting and resulted in a -1 PRF signal of up to 11% GFP fluorescence, approximately corresponding to the wild-type frameshifting rate of the most efficient event tested here (SARS coronavirus). We tested all different downstream regions with synthetic slippery sequences resembling common types of -1 PRF sites and found pronounced differences in combinatorial preferences (Fig 4G), showing that also fully synthetic frameshifting events exhibit the combinatorial preferences and context-dependent peculiarities observed for native PRF events.

Prediction of frameshifting potential and efficiency

Having accurate quantitative measurements for large collections of frameshifting sites, we aimed to predict frameshifting efficiency using machine learning approaches. Based on previous studies and our own findings, we used the identity of slippery site positions, tAI of codons around the frameshifting site, amino acid class, minimum free energy and pairedness of positions downstream of the frameshifting site (see Methods), alone and combined, as features and built computational models based on Gradient Boosting Decision Trees (XGBoost (Chen and Guestrin, 2016); Fig 5A and Methods). These models were either estimating whether a sequence has the ability to induce frameshifting (i.e. yield a fluorescent signal above threshold (1.3%, see Figure S2A-C); classification) or quantitatively predicting PRF efficiency (approximated by GFP fluorescence intensity in our assay; regression). We achieved high accuracy (up to an area under the receiver operating characteristic curve (ROC AUC)=0.93 for the classification of a sequence as yielding a PRF signal or not, and Pearson r=0.81 for the comparison between measured and predicted frameshifting signal (% GFP fluorescence)) when training our model on variants of specific frameshifting events and predicting unseen variants from the same event (Fig 5BC). Despite the differences in the regulatory characteristics between individual PRF events, we could predict frameshifting ability (Fig 5D, Fig S10A, ROC AUC=0.89; area under the Precision-Recall-Curve=0.8) and frameshifting signal (% GFP fluorescence; Fig 5E, Pearson r=0.53) also for a pool of all sequence variants tested, no matter whether they yielded a detectable frameshifting signal in our assay or not (Fig 5BC, Fig S10A; "-1 PRF events" vs. "all library variants").

To determine to what extent a model trained on one PRF event could predict frameshifting potential and -1 PRF signal in a different context, we trained our model (based on the full set of features) on sequence variants derived from one event and predicted on variants of another event (Fig 5F, Fig S10B). While performance is usually best for variants of the same PRF event, this approach revealed functional similarities that allowed relatively accurate prediction on "foreign" PRF events (ROC AUC up to 0.94, Pearson r up to 0.69). There is considerable overlap with other attempts of functional clustering, e.g. based on sensitivities to mutations (Fig 2A), with SRV1 and HERV-K10 PRF sites showing similar behavior and HIV variants yielding good prediction scores for models trained on data from any retroviral PRF event (HIV, SIVmac239 and SRV1; Fig 5F). The SARS PRF event, however, is not amenable to prediction based on models trained on data from other PRF events. Prediction accuracy also reflects the specific components of particular importance for a PRF event, e.g. the slippery site for HIV (Fig S10C) and the downstream region for SRV1 (Fig S10D). In these cases,

Figure 5. Prediction of frameshifting potential and efficiency.
A. Outline of the process of generating features and building and testing predictive models. BC. Prediction scores (area under the receiver operating characteristic curve ("ROC AUC") in the case of classification (B) and Pearson correlation coefficients in the case of regression (C)) on held-out test data for the indicated PRF event(s) or the entire library using different feature sets (slip: nucleotide identities in the canonical XXXYYYZ slippery site; tai: tAI scores around the slippery site, aa: amino acid class ("unipolar", "polar", "charged") around the slippery site, dg: MFE of downstream regions, sec: predicted pairedness of downstream positions). D. ROC curve showing performance of a classifier trained on 80% of all designed library variants passing filtering and testing on the remaining 20%. E. For the test set (20% of all designed library variants of all -1 PRF events (left) or HIV-1 (right) passing filtering) the measured % GFP fluorescence is plotted against the model prediction (trained on 80% of all library variants of all -1 PRF events (left) or HIV-1 (right) passing filtering). F. Prediction scores (ROC AUC) on held-out test data (20%) for the indicated PRF event (columns) using the full set of features and training the model on the training data (80%) from the indicated PRF event (rows).

frameshifting efficiency can be predicted using models trained on data from almost any other PRF
event.

Interpretation of the prediction models provides an additional approach to identify the most important
properties of a (potential) PRF site. Here, we used Shapely (SHAP) values for determining the
contribution of each feature to the prediction result of every sample (Fig S11A; Lundberg et al., 2020).
The features of the different subsets driving the prediction provide additional information about the

effect of specific sequence properties on the readout (Fig S11A), e.g. recapitulating earlier observations that XXXCCCZ ("Y=C") and XXXGGGZ ("Y=G") slippery sites are disfavored (Fig S11A upper left, cf. Fig 2D). For a model trained on the MFE of different lengths of downstream region, stronger secondary structure not always led to prediction of higher PRF efficiency (Fig S11A lower left). When taking into account regions of e.g. 60 nt after the slippery site ("dg60"), a lower feature value favored prediction of PRF (Fig S11B).). While the positive effect of strong secondary structure (low MFE) in the first 60 nt on PRF prediction was strongly enhanced if also the downstream region as a whole showed low MFE (Fig S11B), in general we find strong secondary structure of the entire downstream region (120 nt) to be associated with prediction of no PRF (Fig S11A). This could either be explained through an effect on frameshifting efficiency or on RNA stability and overall expression levels. The latter option is unlikely to explain the effect in full as we do not observe a correlation between MFE of the downstream region and steady-state RNA levels (Fig S11C).

HIV clinical isolates exhibit subtype-specific differences in frameshifting efficiency and associations with viral load in patients

The HIV gag-pol frameshifting site is arguably one of the most intensely studied examples of PRF and – due to its critical importance for the viral replication cycle – has been repeatedly suggested as an antiviral drug target (Brakier-Gingras et al., 2012; Hung et al., 1998). To assess the natural variation in frameshifting rates in HIV1 we assembled a set of 581 sequences from clinical isolates between 1976 and 2014 (http://www.hiv.lanl.gov/, Methods), which differ in the sequence surrounding the frameshifting site, but not in the slippery site itself (Fig S12A). Frameshifting rates of the variants are distributed around the rate observed for the lab strain HXB2 (Fig 6A). This constitutes actual differences in frameshifting rates and not only experimental variability, as isolated clones show remarkably good correlation given the small differences in PRF signal we are measuring (Fig 6B). Secondary structure showed the best correlation with frameshifting rates when considering the first 30 nucleotides after the frameshifting site (Fig S12B), matching the region of high sequence conservation (Fig S12A). We grouped the HIV variants based on subtype and found significant differences between the groups (Fig 6C, $p<8*10^{-11}$, one-way ANOVA), most notably higher frameshifting rates in subtype C ($p<6*10^{-12}$ for the difference between C and B), in contrast to an earlier report (Baril et al., 2003). HIV subtypes show distinct geographical distributions (Hemelaar et al., 2006), and consequently we also observed differences between countries of origin (Fig S12C, $p<2*10^{-4}$), but no change in frameshifting rates over time (Fig S12D, p=0.4).

Optimal gag-pol frameshifting rates have been proposed to be critical for virulence (Dinman and Wickner, 1992; Felsenstein and Goff, 1988). In order to link the frameshifting rate measured in our reporter assay with the replicative success of the corresponding HIV isolate we compared the viral

Figure 6. Testing of PRF sites from HIV clinical isolates reveals subtype specific differences and associations with viral load in patients.

A. Distribution of percent GFP fluorescence of all measured HIV variants. B. Percent GFP fluorescence as determined by FACSseq is plotted against the value determined by measuring the corresponding isolated clone by FACS. C. Boxplot showing percent GFP fluorescence of HIV gag-pol PRF variants coming from the indicated subtypes (n=8-112). D. Viral load (HIV titer) is plotted against percent GFP fluorescence determined for the natural gag-pol PRF site variant from the corresponding clinical isolate; vertical line: HIV HXB2 wild-type percent GFP fluorescence. E. Predicted vs measured percent GFP fluorescence for 20% of the HIV variants (held-out test set) with the full feature set except for the identity of slippery site positions (left) or after removing MFE features (middle) or all secondary structure derived features (right) features.

load in patients (where available) to frameshifting rates of the causal HIV isolate (Fig 6D). Although viral load is influenced by many factors, we nevertheless observed a clear trend for patients with high viral load to have frameshifting rates close to "wild type" (the lab strain HXB2). These data present optimality of frameshifting rates as a hallmark of HIV infection and as being associated with infectious success, underscoring the potential for drugs altering the efficiency of gag-pol frameshifting.

In order to predict frameshifting efficiency of naturally occurring variants of the HIV gag-pol frameshifting site, we built a model based on the feature sets described above. Despite eliminating the most important feature of the HIV gag-pol PRF site (the slippery sequence) by restricting our analysis to clinical isolates with no sequence alterations in and around the canonical UUUUUUA and despite the narrow range of frameshifting rates (Fig 6A), we could accurately predict frameshifting rates of unseen variants (Pearson r=0.6; Fig 6E, left). Interestingly, removing the features based on pairedness of downstream positions and minimum free energy led to similar agreement between measured and predicted values (Fig 6E, right). This is in line with the limited importance of the exact downstream structure in comparison to other PRF events tested. Including designed HIV variants in the training set yielded similar prediction accuracy (Fig S12E), indicating that our model can learn the

relevant rules from our set of natural variants alone. Taken together, these results show that a model
based on data from our assay is sensitive enough to detect even subtle differences between HIV
variants and allows prediction of frameshifting rates with an accuracy of clinical relevance.

Discussion

Here, we combined fluorescent frameshifting reporters with rational design of DNA sequences and
high-throughput testing to systematically decipher the rules governing PRF. Our approach aims at a
systematic and comparative assessment of the commonalities and peculiarities in the regulation of
frameshifting efficiency across PRF events. It is complementary to in-depth analyses of individual
frameshifting sites yielding detailed structural or kinetic insights (Caliskan et al., 2014, 2017; Chen et
al., 2014; Choi et al., 2020; Lyon et al., 2019; Yan et al., 2015). The high-throughput nature of our
approach entails particular limitations. While we show that our assay is highly sensitive to changes in
the PRF signal (e.g. in the testing of natural isolates of HIV) and therefore able to identify even subtle
effects of sequence alterations on frameshifting, our detection limit is higher than in assays using
amplification methods and/or overexpression of the reporter.

Prediction of PRF events has up to now been largely limited to identifying sequences matching the
canonical pattern for slippery sites, followed by a downstream secondary structure (Belew et al., 2008;
Theis et al., 2008). To the best of our knowledge, no attempts at quantitatively predicting the effect of
sequence variation exist to date. We used our measurements of frameshifting signal of thousands of
variants of known PRF sites to build a computational model based on known and novel sequence
features affecting frameshifting, leading to accurate prediction of frameshifting potential (up to ROC
AUC=0.93) and frameshifting rates (up to Pearson r=0.81). In many cases sequence properties other
than slippery site identity or downstream secondary structure rigidity yielded high prediction scores,
emphasizing the importance to include additional features such as amino acid properties and tRNA
availability in the investigation of PRF regulation.

By controlled sequence and structure manipulations in multiple contexts we aimed to dissect and
directly compare the regulatory architecture of 15 frameshifting events, revealing a great diversity in
regulatory strategies involving upstream and downstream sequence and structural elements and
amino acid properties. We demonstrate a high degree of optimization and specialization of regulatory
mechanisms. For example, no universally optimal slippery site exists, but in each of the PRF events
tested the native slippery site was found to be optimal given the sequence context.

A correlation between secondary structure rigidity and PRF efficiency has been shown previously
(Bidou et al., 1997; Garcia-Miranda et al., 2016; Wu et al., 2018), while also other properties of
secondary structures like their plasticity have been reported to be associated with frameshifting rates

(Halma et al., 2019; Ritchie et al., 2012). Our results corroborate the functional link between minimum

free energy of downstream structures and show that this is true for most PRF events tested. However,

this correlation generally depended on the length of the downstream region taken into account, in

agreement with the specific association between PRF efficiency and thermodynamic stability of the

start of the HIV stem loop structure reported earlier (Mouzakis et al., 2013). Furthermore, the

correlation between PRF signal and MFE only held true when testing small variations of the native

sequence and structure; when sequences and structures different from the native one were tested,

this correlation was lost.

Slippery site and a downstream stimulatory signal are central – and therefore intensely studied –

regulatory elements present in all known PRF events. Here, we expanded the repertoire of regulatory

inputs and demonstrated that both tRNA availability as well as amino acid properties can affect PRF

efficiencies across native sequence contexts.

Ribosome stalling due to limited tRNA availability has the potential to lead to slippage of the ribosome

(Gallant and Lindsley, 1992; Weiss et al., 1988) and contributes to non-programmed frameshifting in

huntingtin (Girstmair et al., 2013). Availability of tRNAs decoding slippery site codons have been

shown to influence HIV gag-pol frameshifting efficiency (Korniy et al., 2019) and HIV-1 itself has the

ability to modulate the tRNA pool of the host cell (van Weringh et al., 2011). Our observation that

limited tRNA availability in the codons leading up to the slippery site progressively contributes to PRF

efficiency is in line with these data and expands the notion of an influence of the tRNA pool on

frameshifting beyond the slippery site codons and individual PRF events.

In addition, our analyses highlighted amino acid properties around the frameshifting sites as another

factor influencing PRF efficiency. Specifically, a charged amino acid at the last position before the

slippery site was associated with reduced frameshifting rate. Together with reports of a frameshift-

promoting effect of specific amino-acid sequences upstream of the frameshifting site in the case of

+1 PRF in fungal OAZ genes (Yordanova et al., 2015) and the copA gene of Escherichia coli (Meydan

et al., 2017), this observation provides additional evidence for a role of the nascent peptide in

regulating ribosomal frameshifting.

In general, our results reveal functional groups of PRF events that exhibit greater similarity in terms

of their sensitivities to mutations in regulatory regions and congruence in the preferences for

secondary structure. These functional groups also manifest themselves in a higher prediction

accuracy within these groups. The groups resemble similarities also according to other criteria, e.g.

in their native secondary structure. PRF events, whose downstream stimulatory sequence has been

shown to be a stem loop (e.g. HIV-1 (Parkin et al., 1992; Staple and Butcher, 2003) and SIVmac239

(Marcheschi et al., 2007)) tend to be more resilient and less affected by small sequence changes than

those with a downstream pseudoknot (e.g. SRV-1 (ten Dam et al., 1994; Du et al., 1997) as well as

SARS and other coronaviruses (Brierley et al., 1989; Plant et al., 2005)). Typically, these cases exhibit
higher fluorescence signal in our assay (Fig 1C), but greater sensitivities to even small sequence
changes (Fig 2B, 3C, S4AB, S5A) and requirements for a more extensive downstream region. This
might also entail a greater sensitivity to the experimental context and therefore explain the lack of
signal (above the noise threshold) for previously reported cases depending on a pseudoknot as
frameshift stimulatory signal like Rous sarcoma virus (Jacks and Varmus, 1985; Marczinke et al.,
1998). In other cases, low wild-type rates of frameshifting, requirements for additional trans-acting
factors or an effect on overall expression levels might explain the lack of signal in our assay.

Comparing the HIV gag-pol site with other PRF events shows the remarkable robustness and
exceptional tolerance for various non-native downstream stimulatory signals. More than in other
cases, maintaining the wild-type slippery site UUUUUUA is crucial for frameshifting, and tends to be
absolutely invariable across otherwise diverse natural isolates (Biswas et al., 2004). The quantitative
nature of our assay allows us nevertheless to accurately quantify the effect of sequence variation
away from the slippery site, revealing sub-type specific differences. Furthermore, we devised a
machine learning model that is able to predict the natural variation in HIV frameshifting rates for novel
variants (Pearson r=0.60). Precise stoichiometry of frameshifted and non-frameshifted product is
crucial for HIV virulence, and the ability to predict frameshifting rates of naturally occurring variants of
the gag-pol PRF site therefore has important clinical implications, both in the case of HIV as well as
in the case of PRF events in other viruses.

This high-throughput investigation of PRF also constitutes an addition to the experimental toolbox for
studying translational recoding and highlights many starting points for further investigations into the
regulatory basis and functional importance of frameshifting. In addition to deciphering the effect of cis-
regulatory elements, regulation by trans-acting factors like the recently discovered case of the
frameshift repressor shiftless (Wang et al., 2019) constitutes another area where a systematic
approach like the one presented here can yield novel insights into the mode of action and its
specificity. Libraries of frameshifting reporters provide a platform for screening putative modifiers of
ribosomal frameshifting, offering a powerful tool to identify ways to generally or selectively control
frameshifting and opening new possibilities for interfering with viral replication and for controlling
cellular processes depending on translational frameshifting.

Methods

Synthetic library design

General design notes

Oligonucleotides were designed to maintain a constant length of 210 nt. Restriction sites used for cloning were excluded from the design. All the variants were composed of an 18 nt forward primer, 12 nt barcode sequence, 162 nt variable region and 18 nt reverse primer sequences. DNA barcodes were designed to differ from any other barcode in the library in at least 3 nt.

Selection of frameshifting sites for systematic sequence manipulations

60 PRF events reported previously or listed in FSDB (Moon et al., 2007) were included in the library design. Sequences were aligned in a way that the site of frameshifting would be at the same position in all variants (slippery site ending at position 42). In case there were stop codons downstream of the PRF site in any of the three frames, additional versions with stop codons mutated (changed to TGG) were included for testing only. Out of these, 15 were selected for systematic sequence manipulations based on the amount of experimental evidence supporting them, lack of stop codons in the relevant reading frames, biological significance and mechanistic diversity (i.e. covering +1, -1 and -2 frameshifting sites). The HIV and OAZ frameshifting sites were tested beforehand in isolation to confirm that our experimental setup is able to reproduce well established cases of -1 as well as +1 PRF.

Design of individual rationally designed subsets

For each of the subsets in the libraries, systematic sequence manipulations were performed on the set of 15 previously reported frameshifting sites. Design of subsets was carried out in Python.

Multiple barcode controls: We added multiple variants to the library that contained the same variable region, but different barcodes, in order to gauge potential effects of the barcode and the technical noise of our assay.

Slippery site: We tested all possible variations of the different slippery site patterns found in previously reported frameshifting events (XXXYYYZ, XYYZZZZ, influenza, OAZ (for +1 PRF)). In the case of OAZ, both the first and the second codon in the slippery site UCC UGA were replaced with all other codons. In addition, we created all possible point mutations in -1 PRF slippery sites that would not give rise to a stop codon.

Immediate downstream region/"spacer": We created all possible point mutations in the six nucleotides

downstream of PRF slippery sites that would not give rise to a stop codon. In addition, we removed

up to 6 bps immediately downstream (in 1 or 3 bp increments, depending on the presence of stop

codons in downstream reading frames) or introduced up to 12 bps from five different constant

sequences (in 1 or 3 bp increments, depending on the presence of stop codons in downstream

reading frames).

Immediate upstream region: The first two codons preceding the slippery site (in the original frame)

were replaced with all other codons (except stop codons).

Upstream region: The upstream region (excluding the last two codons before the slippery site) was

randomly recoded five times. Moreover, variants were added for all 15 PRF events in which the same

part of the upstream region was replaced by four different constant sequences in increments of six

bps, starting from the beginning of the variable region. In addition, we included the following structural

variations in the design: We designed sequences 30 nt in length predicted (using ViennaRNA 2.0

(Lorenz et al., 2011) and antaRNA (Kleinkauf et al., 2015a, 2015b)) to fold into the respective native

structure (but having a different primary sequence or to have no secondary structure, a 9 bp long

stem-loop at the 5' or at the 3' end (6 sequences per set) and replaced the endogenous region in all

15 PRF events.

Downstream region: Variants were added for all 15 PRF events in which the same part of the

downstream region was replaced by three different constant sequences (different sets were used for

predicted +1 and -1 PRF sites to avoid stop codons in the relevant frames) in increments of six bps,

starting from the end of the variable region. For scanning mutagenesis, 30 positions (leaving the first

6 bp after the slippery site unchanged) were replaced with either A or C (G and T were avoided as

they often give rise to stop codons in one of the frames). In addition, we included the following

structural variations in the design: We used ViennaRNA 2.0 or pKiss (Janssen and Giegerich, 2015)

(which also predicts pseudoknots) to determine the native secondary structure of each of the 15 PRF

events. We then created for each PRF event two sets of ten sequences each (using antaRNA) that

are predicted to fold into the downstream secondary structure as determined using ViennaRNA and

pKiss, respectively. Furthermore we designed variants of the predicted HIV, SRV1 and nsp2F

(PRRSV) downstream secondary structure, e.g. by extending or reducing the length of hairpins, and

completely synthetic secondary structures and again used antaRNA to create sets of corresponding

sequences that we used to replace the native downstream structure in all 15 PRF events.

Completely synthetic variants: We designed a set of four upstream regions, three potential slippery

sites (resembling previously reported cases of the patterns XXXYYYZ, XYYZZZZ and XXYYYYY),

and created all possible combinations with a set of 57 artificial downstream sequences.

Frameshifts: We created variants in which the slippery site was moved by one, two or three positions
in the 5' or 3' direction in order to verify the frame-dependence of the effect.

Combinatorial variants: Upstream, slippery site and downstream regions from the whole set of 15
PRF sites were recombined, either preserving the original sequence fully or replacing potential stop
codons in all frames with TGG.

gag-pol frameshifting sites from HIV clinical isolates

We retrieved >30,000 full length sequences of the HIV genome from clinical isolates
(http://www.hiv.lanl.gov/) and selected gag-pol frameshifting sites which had the canonical sequence
around the slippery site (TAATTTTTA), but differed in at least one position in the 30 nucleotides
upstream or the 120 nucleotides downstream. We chose to exclude variability in the slippery site to
screen for more subtle differences in frameshifting rates across clinical isolates.

Experimental procedures

K562 cell culture

K562 cells were acquired from ATCC. Cells were grown in Iscove's Modified Dulbecco Medium
supplemented with 10% fetal bovine serum (SIGMA) and 1% Penicillin-Streptomycin solution
(SIGMA). The cells were split when reaching a concentration of $\sim 10^6$ cells/ml. The cells were grown
in an incubator at 37°C and 5% CO_2. Cells were frozen in batches of 4×10^6 cells in growth medium
supplemented with 5% DMSO.

Construction of the master plasmid

Master plasmids for library insertion were constructed by amplifying parts from existing vectors and
cloning the parts sequentially into pZDonor 3.1. The master plasmid contained the EF1alpha
promoter, *mCherry*, a designed multiple cloning site containing restriction sites for library cloning
(RsrII and AscI), *gfp* and the SV40 terminator sequence. In the case of the master plasmid for cloning
the stop-vector library, two complementary DNA oligonucleotides (GACTGATAGCTGACTAGTCG
and GTCCGACTAGTCAGCTATCAG) were synthesized (IDT), annealed and introduced into the
vector described above (cut with RsrII) such that the mCherry-proximal RsrII site was destroyed and
the downstream RsrII site (after the inserted stop cassette) was subsequently used for inserting the
library.

Synthetic library cloning

The cloning steps were performed essentially as described previously (Vainberg Slutskin et al., 2018). We used Agilent oligo library synthesis technology to produce a pool of 17,809 different fully designed single-stranded 210-oligomers (Agilent Technologies, Santa Clara, CA), which was provided as a single pool of oligonucleotides (10 pmol). The two subsets of this pool corresponding to the designed and native libraries tested here were defined by unique amplification primers. The reverse primer contained a spacer of either 4 or 5 nucleotides to create the offset in the *gfp* reading frame. The oligo pool was dissolved in 200 µl Tris-ethylene-diamine-tetraacetic acid (Tris-EDTA) and then diluted 1:50 with Tris-EDTA, which was used as template for PCR. We amplified each of the four libraries by performing 8 PCR reactions, each of which contained 19 µl of water, 5 µl of DNA, 10 µl of 5× Herculase II reaction buffer, 5 µl of 2.5 mM deoxynucleotide triphosphate (dNTPs) each, 5 µl of 10 µM forward primer, 5 µl of 10 µM reverse primer, and 1 µl Herculase II fusion DNA polymerase (Agilent Technologies). The parameters for PCR were 95°C for 1 min, 14 cycles of 95°C for 20 s, and 68°C for 1 min, each, and finally one cycle of 68°C for 4 min. The oligonucleotides were amplified using library-specific common primers in the length of 35 nt, which have 18-nt complementary sequence to the single-stranded 210-mers and a tail of 17 nt containing RsrII (forward primer) and AscI (reverse primer) restriction sites. The PCR products were concentrated using Amicon Ultra, 0.5 ml 30K centrifugal filters (Merck Millipore). The concentrated DNA was then purified using a PCR mini-elute purification kit (Qiagen) according to the manufacturer's protocol. Purified library DNA (540 ng total) was cut with the unique restriction enzymes RsrII and AscI (Fermentas FastDigest) for 2 hours at 37°C in two 40-µl reactions containing 4 µl fast digest (FD) buffer, 1 µl RsrII enzyme, 1 µl AscI enzyme, 18 µl DNA (15 ng/µl) and 16 µl water, followed by heat inactivation for 20 min at 65°C. Digested DNA was separated from smaller fragments and uncut PCR products by electrophoresis on a 2.5% agarose gel stained with GelStar (Cambrex Bio Science Rockland). Fragments were cut from the gel and eluted using electroelution Midi GeBAflex tubes (GeBA, Kfar Hanagid, Israel). Eluted DNA was precipitated using sodium acetate–isopropanol. The master plasmid was cut with RsrII and AscI (Fermentas FastDigest) in a reaction mixture containing 6 µl FD buffer, 3 µl of each enzyme and 3.5 µg of the plasmid in a total volume of 60 µl. After incubation for 2.5 hours at 37°C, 3 µl FD buffer, 3 µl alkaline phosphatase (Fermentas) and 24 µl water were added and the reactions were incubated for an additional 30 mins at 37°C followed by 20 min at 65°C. Digested DNA was purified using a PCR purification kit (Qiagen). The digested plasmid and DNA library were ligated for 30 min at room temperature in a 10 µl reactions, containing 150 ng plasmid and the library in a molar ratio of 1:1, 1 µl CloneDirect 10× ligation buffer, and 1 µl CloneSmart DNA ligase (Lucigen Corporation), followed by heat inactivation for 15 min at 70°C. Ligated DNA was transformed into *E. coli* 10G electrocompetent cells (Lucigen) divided into aliquots (23 µl each, plus 2 µl of the ligation mix), which were then plated on 4 Luria broth (LB) agar (200 mg/ml amp) 15-cm plates per transformation reaction

(25 µl). The rationally designed (12,809 variants) and the native (5000 variants) parts of the library

were cloned separately. For the rationally designed and the native library we collected around 10^6 and

4×10^5 colonies, respectively, for each frame the day after transformation by scraping the plates into

LB medium. Library-pooled plasmids were purified using a NucleoBond Xtra maxi kit (Macherey

Nagel). To ensure that the collected plasmids contain only a single insert of the right size, we

performed colony PCR (at least 48 random colonies per library).

Transfection into K562 cells and genomic integration

The purified plasmid libraries were transfected into K562 cells and genomically integrated using the

Zinc Finger Nuclease (ZFN) system for site-specific integration and the CompoZr® Targeted

Integration Kit - AAVS1 (SIGMA). Transfections were carried out using Amaxa® Cell Line

Nucleofector® Kit V (LONZA). To ensure library representation we performed 15 nucleofections of

the purified plasmid library for each rationally designed library (frame -1 and frame +1) and 5 for each

native library. For each nucleofection, 4×10^6 cells were centrifuged and washed twice with 20 ml of

Hank's Balanced Salt Solution (HBSS, SIGMA). Cells were resuspended in 100 µl solution (warmed

to room temperature) composed of 82 µl solution V and 19 µl supplement (Amaxa® Cell Line

Nucleofector® Kit V). Next, the cells were mixed with 2.75 µg of donor plasmid and 0.6 µg ZFN mRNA

(prepared in-house) just prior to transfection. Nucleofection was carried out using program T-16 on

the NucleofectorTM device, immediately mixed with ~0.5 ml of pre-cultured growth medium and

transferred to a 6 well plate with additional 1.5 ml of pre-cultured growth medium. A purified plasmid

library was also transfected without the addition of ZFN and served as a control to determine when

cells lost non-integrated plasmids.

Sorting the library by FACS

K562 cells were grown for at least 14 days to ensure that non-integrated plasmid DNA was eliminated.

A day prior to sorting, cells were split to ~0.25×106 cells/ml. On the day of sorting, cells were

centrifuged, resuspended in sterile PBS and filtered using cell-strainer capped tubes (Becton

Dickinson (BD) Falcon). Sorting was performed with BD FACSAria II SORP (special-order research

product) at low sample flow rate and a sorting speed of ~18000 cells/s. To sort cells that integrated

the reporter construct successfully and in a single copy (~4% of the population), we determined a

gate according to mCherry fluorescence so that only mCherry-expressing cells corresponding to a

single copy of the construct were sorted (mCherry single population). We collected around 350

cells/variant on average for each library in order to ensure adequate library representation.

Cells sorted for single integration of the transgene were grown for a week before we sorted the
population (combined rationally designed and native library, but separately for -1 PRF reporters, +1
PRF reporters and the stop-vector control) into 16 (in the case of the stop-vector 12) bins according
to GFP fluorescence, after gating for a narrow range of mCherry expression to avoid effects coming
from the influence of the variable region on overall expression level. This resulted in the loss of a
significant fraction of variants (as only around 50% of the cells fell into the narrow range of mCherry
fluorescence) but ensured that the subsequent sorting by GFP fluorescence was not biased by
expression level. Each bin was defined to span a range of GFP fluorescence values such that it
contains between 1-10% of the cell population. We collected around 1000 cells/variant on average in
order to ensure adequate library representation. Cells from each bin were grown separately for
freezing and purification of genomic DNA.

Genomic DNA purification, amplification and sample preparation

For each of the bins we purified genomic DNA by centrifuging 5×10^6 cells, washing them with 1 ml
PBS and purifying DNA using DNeasy Blood & Tissue Kit (Qiagen) according to the manufacturer's
protocol. In order to maintain the complexity of the library amplified from gDNA, PCR reactions were
carried out on a gDNA amount calculated to contain a minimum average of 200 copies of each oligo
included in the sample. For each of the 16 bins, we used 15 µg of gDNA as template in a two-step
nested PCR. In the first step, three reactions were performed and each reaction contained 5 µg gDNA,
25 µl Kapa Hifi ready mix X2 (KAPA Biosystems), 2.5 µl 10 µM 5′ primer, and 2.5 µl of 10 µM 3′ primer.
The parameters for the first PCR were 95°C for 5 min, 18 cycles of 94°C for 30s, 65°C for 30s, and
72°C for 60s, each, and one cycle of 72°C for 5min. In the second PCR step, each reaction contained
2.5 µl of the first PCR product, 25 µl of Kapa Hifi ready mix X2 (KAPA Biosystems), 2.5 µl 10 µM 5′
primer, and 2.5 µl 10 µM 3′ primer. The PCR program was 95°C for 5 min, 24 cycles of 94°C for 30s
and 72°C for 30s, each, and one cycle of 72°C for 5min. Specific primers corresponding to the constant
region of the plasmid were used. The 5′ primer contained a unique upstream 8-nt bin barcode
sequence, and three different barcodes were used for each bin. The 3′ primer was common to all bins.
Multiple PCR reaction products of each bin were combined. The concentration of the PCR samples
was measured using a monochromator (Tecan i-control), and the samples were mixed in ratios
corresponding to their ratio in the population, as defined when sorting the cells into the 16 bins.
Sample preparation including gel elution and purification were performed as described above for
amplicons from cDNA.

RNA purification, cDNA synthesis and sample preparation

For the cell population sorted for single integration of the reporter construct we performed RNA purification by centrifuging 10^7 cells, washing them with PBS, splitting into two tubes and purifying RNA using NucleoSpin RNA II kit (MACHEREY-NAGEL) according to the manufacturer's protocol. We prepared cDNA in 2 reverse transcription reaction for each replicate using SuperScript® III First-Strand Synthesis System (Thermo Fisher Scientific) with random hexamer primers and 5 µg of input RNA (per reaction) according to the manufacturer protocol. For amplification of the library variants, 3 PCR reactions of 50 ul total volume were performed. Each reaction contained 5 ul cDNA, 25 µl of Kapa Hifi ready mix X2 (KAPA Biosystems), 2.5 µl 10 µM 5′ primer, and 2.5 µl 10 µM 3′ primer. The PCR program was 95∘C for 5 min, 20 cycles of 94∘C for 30s and 72∘C for 30s, each, and one cycle of 72∘C for 5min. Specific primers corresponding to the constant region upstream and downstream of the splice sites were used. The PCR products were separated from potential unspecific fragments by electrophoresis on a 1.5% agarose gel stained with EtBr, cut from the gel, and cleaned in 2 steps: gel extraction kit (Qiagen) and SPRI beads (Agencourt AMPure XP). The sample was assessed for size and purity at the Tapestation, using high sensitivity D1K screenTape (Agilent Technologies, Santa Clara, California). We used 20 ng library DNA for library preparation for NGS; specific Illumina adaptors were added, and DNA was amplified using 14 amplification cycles. The sample was reanalyzed using Tapestation.

For RT-PCR analysis we performed RNA purification on 5 million cells from individually constructed and integrated reporter constructs, followed by cDNA synthesis as described above (one reaction). PCR was performed using primers within the mCherry and GFP coding region, yielding a PCR product of about 1000 bp length. 40 cycles of PCR were used to obtain a strong signal even for less prominent bands. PCR products were analyzed on a 1% agarose gel.

Computational analyses

Mapping next generation sequencing reads

To unambiguously identify the variant of origin, a unique 12-mer barcode sequence was placed at the 5' end of each variable region. DNA was sequenced on a NextSeq-500 sequencer. We determined for each read (PE150) its bin barcode and its variant barcode and discarded all the reads that could not be assigned to a bin and a library variant of origin. In a second step, for each variant the mapped reads (paired-end, 150 bases for each mate) were aligned to the designed sequence (in the context of the reporter) and only those ones were counted that mapped without a mismatch over the whole length (disregarding the last 5 bases). Reads showing a mismatch, in particular a single nucleotide deletion or insertion, were quantified separately and if they passed the same threshold applied to correct library variants a PRF readout (% GFP fluorescence) was calculated for them. These

additional variants were only used in comparisons with the original designed variant and not

incorporated in the general pool of library variants considered for the analyses. For all three libraries

(-1 PRF reporters, +1 PRF reporters and the stop-vector control) around 40 milllion successfully

mapped reads were used for subsequent analysis. In addition, we sequenced the entire cell

population using greater read lengths (PE300) to cover the entire length of the variable region and to

obtain a measure of the fraction of reads per variant containing a synthesis or cloning error. Mapping

was performed using custom-made Python scripts.

Computing a measure of frameshifting efficiency (% GFP fluorescence) and related readouts

We applied a number of filters to the raw sequencing data to reduce experimental noise. First, variants

with less than 20 reads (perfect matches along the whole length) mapped across bins were removed.

Second, for bins with a read count of less than three, the bin value was set to zero. Third, for each

variant we set to zero bins surrounded by zero values as these constituted isolated bins unlikely to

come from the actual distribution. Forth, to reduce bias coming from the "open end"-bins at the

extreme values, we set their count to match their neighbor's if it was higher, as these bins are defined

as containing the tails of the distribution of variants with no GFP fluorescence and maximal GFP

expression, respectively, and peaks in these extreme bins are considered experimental noise. For

each variant, we normalized the values across the 16 bins by dividing the filtered read counts for each

bin by the sum of all filtered read counts. To determine whether a distribution is unimodal we applied

a Savitzky-Golay filter for smoothing the data and removing local maxima introduced by experimental

noise. We detected peaks in the smoothed distribution by a simple approach in which a point is

considered a maximum peak if it has the maximal value, and was preceded by a value lower by 5%.

Variants without detectable peaks (due to the read number not reaching the threshold) were excluded

at this stage. Variants with more than one peak were filtered out at the level of analysis. For each bin,

we calculated the median of the log2 of GFP fluorescence intensity as measured by FACS for all the

cells sorted into that bin and used this as the GFP fluorescence value associated with this bin. For

each variant with one peak after smoothing, we calculated the weighted average for the distribution

of reads across bins (using unsmoothed read counts normalized for each variant and the GFP

fluorescence value for the bins, as described above), resulting in what is referred to in the main text

and figures as the "GFP fluorescence" (in log2).

As described in the main text and illustrated in Figure S2A-C, we set the lowest value obtained (the

variant with the lowest weighted average of the distribution) to 0% GFP fluorescence and the highest

value obtained (the variant with the highest weighted average of the distribution) to 100% GFP

fluorescence. Accordingly, a percentage was assigned to each variant based on the GFP

fluorescence (log2 of the weighted average of the reads distribution) of a variant as follows:

$[\text{\% GFP fluorescence}]_{\text{variant X}} = ([\text{GFP fluorescence}]_{\text{variant X}} - \min([\text{GFP fluorescence}]_{\text{all variants}}))\ /$

$(\max([\text{GFP fluorescence}]_{\text{all variants}}) - \min([\text{GFP fluorescence}]_{\text{all variants}})) * 100$

"% wild-type GFP fluorescence" of a variant was calculated as follows:

$[\text{\% wild-type GFP fluorescence}]_{\text{variant X}} = [\text{\% GFP fluorescence}]_{\text{variant X}}\ /\ [\text{\% GFP fluorescence}]_{\text{wild-type version}}$

$_{\text{of variant X}} * 100$ As a measure for steady state RNA expression levels, we computed the log ratio of

RNA/DNA reads for all variants with at least 20 DNA reads past filtering: $\log2([\text{RNA reads}]_{\text{variant X}}\ /$

$[\text{DNA reads}]_{\text{variant X}})$.

Machine learning approaches

Machine learning procedures were carried out using the python scikit-learn (version 0.18.2) and

XGBoost package. Initially, from all duplicated sequences (e.g. barcode control sets), which passed

filtering, a single variant was randomly chosen for all subsequent steps to avoid biases resulting from

having duplicated sequences. We created sets of variants corresponding either to the full rationally

designed library ("full library"), the -1 PRF sites from the main set of events tested positive in our

assay combined ("frameshifting events") or to all variants based on one of these PRF events in

isolation. 10% of the variants in the individual sets (20% in case of the HIV variants) were put aside

and used only for evaluation of models built using the other 90% (80%). We chose Gradient Boosting

Decision Trees (XGBoost (Chen and Guestrin, 2016)) as the prediction algorithm because it can

capture non-linear interactions between features, allows combining different types of continuous and

categorical features and has proven to be a powerful approach in predicting the effect of regulatory

regions (Mikl et al., 2019; Vainberg Slutskin et al., 2018).

For predictions based on secondary structure we used the fold function from the Vienna RNA package

2.0 and extracted both the minimal free energy ("dg"; calculated for different lengths of downstream

(10-120 nt) regions) and the predicted pairedness for each position ("sec"; for each position in the

downstream region this set of features contains the information whether the position is predicted

(Vienna RNAfold) to be paired or not (0 or 1)).

For predictions based on tRNA adaptation index (tAI), we computed (based on Waldman et al., 2010)

the tAI of codons in the original and -1 frame as well as the (per position) difference between frame -

1 and frame 0 and used the information for 14 starting from the first codon in the slippery site (in the

case of the canonical -1 pattern XXXYYYZ, XXY in frame 0 and XXX in frame -1) as features.

For predictions based on amino acid identity, information on the type of amino acids around the

slippery site (from position -3 relative to the site of frameshifting until the end of the slippery site) was

used ("unpolar", "polar" or "charged").

Different hyperparameter settings for learning rate, n_estimators and max_depth were tested in a
systematic and combinatorial fashion using 10-fold cross-validation for each set of variants and
features independently. Typically around 100 tests were performed and the best set of
hyperparameters used for subsequent steps.

At the end, the model was tested by training it on the entire training set (90% of all relevant unique
library variants, 80% in the case of HIV clinical isolates) and scoring the correlation (Pearson) of
predicted vs. measured values based on the held-out test set (10% of relevant unique library variants
from the relevant set, 20% in the case of HIV clinical isolates), which had not been used at any stage
during development of the model. This approach was used for each of the features sets in isolation
as well as all of the features combined.

For classification, the same pipeline was used. Here, XGBClassifier was trained and the performance
scored on the test set, reporting the area under the ROC curve as a metric.

Feature importance and effect on the model was determined using SHAP analysis (Lundberg et al.,
2020).

General data analysis

For data analysis, we used python 2.7.11 with pandas 0.20.3, numpy 1.13.1, seaborn 0.6, scipy 0.17,
scikit-learn 0.18.2 and shap 0.28.5. Confidence intervals were calculated by bootstrapping (1000
iterations).

Data availability

All data generated in this study are available in the NCBI gene expression omnibus (GEO) under
accession GSE145684.

Code availability

The code used to process the data, generate the figures and train and test the predictive models will
be made available as a GitHub repository (https://github.com/martinmikl/).

Acknowledgements

The authors thank Adina Weinberger, Orna Dahan, Alexey Gritsenko and Roni Rak for helpful
discussions, Ronit Nir, Tali Avnit-Sagi and Maya Lotan-Pompan for technical advice and the Moshe
Oren lab for advice and reagents. This work was supported by an EMBO long-term fellowship (to
M.M.). E.S. is supported by the Crown Human Genome Center; the Else Kroener Fresenius
Foundation; D. L. Schwarz; J. N. Halpern; L. Steinberg; J. Benattar; Aliza Moussaieff; Adelis
Foundation; and grants funded by the European Research Council and the Israel Science Foundation.

Author contributions

Conceptualization: M.M., Y.P. and E.S.; Methodology, Software and Formal Analysis: M.M.;
Investigation: M.M.; Writing: M.M., Y.P. and E.S.; Funding Acquisition: M.M., Y.P. and E.S.;
Supervision: Y.P. and E.S.

The authors declare no competing interests.

References

Advani, V.M., and Dinman, J.D. (2016). Reprogramming the genetic code: The emerging role of
ribosomal frameshifting in regulating cellular gene expression. BioEssays News Rev. Mol. Cell. Dev.
Biol. 38, 21–26.

Baril, M., Dulude, D., Gendron, K., Lemay, G., and Brakier-Gingras, L. (2003). Efficiency of a
programmed −1 ribosomal frameshift in the different subtypes of the human immunodeficiency virus
type 1 group M. RNA 9, 1246–1253.

Belew, A.T., Hepler, N.L., Jacobs, J.L., and Dinman, J.D. (2008). PRFdb: a database of
computationally predicted eukaryotic programmed -1 ribosomal frameshift signals. BMC Genomics 9,
339.

Belew, A.T., Advani, V.M., and Dinman, J.D. (2011). Endogenous ribosomal frameshift signals
operate as mRNA destabilizing elements through at least two molecular pathways in yeast. Nucleic
Acids Res. 39, 2799–2808.

Belew, A.T., Meskauskas, A., Musalgaonkar, S., Advani, V.M., Sulima, S.O., Kasprzak, W.K.,
Shapiro, B.A., and Dinman, J.D. (2014). Ribosomal frameshifting in the CCR5 mRNA is regulated by
miRNAs and the NMD pathway. Nature 512, 265–269.

Bidou, L., Stahl, G., Grima, B., Liu, H., Cassan, M., and Rousset, J.P. (1997). In vivo HIV-1
frameshifting efficiency is directly related to the stability of the stem-loop stimulatory signal. RNA N.
Y. N 3, 1153–1158.

Biswas, P., Jiang, X., Pacchia, A.L., Dougherty, J.P., and Peltz, S.W. (2004). The Human
Immunodeficiency Virus Type 1 Ribosomal Frameshifting Site Is an Invariant Sequence Determinant
and an Important Target for Antiviral Therapy. J. Virol. 78, 2082–2087.

Brakier-Gingras, L., Charbonneau, J., and Butcher, S.E. (2012). Targeting frameshifting in the human
immunodeficiency virus. Expert Opin. Ther. Targets 16, 249–258.

Brierley, I., and Pennell, S. (2001). Structure and function of the stimulatory RNAs involved in
programmed eukaryotic-1 ribosomal frameshifting. Cold Spring Harb. Symp. Quant. Biol. 66, 233–
248.

Brierley, I., Digard, P., and Inglis, S.C. (1989). Characterization of an efficient coronavirus ribosomal
frameshifting signal: requirement for an RNA pseudoknot. Cell 57, 537–547.

Caliskan, N., Katunin, V.I., Belardinelli, R., Peske, F., and Rodnina, M.V. (2014). Programmed -1
frameshifting by kinetic partitioning during impeded translocation. Cell 157, 1619–1631.

Caliskan, N., Peske, F., and Rodnina, M.V. (2015). Changed in translation: mRNA recoding by -1
programmed ribosomal frameshifting. Trends Biochem. Sci. 40, 265–274.

Caliskan, N., Wohlgemuth, I., Korniy, N., Pearson, M., Peske, F., and Rodnina, M.V. (2017).
Conditional Switch between Frameshifting Regimes upon Translation of dnaX mRNA. Mol. Cell 66,
558-567.e4.

Chen, T., and Guestrin, C. (2016). XGBoost: A Scalable Tree Boosting System. Proc. 22nd ACM
SIGKDD Int. Conf. Knowl. Discov. Data Min. - KDD 16 785–794.

Chen, J., Petrov, A., Johansson, M., Tsai, A., O'Leary, S.E., and Puglisi, J.D. (2014). Dynamic
pathways of −1 translational frameshifting. Nature 512, 328–332.

Choi, J., O'Loughlin, S., Atkins, J.F., and Puglisi, J.D. (2020). The energy landscape of −1 ribosomal
frameshifting. Sci. Adv. 6, eaax6969.

Clark, M.B., Jänicke, M., Gottesbühren, U., Kleffmann, T., Legge, M., Poole, E.S., and Tate, W.P.
(2007). Mammalian Gene PEG10 Expresses Two Reading Frames by High Efficiency –1
Frameshifting in Embryonic-associated Tissues. J. Biol. Chem. 282, 37359–37369.

ten Dam, E., Brierley, I., Inglis, S., and Pleij, C. (1994). Identification and analysis of the pseudoknot-
containing gag-pro ribosomal frameshift signal of simian retrovirus-1. Nucleic Acids Res. 22, 2304–
2310.

Dinman, J.D. (2012). Control of gene expression by translational recoding. Adv. Protein Chem. Struct.
Biol. 86, 129–149.

Dinman, J.D., and Wickner, R.B. (1992). Ribosomal frameshifting efficiency and gag/gag-pol ratio are
critical for yeast M1 double-stranded RNA virus propagation. J. Virol. 66, 3669–3676.

Du, Z., Holland, J.A., Hansen, M.R., Giedroc, D.P., and Hoffman, D.W. (1997). Base-pairings within
the RNA pseudoknot associated with the simian retrovirus-1 gag-pro frameshift site. J. Mol. Biol. 270,
464–470.

Felsenstein, K.M., and Goff, S.P. (1988). Expression of the gag-pol fusion protein of Moloney murine
leukemia virus without gag protein does not induce virion formation or proteolytic processing. J. Virol.
62, 2179–2182.

Firth, A.E., and Atkins, J.F. (2009). A conserved predicted pseudoknot in the NS2A-encoding
sequence of West Nile and Japanese encephalitis flaviviruses suggests NS1' may derive from
ribosomal frameshifting. Virol. J. 6, 14.

Firth, A.E., and Brierley, I. (2012). Non-canonical translation in RNA viruses. J. Gen. Virol. 93, 1385–
1409.

Gallant, J.A., and Lindsley, D. (1992). Leftward ribosome frameshifting at a hungry codon. J. Mol.
Biol. 223, 31–40.

Garcia-Miranda, P., Becker, J.T., Benner, B.E., Blume, A., Sherer, N.M., and Butcher, S.E. (2016).
Stability of HIV Frameshift Site RNA Correlates with Frameshift Efficiency and Decreased Virus
Infectivity. J. Virol. 90, 6906–6917.

Girstmair, H., Saffert, P., Rode, S., Czech, A., Holland, G., Bannert, N., and Ignatova, Z. (2013).
Depletion of Cognate Charged Transfer RNA Causes Translational Frameshifting within the
Expanded CAG Stretch in Huntingtin. Cell Rep. 3, 148–159.

Halma, M.T.J., Ritchie, D.B., Cappellano, T.R., Neupane, K., and Woodside, M.T. (2019). Complex
dynamics under tension in a high-efficiency frameshift stimulatory structure. Proc. Natl. Acad. Sci. U.
S. A.

Hemelaar, J., Gouws, E., Ghys, P.D., and Osmanov, S. (2006). Global and regional distribution of
HIV-1 genetic subtypes and recombinants in 2004. AIDS Lond. Engl. 20, W13-23.

Hofacker, I.L., Priwitzer, B., and Stadler, P.F. (2004). Prediction of locally stable RNA secondary
structures for genome-wide surveys. Bioinforma. Oxf. Engl. 20, 186–190.

Hung, M., Patel, P., Davis, S., and Green, S.R. (1998). Importance of ribosomal frameshifting for
human immunodeficiency virus type 1 particle assembly and replication. J. Virol. 72, 4819–4824.

Ivanov, I.P., and Atkins, J.F. (2007). Ribosomal frameshifting in decoding antizyme mRNAs from
yeast and protists to humans: close to 300 cases reveal remarkable diversity despite underlying
conservation. Nucleic Acids Res. 35, 1842–1858.

Ivanov, I.P., Gesteland, R.F., and Atkins, J.F. (1998). A second mammalian antizyme: conservation
of programmed ribosomal frameshifting. Genomics 52, 119–129.

Ivanov, I.P., Matsufuji, S., Murakami, Y., Gesteland, R.F., and Atkins, J.F. (2000). Conservation of
polyamine regulation by translational frameshifting from yeast to mammals. EMBO J. 19, 1907–1917.

Ivanov, I.P., Gesteland, R.F., and Atkins, J.F. (2006). Evolutionary specialization of recoding:
Frameshifting in the expression of S. cerevisiae antizyme mRNA is via an atypical antizyme shift site
but is still +1. RNA 12, 332–337.

Jacks, T., and Varmus, H.E. (1985). Expression of the Rous sarcoma virus pol gene by ribosomal
frameshifting. Science 230, 1237–1242.

Jacks, T., Power, M.D., Masiarz, F.R., Luciw, P.A., Barr, P.J., and Varmus, H.E. (1988a).
Characterization of ribosomal frameshifting in HIV-1 gag-pol expression. Nature 331, 280–283.

Jacks, T., Madhani, H.D., Masiarz, F.R., and Varmus, H.E. (1988b). Signals for ribosomal
frameshifting in the Rous sarcoma virus gag-pol region. Cell 55, 447–458.

Janssen, S., and Giegerich, R. (2015). The RNA shapes studio. Bioinformatics 31, 423–425.

Ketteler, R. (2012). On programmed ribosomal frameshifting: the alternative proteomes. Front. Genet.
3, 242.

Kleinkauf, R., Houwaart, T., Backofen, R., and Mann, M. (2015a). antaRNA – Multi-objective inverse
folding of pseudoknot RNA using ant-colony optimization. BMC Bioinformatics 16, 389.

Kleinkauf, R., Mann, M., and Backofen, R. (2015b). antaRNA: ant colony-based RNA sequence
design. Bioinformatics 31, 3114–3121.

Kollmus, H., Honigman, A., Panet, A., and Hauser, H. (1994). The sequences of and distance between
two cis-acting signals determine the efficiency of ribosomal frameshifting in human immunodeficiency
virus type 1 and human T-cell leukemia virus type II in vivo. J. Virol. 68, 6087–6091.

Korniy, N., Goyal, A., Hoffmann, M., Samatova, E., Peske, F., Pöhlmann, S., and Rodnina, M.V.
(2019). Modulation of HIV-1 Gag/Gag-Pol frameshifting by tRNA abundance. Nucleic Acids Res. 47,
5210–5222.

Kurian, L., Palanimurugan, R., Gödderz, D., and Dohmen, R.J. (2011). Polyamine sensing by nascent
ornithine decarboxylase antizyme stimulates decoding of its mRNA. Nature 477, 490–494.

Lin, Z., Gilbert, R.J.C., and Brierley, I. (2012). Spacer-length dependence of programmed -1 or -2
ribosomal frameshifting on a U6A heptamer supports a role for messenger RNA (mRNA) tension in
frameshifting. Nucleic Acids Res. 40, 8674–8689.

Lorenz, R., Bernhart, S.H., Höner Zu Siederdissen, C., Tafer, H., Flamm, C., Stadler, P.F., and
Hofacker, I.L. (2011). ViennaRNA Package 2.0. Algorithms Mol. Biol. AMB 6, 26.

Lundberg, S.M., Erion, G., Chen, H., DeGrave, A., Prutkin, J.M., Nair, B., Katz, R., Himmelfarb, J.,
Bansal, N., and Lee, S.-I. (2020). From local explanations to global understanding with explainable AI
for trees. Nat. Mach. Intell. 2, 56–67.

Lyon, K., Aguilera, L.U., Morisaki, T., Munsky, B., and Stasevich, T.J. (2019). Live-Cell Single RNA
Imaging Reveals Bursts of Translational Frameshifting. Mol. Cell 75, 172-183.e9.

Manktelow, E., Shigemoto, K., and Brierley, I. (2005). Characterization of the frameshift signal of Edr,
a mammalian example of programmed -1 ribosomal frameshifting. Nucleic Acids Res. 33, 1553–1563.

Marcheschi, R.J., Staple, D.W., and Butcher, S.E. (2007). Programmed Ribosomal Frameshifting in
SIV is Induced by a Highly Structured RNA Stem-Loop. J. Mol. Biol. 373, 652–663.

Marczinke, B., Fisher, R., Vidakovic, M., Bloys, A.J., and Brierley, I. (1998). Secondary structure and
mutational analysis of the ribosomal frameshift signal of rous sarcoma virus. J. Mol. Biol. 284, 205–
225.

Matsufuji, S., Matsufuji, T., Miyazaki, Y., Murakami, Y., Atkins, J.F., Gesteland, R.F., and Hayashi, S.
(1995). Autoregulatory frameshifting in decoding mammalian ornithine decarboxylase antizyme. Cell
80, 51–60.

Melian, E.B., Hinzman, E., Nagasaki, T., Firth, A.E., Wills, N.M., Nouwens, A.S., Blitvich, B.J., Leung,
J., Funk, A., Atkins, J.F., et al. (2010). NS1′ of Flaviviruses in the Japanese Encephalitis Virus
Serogroup Is a Product of Ribosomal Frameshifting and Plays a Role in Viral Neuroinvasiveness. J.
Virol. 84, 1641–1647.

Melian, E.B., Hall-Mendelin, S., Du, F., Owens, N., Bosco-Lauth, A.M., Nagasaki, T., Rudd, S., Brault,
A.C., Bowen, R.A., Hall, R.A., et al. (2014). Programmed ribosomal frameshift alters expression of
west nile virus genes and facilitates virus replication in birds and mosquitoes. PLoS Pathog. 10,
e1004447.

Meydan, S., Klepacki, D., Karthikeyan, S., Margus, T., Thomas, P., Jones, J.E., Khan, Y., Briggs, J.,
Dinman, J.D., Vázquez-Laslop, N., et al. (2017). Programmed Ribosomal Frameshifting Generates a
Copper Transporter and a Copper Chaperone from the Same Gene. Mol. Cell 65, 207–219.

Mikl, M., Hamburg, A., Pilpel, Y., and Segal, E. (2019). Dissecting splicing decisions and cell-to-cell
variability with designed sequence libraries. Nat. Commun. 10, 4572.

Moomau, C., Musalgaonkar, S., Khan, Y.A., Jones, J.E., and Dinman, J.D. (2016). Structural and
Functional Characterization of Programmed Ribosomal Frameshift Signals in West Nile Virus Strains
Reveals High Structural Plasticity Among cis-Acting RNA Elements. J. Biol. Chem. 291, 15788–
15795.

Moon, S., Byun, Y., and Han, K. (2007). FSDB: a frameshift signal database. Comput. Biol. Chem.
31, 298–302.

Mouzakis, K.D., Lang, A.L., Vander Meulen, K.A., Easterday, P.D., and Butcher, S.E. (2013). HIV-1
frameshift efficiency is primarily determined by the stability of base pairs positioned at the mRNA
entrance channel of the ribosome. Nucleic Acids Res. 41, 1901–1913.

Nam, S.H., Copeland, T.D., Hatanaka, M., and Oroszlan, S. (1993). Characterization of ribosomal
frameshifting for expression of pol gene products of human T-cell leukemia virus type I. J. Virol. 67,
196–203.

Nikolic, E.I.C., King, L.M., Vidakovic, M., Irigoyen, N., and Brierley, I. (2012). Modulation of ribosomal
frameshifting frequency and its effect on the replication of Rous sarcoma virus. J. Virol. 86, 11581–
11594.

Ono, M., Yasunaga, T., Miyata, T., and Ushikubo, H. (1986). Nucleotide sequence of human
endogenous retrovirus genome related to the mouse mammary tumor virus genome. J. Virol. 60, 589–
598.

Parkin, N.T., Chamorro, M., and Varmus, H.E. (1992). Human immunodeficiency virus type 1 gag-pol
frameshifting is dependent on downstream mRNA secondary structure: demonstration by expression
in vivo. J. Virol. 66, 5147–5151.

Plant, E.P., Pérez-Alvarado, G.C., Jacobs, J.L., Mukhopadhyay, B., Hennig, M., and Dinman, J.D.
(2005). A three-stemmed mRNA pseudoknot in the SARS coronavirus frameshift signal. PLoS Biol.
3, e172.

Pollom, E., Dang, K.K., Potter, E.L., Gorelick, R.J., Burch, C.L., Weeks, K.M., and Swanstrom, R.
(2013). Comparison of SIV and HIV-1 Genomic RNA Structures Reveals Impact of Sequence
Evolution on Conserved and Non-Conserved Structural Motifs. PLOS Pathog. 9, e1003294.

Ritchie, D.B., Foster, D.A.N., and Woodside, M.T. (2012). Programmed -1 frameshifting efficiency
correlates with RNA pseudoknot conformational plasticity, not resistance to mechanical unfolding.
Proc. Natl. Acad. Sci. U. S. A. 109, 16167–16172.

Ritchie, D.B., Cappellano, T.R., Tittle, C., Rezajooei, N., Rouleau, L., Sikkema, W.K.A., and
Woodside, M.T. (2017). Conformational dynamics of the frameshift stimulatory structure in HIV-1.
RNA N. Y. N 23, 1376–1384.

Rom, E., and Kahana, C. (1994). Polyamines regulate the expression of ornithine decarboxylase
antizyme in vitro by inducing ribosomal frame-shifting. Proc. Natl. Acad. Sci. U. S. A. 91, 3959–3963.

Staple, D.W., and Butcher, S.E. (2003). Solution structure of the HIV-1 frameshift inducing stem-loop
RNA. Nucleic Acids Res. 31, 4326–4331.

Su, M.-C., Chang, C.-T., Chu, C.-H., Tsai, C.-H., and Chang, K.-Y. (2005). An atypical RNA
pseudoknot stimulator and an upstream attenuation signal for -1 ribosomal frameshifting of SARS
coronavirus. Nucleic Acids Res. 33, 4265–4275.

Theis, C., Reeder, J., and Giegerich, R. (2008). KnotInFrame: prediction of -1 ribosomal frameshift
events. Nucleic Acids Res. 36, 6013–6020.

Thiel, V., Ivanov, K.A., Putics, A., Hertzig, T., Schelle, B., Bayer, S., Weissbrich, B., Snijder, E.J.,
Rabenau, H., Doerr, H.W., et al. (2003). Mechanisms and enzymes involved in SARS coronavirus
genome expression. J. Gen. Virol. 84, 2305–2315.

Tholstrup, J., Oddershede, L.B., and Sørensen, M.A. (2012). mRNA pseudoknot structures can act
as ribosomal roadblocks. Nucleic Acids Res. 40, 303–313.

Tosaka, Y., Tanaka, H., Yano, Y., Masai, K., Nozaki, M., Yomogida, K., Otani, S., Nojima, H., and
Nishimune, Y. (2000). Identification and characterization of testis specific ornithine decarboxylase
antizyme (OAZ-t) gene: expression in haploid germ cells and polyamine-induced frameshifting. Genes
Cells Devoted Mol. Cell. Mech. 5, 265–276.

Tu, C., Tzeng, T.H., and Bruenn, J.A. (1992). Ribosomal movement impeded at a pseudoknot
required for frameshifting. Proc. Natl. Acad. Sci. U. S. A. 89, 8636–8640.

Vainberg Slutskin, I., Weingarten-Gabbay, S., Nir, R., Weinberger, A., and Segal, E. (2018).
Unraveling the determinants of microRNA mediated regulation using a massively parallel reporter
assay. Nat. Commun. 9.

Waldman, Y.Y., Tuller, T., Shlomi, T., Sharan, R., and Ruppin, E. (2010). Translation efficiency in
humans: tissue specificity, global optimization and differences between developmental stages.
Nucleic Acids Res. 38, 2964–2974.

Wang, X., Xuan, Y., Han, Y., Ding, X., Ye, K., Yang, F., Gao, P., Goff, S.P., and Gao, G. (2019).
Regulation of HIV-1 Gag-Pol Expression by Shiftless, an Inhibitor of Programmed -1 Ribosomal
Frameshifting. Cell 176, 625-635.e14.

Wang, Y., Wills, N.M., Du, Z., Rangan, A., Atkins, J.F., Gesteland, R.F., and Hoffman, D.W. (2002).
Comparative studies of frameshifting and nonframeshifting RNA pseudoknots: a mutational and NMR
investigation of pseudoknots derived from the bacteriophage T2 gene 32 mRNA and the retroviral
gag-pro frameshift site. RNA N. Y. N 8, 981–996.

Weingarten-Gabbay, S., Elias-Kirma, S., Nir, R., Gritsenko, A.A., Stern-Ginossar, N., Yakhini, Z.,
Weinberger, A., and Segal, E. (2016). Systematic discovery of cap-independent translation
sequences in human and viral genomes. Science 351, aad4939.

Weingarten-Gabbay, S., Nir, R., Lubliner, S., Sharon, E., Kalma, Y., Weinberger, A., and Segal, E.
(2019). Systematic interrogation of human promoters. Genome Res. 29, 171–183.

Weiss, R., Lindsley, D., Falahee, B., and Gallant, J. (1988). On the mechanism of ribosomal
frameshifting at hungry codons. J. Mol. Biol. 203, 403–410.

van Weringh, A., Ragonnet-Cronin, M., Pranckeviciene, E., Pavon-Eternod, M., Kleiman, L., and Xia,
X. (2011). HIV-1 modulates the tRNA pool to improve translation efficiency. Mol. Biol. Evol. 28, 1827–
1834.

Wills, N.M., Moore, B., Hammer, A., Gesteland, R.F., and Atkins, J.F. (2006). A functional -1 ribosomal
frameshift signal in the human paraneoplastic Ma3 gene. J. Biol. Chem. 281, 7082–7088.

Wu, B., Zhang, H., Sun, R., Peng, S., Cooperman, B.S., Goldman, Y.E., and Chen, C. (2018).
Translocation kinetics and structural dynamics of ribosomes are modulated by the conformational
plasticity of downstream pseudoknots. Nucleic Acids Res. 46, 9736–9748.

Yan, S., Wen, J.-D., Bustamante, C., and Tinoco, I. (2015). Ribosome excursions during mRNA
translocation mediate broad branching of frameshift pathways. Cell 160, 870–881.

Yordanova, M.M., Wu, C., Andreev, D.E., Sachs, M.S., and Atkins, J.F. (2015). A Nascent Peptide
Signal Responsive to Endogenous Levels of Polyamines Acts to Stimulate Regulatory Frameshifting
on Antizyme mRNA. J. Biol. Chem. 290, 17863–17878.

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