Quantitative analysis of transcription start site selection in *Saccharomyces*
***cerevisiae* reveals control by DNA sequence, RNA Polymerase II activity, and NTP**
levels

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Abstract

Transcription start site (TSS) selection is a key step in gene expression and occurs at
many promoter positions over a wide range of efficiencies. Here, we develop a
massively parallel reporter assay to quantitatively dissect contributions of promoter
sequence, NTP substrate levels, and RNA polymerase II (Pol II) activity to TSS
selection by "promoter scanning" in *Saccharomyces cerevisiae* (Pol II MAssively
Systematic Transcript End Readout, "Pol II MASTER"). Using Pol II MASTER, we
measure the efficiency of Pol II initiation at 1,000,000 individual TSS sequences in a
defined promoter context. Pol II MASTER confirms proposed critical qualities of *S.*
cerevisiae TSS -8, -1, and +1 positions quantitatively in a controlled promoter context.
Pol II MASTER extends quantitative analysis to surrounding sequences and finds they
tune initiation over a wide range of efficiencies. These results enable the development
of a predictive model for initiation efficiency based on sequence. We show that genetic
perturbation of Pol II catalytic activity alters initiation efficiency mostly independently of
TSS sequence, but selectively modulates preference for initiating nucleotide.
Intriguingly, we find that Pol II initiation efficiency is directly sensitive to GTP levels at
the first five transcript positions and to CTP and UTP levels at the second position
genome-wide. These results suggest individual transcripts will be differentially sensitive
to alterations in NTP levels, representing a cryptic layer of potential regulation at the
level of Pol II biochemical properties. The results establish Pol II MASTER as a method
for quantitative dissection of transcription initiation in eukaryotes.

Introduction

In transcription initiation, RNA polymerase II (Pol II) binds promoter DNA through interactions with core promoter elements, unwinds a turn of promoter DNA forming a Pol II-promoter open complex containing a single-stranded "transcription bubble", and selects a promoter position within a region competent for initiation to serve as the transcription start site (TSS). At the majority of Pol II promoters in eukaryotes, TSS selection occurs at multiple positions [1-11]. Thus, the overall rate of gene expression at the majority of Pol II promoters is determined by the efficiency of initiation from several distinct TSS positions. In addition, studies have suggested that alternative TSS selection can lead to differences in mRNA features, translation activity and subsequential protein levels and functions [3,12,13], and therefore is widespread in different cell types [6], developmental processes [4,12,14,15], growth conditions [16], responses to environmental changes, and cancers [17-19].

Pol II initiation in yeast proceeds by a promoter scanning mechanism, where the Pol II Pre-initiation Complex (PIC), comprising Pol II and initiation factors, assembles upstream of the initiation region and then scans downstream to select a TSS position [20-27]. The efficiency of initiation at a given position within a region allowing initiation depends on multiple features. Location relative to the core promoter region, with bases scanned from upstream to downstream from the core promoter, is critically important, as is DNA sequence, with the template base specifying the TSS position and position immediately upstream of the TSS (positions +1 and -1, respectively) making the largest contributions. In particular, there is a strong preference for an R:Y base pair at position +1 and Y:R base pair at position -1 (reflected as a $Y_{-1}R_{+1}$ "initiator" sequence on the coding strand; Y=pYrimidine; R=puRine), and this preference is near universal for RNAPs[1,10,11,16,21,22,28-44].

It has long been recognized that DNA sequences at positions beyond the -1/+1 also contribute to the efficiency of initiation by Pol II. However, the sequence preferences at these positions are difficult to determine from genomic usage alone given differences in context, might extend to higher-order interactions between positions, and likely include species-specific determinants. For example, in *S. cerevisiae*, genome-wide transcriptomic data have revealed a preference for an A:T base pair at position -8, reflected as an A on the transcribed strand and a T on the template strand [1,16,21,34,35]. Furthermore, promoters with high expression and/or with initiation focused primarily in a single, efficient TSS tend to show additional A enrichment at positions -7 to -3 on the transcribed strand (T on the template strand) [1,16,21,45,46]. Several factors confound efforts to directly measure the contribution of DNA sequence to the efficiency of initiation by Pol II including promoter chromatin context and TSS position within a promoter. These attributes together might be considered a promoter's architecture. Promoter architectural factors are especially apparent in yeast – where initiation occurs by promoter scanning – because TSSs are proposed to be selected in a polar fashion from upstream to downstream.

An elegant and important study of promoter scanning from Kuehner and Brow established that TSS usage is determined by TSS priority during the scanning process

[22]. As noted above, scanning proceeds directionally from upstream sequences to those
downstream. Therefore, sequences are examined by the transcription machinery in the
order in which they are scanned with upstream TSSs having priority, regardless of
innate TSS strength. Kuehner and Brow introduced the concept of "TSS efficiency,"
which accounts for how much Pol II reaches a particular TSS in order to determine
innate TSS strength (see **Figure 1**). TSS efficiency is defined as the usage of a
particular TSS normalized to the usage of that position and all downstream positions.
Downstream TSS usage is used to estimate the flux of scanning Pol II that scans past a
particular TSS. This measure is essential for comparison of TSSs embedded in different
promoter contexts and different positions relative to other TSSs.

Imbalanced promoter sequence distributions imposed by evolutionary processes also
limit the ability to determine sequence-activity relationships for initiation. For example, it
has been observed that yeast promoters have uneven distribution of bases across
promoters and this is most obvious in the enrichment for T on the transcribed strand
upstream of the median TSS position (reflecting the middled of the TSS distribution) and
A on the transcribed strand downstream of the median TSS position at highly expressed
promoters, and a paucity of G/C content in general [21,45-48]. Therefore, the biased
distribution of bases in yeast promoters leads to a biased distribution of TSS sequence
motifs, where preferred TSS motifs, due to increased probability of a -8A in additional to
evolutionary selection for it, show enrichment downstream of the median TSS position,
and less-preferred motifs show enrichment upstream of the median TSS position [21]. We
have found that hyperactive Pol II mutants ("gain-of-function" or GOF) and hypoactive
Pol II mutants ("loss-of-function" or LOF) were observed to shift TSS usage distributions
upstream or downstream, respectively, with differences in aggregate usage of particular
TSS motifs. Differences in usage due to biased promoter sequence distribution cannot
be disentangled from actual differences in TSS efficiencies. Moreover, other properties
such as the biochemistry of transcription itself due to availability of NTP substrates, the
biochemical properties of scanning processivity [20,25,27], promoter identity [49-51], or
promoter chromatin could also contribute to initiation output [52]. Together, all of these
factors would make contributions of primary DNA sequence more difficult to ascertain.

In order to remove contextual differences among promoters we have developed a
system to dissect determinants of initiation efficiency within a defined, controlled
promoter context. Here, we develop "Pol II MASTER" based on bacterial MASTER
(MAssively Systematic Transcript End Readout) [42,53-56], which allows determination of
initiation at base pair resolution and attribution of RNA transcripts to nucleic acid
barcoded promoter variants in massively parallel fashion. We apply Pol II MASTER to
initiation by promoter scanning to investigate the initiation efficiency of ~80,000
promoter variants within an appropriate TSS region in Pol II WT and catalytic mutants
and upon manipulation of NTP levels. We show that this system enables determination
of the interface between initiation factor activity, transcription substrates, promoter
sequence, and promoter output. We recapitulate the large impact of known base
positions relative to the TSS on initiation efficiency (-8, -1/+1) while revealing the wide
range of effects other positions have in initiation (-11 to -9, -7 to -2 and +2 to +4). We
identify a distinct hierarchy in $Y_{-1}R_{+1}$ preferences, and detect interactions between
bases at neighboring positions, suggesting potential mechanistic coupling between

positions. We find that Pol II mutant classes increase or decrease initiation efficiencies
for all possible sequences. This result is consistent with predictions that the primary
effects of altered TSS selection (directional shifts in TSS distributions) are driven by
initiation efficiency changes across all sequences and not on TSS sequence selection
per se. Our results further show that Pol II activity level does contribute to selective
efficiency of initiation at sequences +1A vs +1G and find that initiation efficiency is
selectively sensitive to individual NTP levels across genomic TSSs. Our findings
demonstrate that Pol II MASTER provides a quantitative platform for how initiation factor
activity together with promoter sequence contribute to Pol II transcription initiation in
vivo.

Results

A high-throughput system for studying TSS sequence effects on Pol II initiation
output

TSSs are identified at yeast promoters by scanning from upstream near the core-
promoter to downstream (**Figure 1A**). While promoters are melted starting around +20
from the TATA box (if present), there is a distance restriction that reduces usage of
TSSs until they are within a region ~40-140 nt downstream from the core promoter [57]
(**Figure 1B**). Processivity of the scanning process, likely determined by TFIIH activity, is
expected to limit TSS usage on the downstream edge of promoters (**Figure 1A**,
"unreachable TSS"). Once initiation happens at any site within this window of
opportunity for initiation (**Figure 1A, B**), Pol II "flux" – the amount of Pol II progressing to
TSSs downstream – is necessarily reduced. Given these potential variables in TSS
context and the "first come-first served" basis of promoter scanning, determining how
some TSSs are stronger than others is difficult if only considering TSS usage at
genomic promoters. In order to specifically dissect how TSS sequence controls initiation
efficiency and how sequence interacts with Pol II catalytic activity or NTP levels in a
controlled context, we have established a massively parallel promoter variant assay "Pol
II MASTER" where we have embedded almost all possible sequences within a 9 bp
randomized TSS region (**Figure 1C**) constructed on plasmids and introduced into yeast
strains with wild type (WT) or mutated Pol II. The sequence libraries constructed are
illustrated in **Figure 1C** and are referred to by their base compositions relative to the
transcribed strand unless otherwise specifically noted. The libraries are named by
identities of bases at positions -8, -1, and +1. The "AYR" library has composition $A_{-8}NNNNNNY_{-1}R_{+1}$ (N=A, C, G, or T, Y=C or T, R=A or G) relative to the transcribed
strand, with "BYR" having composition $B_{-8}NNNNNNY_{-1}R_{+1}$ (B=C, G, or T), *etc*. Our three
libraries comprise 81,920 promoter variants, allowing up to 983,040 individual TSS
sequences to be evaluated, as each promoter variant may allow initiation from different
positions within or right after the randomized region.

This randomized TSS region was inserted into a controlled promoter context containing
specific functionalities (**Figure 1C**). First, the *GAL1* UAS was utilized to allow
expression control of libraries. Second, the TATA-box-to-TSS region of *SNR37* gene
was used as a "scanning region" to direct initiation within the downstream randomized

TSS region. This is because almost no RNA 5′ ends are observable from this scanning
region within its normal promoter context. <u>Third</u>, the native, highly-efficient *SNR37* TSS
region was inserted downstream of the randomized TSS region as a "Flux Detector"
(FD). Here, we employ the approach of Kuehner and Brow that a highly efficient
initiation region placed downstream an upstream TSS region may capture any
polymerases that happen to scan past the randomized TSS region. As a metric to
compare strength of different TSSs, we employ "TSS efficiency" as defined by Kuehner
and Brow [22], which measures the usage of a TSS relative to usage at that TSS and all
downstream positions (**Figure 1C**). This metric serves two key purposes: it allows
upstream and downstream starts to be compared as it takes priority effects into account
(upstream TSSs reduce the amount of polymerases scanning to downstream sites), and
it allows normalization of promoter usage using RNA only, engendering comparison
across promoters and libraries. <u>Fourth</u>, a 24 bp DNA barcode containing 20 positions of
randomized bases and 4 interspersed fixed bases (to exclude low-complexity
sequences) allows RNA products to be assigned to respective promoter variants. An
RNA barcode is critical as bases upstream of the TSS will not be present in the
transcribed RNA yet still contribute to TSS efficiency. <u>Fifth</u>, the *GFP* coding region and
CYC1 terminator were added to support termination and stabilize RNA products.
Libraries were constructed by PCR sewing followed by cloning into a plasmid backbone.
After amplification in *E. coli*, plasmid libraries were transformed into Pol II WT and
mutant yeast strains in triplicate. Library expression was induced by addition of
galactose to the medium (4% final) for three hours. Both plasmid DNA and RNA
products were extracted from harvested yeast cells and amplified for DNA-seq and
TSS-seq (**Figure S1A**).

Several measures indicate high level of reproducibility and coverage depth of library
variants (**Figure 1D, Figure S1**). Base coverage in the randomized region was highly
even (**Figure S1B**). Correlation analysis for DNA-seq reads per promoter variant
suggested that yeast transformation did not alter promoter variant distribution (**Figure
S1C, D**). Bulk primer extension of libraries illustrated their average behavior and the
amount of initiation deriving from the randomized region (**Figure S1E**). As designed,
only a very small fraction of initiation was generated from the barcode region or further
downstream, validating that the flux detector captured scanning polymerases (**Figure
S1E, F**). Aggregate distribution of reads in our three libraries shows that as TSSs
decreased in efficiency from the most efficient library ("AYR") to least ("ARY") reads
shift from the designed +1 TSS to downstream positions (**Figure 1D**, left). The apparent
shift of TSS usage to position -1 in the ARY library was because the purine at the
designed -1 position serves as a +1 for newly created TSSs. **Figure 1D** (middle)
illustrates the aggregate TSS efficiency of each library based on usage at the designed
+1 TSS relative to usage at that position plus all downstream usage. Correlation
analysis for efficiency of library "major" TSSs (designed +1 TSS of "AYR" and "BYR"
libraries, +2 TSS of "ARY" library) demonstrated that biological replicates were highly
reproducible (**Figure 1D**, right). Therefore, we summed reads from the three biological
replicates, keeping TSSs that contained at least five TSS-seq reads in each replicate
and whose Coefficient of Variation (CV) in TSS-seq reads across replicates was less

than 0.5 (as a proxy for reproducible behavior) (**Figure S1G**). As a result, 97% of
possible TSS promoter variants were covered in each library on average (**Table S3**).

Sequence-dependent control of TSS efficiency in *S. cerevisiae*

To ask how our libraries recapitulated known TSS efficiency measurements, we first
examined core sequences in our library matching the *SNR14* TSS and its variants
examined by Kuehner and Brow [22] for TSS efficiency (**Figure 2A**). Our randomized
library contains the *SNR14* TSS sequence embedded in our *SNR37* context along with
all single substitution variants of this sequence, including the subset previously
examined at *SNR14.* We found that Pol II MASTER recapitulated the single base effects
on TSS efficiency previously observed while also indicating that single base changes
around a TSS can have large effects on TSS efficiency.

Examining all designed +1 TSS variants present in our library, we focused our analysis
on positions -8, -1 and +1, which *in vivo* genome-wide data suggested are important
determinants for TSS selection (**Figure 2B, Figure S2A**). We first determined the TSS
efficiencies of the designed +1 TSS for all promoter variants, dividing all variants into
groups defined by bases at positions -8, -1 and +1 relative to TSS. This allowed us to
examine the TSS efficiency distribution within each subgroup. Our results not only
recapitulated the importance of these three positions in our defined promoter context
but also demonstrated that surrounding positions have a considerable impact on TSS
efficiency (note the wide range of efficiencies within each TSS group, **Figure 2B**). <u>First</u>,
in our controlled context for TSSs at the designed +1 position, we found that $Y_{-1}R_{+1}$ was
essential for initiation above a minimal background relative to $R_{-1}Y_{+1}$. Even in the
presence of an A_{-8}, $R_{-1}Y_{+1}$ variants showed essentially no usage. <u>Second</u>, we quantified
the very large effect of a -8A on TSS efficiency (note that $A_{-8}Y_{-1}R_{+1}$ motifs were much
higher in efficiency in aggregate than non-$A_{-8}Y_{-1}R_{+1}$ TSSs), demonstrating that -8A
alters TSS efficiency. This result is in line with the observation that $A_{-8}C_{-1}A_{+1}$ motif-
containing TSSs have the highest aggregate TSS usage from genomic promoters and
appear to be the most efficient [21]. <u>Third</u>, among $Y_{-1}R_{+1}$ elements we found a clear
hierarchy of efficiency – $C_{-1}A_{+1} > C_{-1}G_{+1} > T_{-1}A_{+1} \approx T_{-1}G_{+1}$ – that was not apparent from
genomic promoter TSS usage or contexts, likely due to promoter sequence biases
within genomic promoters.

As noted above, a wide range of TSS efficiencies were observed within every -8/-1/+1
group, suggesting contributions of other sequence positions to initiation efficiency. To
determine if base identities within -7 to -2 positions have similar effects regardless of -8,
-1 and +1 identity, we rank ordered all individual TSSs for each $N_{-8}N_{-1}N_{+1}$ motif by the
efficiency of their $A_{-8}C_{-1}A_{+1}$ version (**Figure S2A**). This rank ordering by $A_{-8}C_{-1}A_{+1}$
efficiency was predictive of efficiency ranks for -7 to -2 sequences with different bases
at positions -8, -1 and +1. This observation indicates important contributions from
positions beyond positions -8, -1 and +1 and that positions might function independently
to determine TSS efficiency (examined below).

We set out to determine the contributions to Pol II TSS efficiency of individual bases at
each position across our randomized region relative to the designed +1 TSS. To do so,

we calculated the TSS efficiencies of examined TSS variants and divided them into
subgroups based on bases at individual positions relative to the designed TSS +1
position (**Figure S2B**). Comparison of TSS efficiencies across base subgroups
suggested significant individual base effects on TSS efficiency in aggregate at all
examined positions. The usage of other positions in our promoter outside of the
designed TSS +1 position (either within the randomized region or outside of it) allowed
us to examine contribution of bases in our randomized region to the efficiency of these
other TSSs (**Figure S1E, F, Figure 2C**). Our combined libraries allowed us to analyze
efficiencies of nearly a million TSS variants present within them (distribution of
efficiencies is shown in **Figure S2C**). In order to visualize sequence preferences, we
used the median initiation efficiency values of each base subgroup as indicators for
preference, using centered median values to calculate "relative efficiency" and
illustrated this preference in a sequence logo (**Figure 2D**). Datasets of designed +1
TSSs deriving from "AYR", "BYR", and "ARY" libraries allowed us to nearly
comprehensively study preference at positions -8 to +1 (**Figure 2E**, middle and bottom).
Additionally, 10-25% of total TSS usage among different libraries deriving from a TSS at
+4 (**Figure 2C**, brown TSS arrow and sequences) allowed study of positions -11 to -9
relative to this TSS (**Figure 2E**, top). As noted above, positions -8, -1 and +1 were
major determinants for TSS efficiency. Interestingly, position -9 showed a relatively
strong effect in our defined promoter context, which was not described in genome-wide
analyses. At positions -4 to -2, we observed modulation of initiation efficiency, where in
general C and/or G were preferred and T was less-preferred. This overall preference is
consistent with an observation where individually mutating Ts at positions -4 to -2
relative to -38 TSS of *ADH1* promoter to a C significantly increased usage of that TSS
[57]. Though preferences at positions -7 to -5 were statistically significant (**Figure S2B**),
contributions were much lower than other examined positions. Taken together, these
results indicate that positions -9 to -8 and -4 to +1 are two major clusters contributing to
TSS efficiency.

Experiments across species and the description of the canonical initiator element (Inr)
suggest some sequence contributions from downstream positions relative to the TSS
[10,43,44,57-61]. To determine the impact of sequences downstream of the TSS, we
examined motif enrichment of the most efficient -8 TSS variants, whose positions +1 to
+9 are located in the randomized region (**Figure 2C**, green TSS arrow and sequences).
We found an $A(/G)_{+2}G(/C)_{+3}G(/C)_{+4}$ enrichment for the top 10% most efficient TSS, but
not for the next 10% most efficient TSS (**Figure S2D**). These preferences are consistent
with early mutagenesis work in yeast, where A(+2) to C/T, G(+3) to T, or C(+4) to T
substitutions decreased the utilization of a particular TSS, but A(+2) to G or T(+5) to C
substitutions had minor effects [57].

Pairwise nucleotide-position dependencies have been observed for some processes, for
example in 5' splice sites [62-64]. To investigate potential higher order sequence
interactions *i.e.* coupling between positions contributing to TSS efficiency, we examined
all possible pairwise interactions among positions -11 to +1 (**Figure 2F, G, Figure S2E,
F**). "Coupling" would entail a base at one position determining the contribution or effect
of a base at another position. We found evidence for coupling at multiple positions with
the strongest coupling between -9 and -8 positions. Here, an A at one position

suppresses the preference of A at the other position indicative of positional epistasis at
each position where an A at one position might diminish the impact of an A at the other
(**Figure S2E**). Using this -9/-8 interaction as an example, **Figure 2F** shows how
coupling was detected and visualized. We calculated "centered relative efficiency"
values for each base and position and visualized them as a heatmap (**Figure 2F, G**).
We found that evident interactions were mainly observed at neighboring positions
(**Figure 2G**), especially within positions -9 to -7 and within positions -5 to -2. In addition
to its interaction with position -9 described above, position -8 was also observed to
interact with position -7 (**Figure S2F**). These observed interactions suggest that these
positions might work together in TSS selection (see Discussion).

Pol II mutants alter TSS efficiency for all possible TSS motifs while showing
selective effects at +1

Pol II mutants were observed previously to change apparent specificity for A_{-8} versus B_{-8}
(non-A) TSSs in opposite directions depending on Pol II defect as determined by
genomic TSS usage analysis [21]. As we noted in our prior work, this result could be a
consequence of Pol II GOF or LOF mutants shifting TSSs upstream or downstream,
respectively, coupled with the uneven distributions of bases and TSS motifs within
genomic promoters. Furthermore, TSS motifs with a -8A by definition have an increased
probability of having a potential upstream TSS that can be used (the -8A could function
as an upstream +1R TSS but might lack its own -8A). As a result, upstream-shifting
mutants seem to decrease preference for -8A TSSs because they are shifting from
these to use of the -8A as a +1A. To determine how Pol II catalytic activity affects TSS
selection in our controlled promoter context, we measured effects on TSS efficiency in
our promoter variant libraries between WT and Pol II mutants (**Figure 3, Figure S3**).
We introduced variant libraries into two Pol II LOF mutants (*rpb1* F1086S and H1085Q)
and two GOF mutants (*rpb1* E1103G and G1097D). Data for all mutants showed high
coverage of libraries (**Figure S3A**), were highly reproducible (**Figure S3B-E**), and
showed expected upstream or downstream shifts in TSS usage in aggregate across
libraries (**Figure S3B**) with broad effects on efficiencies across sequences (**Figure 3,**
Figure S3F-I). In our previous work [21] we speculated that Pol II mutants with decreased
catalytic activity (LOF) would decrease initiation across all sequences; conversely,
alleles with increased catalytic activity (GOF) would increase initiation across all
sequences. We find that this is indeed the case (**Figure 3A**). In our controlled promoter
context, we find that LOF mutants decreased efficiencies across all sequences, and
GOF mutants increased efficiencies across all sequences, and that mutants did not
show strong effects on -8 (as was observed for usage at genomic promoters [21]).
Instead, we observe differences in effects on +1A vs +1G (**Figure 3A**), and these
effects are apparent across all levels of innate TSS efficiency (**Figure 3B**). Both GOF
and LOF mutants show stronger effects on +1A relative to +1G, consistent with changes
to the Pol II active site altering activity for different initiating NTPs.

Pol II initiation for +1G TSSs is sensitive to changes in GTP levels

We hypothesized that the effects on initiating +1 NTP could be direct effects of Pol II
active site mutants altering activity in a substrate-selective fashion. However, it is also

conceivable that NTP levels within cells might be altered in Pol II mutants. To
investigate if the observed selective effects of Pol II mutants initiating NTP could be due
to altered NTP levels, we measured total NTP levels in WT, F1086S (LOF), and
E1103G (GOF) strains (**Figure 4A**). E1103G showed similar NTP levels as WT while
F1086S increased NTP levels to different extents. While both ATP and GTP in F1086S
were higher than WT, we reasoned that initiation might be more sensitive to GTP levels
as GTP is less in excess of the apparent K_D of Pol II for NTPs than is ATP. If so, this
provides an explanation for decreased deficit of +1G TSSs relative to +1A TSSs for
F1086S; increased GTP level in F1086S may buffer decreases in efficiency for +1G
TSSs.

If initiation is sensitive to GTP concentration in vivo, we hypothesized that reduction in
GTP should result in selectively reduced efficiency for +1G TSSs relative to +1A TSSs.
Therefore, we examined if and how initiation of our promoter variant libraries and
genomic TSSs was affected by treatment of cells with Mycophenolic Acid (MPA), which
is known to deplete GTP through inhibition of yeast inosine monophosphate
dehydrogenase (IMPDH) activities encoded by *IMD3* and *IMD4* [65,66]. We treated our
mixed AYR and BYR libraries with MPA for 20 min prior to Gal induction, while EtOH
treatment alone (the solvent for MPA) was used as a control. MPA decreased GTP level
as expected but also increased ATP, CTP, and UTP levels (**Figure 4B**).

MPA showed effects on promoter variant libraries in aggregate (**Figure S4A**) with an
overall downstream shift in TSS usage relative to EtOH alone. Comparison of replicates
indicated high reproducibility of MPA treatment (**Figure S4A-B**). When examining
specific sequences, we observed a striking and selective decrease in efficiency for
essentially all +1G TSSs, while +1A TSSs were almost entirely unaffected (**Figure 4C**).
MPA was previously shown to depress efficiency of +1G TSSs at *IMD2* as part of its
regulation by GTP levels [67]. Our results extend this observation to all +1G TSSs in our
MASTER library.

Pol II initiation efficiencies are altered genome-wide upon changes to NTP levels

To determine if these striking effects were observed in promoter contexts beyond our
designed library context, we performed TSS-seq on genome-derived RNAs from the
same samples that were analyzed for MASTER libraries. Examination of genomic TSSs
revealed effects of MPA on TSS efficiencies across the genome (**Figure 4D**). We
observed a depression of efficiency on average across all TSSs in MPA vs. EtOH
(vehicle) treatment. Given that we observed strong effects at +1G for MASTER libraries,
we speculated that there might also be effects on TSSs with +2G or potentially beyond,
especially because initial bond formation requires interaction with two substrates, and
subsequent initial elongation will lack stabilization of the product RNA that occurs with a
full-length (~9 nt) RNA:DNA hybrid. Examination of positions +2 through +6 indicates
that G positions through +5 are correlated with decreased initiation efficiency (**Figure
4D, S4D**). Conversely, we observed that a +2C or +2U was correlated with increased
efficiency relative to +2A or +2G. These results are consistent with the observed
increased CTP and UTP levels upon MPA treatment. We hypothesized that A positions
show decreased overall efficiency because most transcripts will have at least one G

within the first 5-6 nucleotides. To examine TSSs more carefully for G effects, we
examined TSSs lacking G entirely for the first 6 positions (**Figure 4E**). TSSs lacking a G
in the first 6 positions were more efficient on average than all TSSs in the presence of
MPA explaining some of the overall suppression of all TSS efficiencies by MPA. We
examined TSSs lacking a G in the first 6 positions and observed reduced effects on
efficiency from MPA treatment relative to all TSSs, consistent with this idea.
Furthermore, this analysis confirmed C and U effects relative to A at the +2 position.

Learned initiation preferences are predictive of TSS efficiencies at genomic promoters

To ask how sequence determinants identified here relate to natural promoters, we
compared our library-defined sequence efficiencies to TSS efficiencies observed at
genomic promoters (**Figure 5**). To limit potentially confounding factors for genomic
promoters, we focused on a single "median" TSS for each promoter in a defined set of
promoter windows. The median TSS is defined as the TSS representing the position of
50th percentile of reads within each promoter window [21], i.e. the TSS representing the
middle of the cumulative distribution function of promoter reads from upstream to
downstream. We found that Pol II sequence preference at positions around median
TSSs was mostly consistent with what we observed in our libraries (**Figure 2E, Figure
5A**). Efficient genomic TSSs appear enriched for A at positions -7 to -5. The A-richness
at positions between -10 to -3 and +5 to +10 has been noted in previous studies from
our lab [21] (**Figure 5B**) and others [1,16,34]. However, As at positions -7 to -5 appeared
neutral in our promoter libraries (**Figure 2E**). The observed A-richness in genome could
reflect selection *in vivo* for additional promoter properties, such as providing an easily
meltable DNA region, lower nucleosome occupancy or reflect a context dependent role
not reflected by our promoter libraries.

We find that the interaction between -9A and -8A discovered in our libraries is also
reflected in genomic TSSs (**Figure 5C**). We first grouped the median TSSs from the top
20% expressed promoters based on the base at position -9 or -8 and then examined
sequence enrichment at the other position. We observed that -9A decreased the
enrichment of -8A relative to that when position -9 is not A (**Figure 5C**, left). Moreover,
when A was absent from the position -8, much higher enrichment for -9A was observed
(**Figure 5C**, right). Together, consistent with -9/-8 interaction discovered in our
controlled context, in the genome the presence of an A at either position -9 or -8
suppresses enrichment of A at the other position within the group of most efficiently
used median TSSs. These results suggest that -9A may function in similar fashion as -
8A, but that -8A may be more effective and therefore has been evolutionarily favored
(see Discussion).

Regression modeling identifies key DNA sequences and interaction for TSS selection regulation

We have found that DNA sequences around the TSS not only additively but also
interactively contribute to TSS efficiency. To quantitatively identify key features
(sequences and interactions) for TSS efficiency, we compiled datasets deriving from all

libraries and predicted TSS efficiency from sequence information by logistic regression
coupled with a forward stepwise selection strategy (**Figure 6A-E, Figure S5A-C**). We
first compiled datasets generated from designed -8 to +2 and +4 TSSs deriving from
"AYR", "BYR", and "ARY" libraries (**Figure S5A**) and split data into training (80%) and
test (20%) sets. For each variant, its sequences at positions -11 to +9 were extracted as
potential predictors for TSS efficiency. We then used a forward stepwise strategy with a
5-fold Cross-Validation (CV) to select robust features (predictors). By evaluating model
performance with R^2, sequences at nine positions (positions -9 to -7 and -4 to +2) and
one interaction (-9/-8 interaction) were identified as robust features and selected for final
modeling (**Figure 6A**). It is worth noting that models with as few as three features –
sequences at positions -9, -8, -1 (or +1) – could explain 74.10% of TSS efficiency
variation.

The final model containing the most predictive features explained 91.60% of the
variance in TSS efficiency for WT test set (20% of total dataset) (**Figure 6B, Figure
S5B**). Principal Component Analysis (PCA) on variables of models trained with
individual replicates of WT and Pol II mutants shows large differences between WT and
Pol II mutants but very small differences between replicates (**Figure S5C**), indicating
that modeling captured features of different Pol II groups. We next asked whether the
features learned by modeling using compiled datasets were consistent with our previous
sequence preference analysis using selected representative datasets with the most
randomized bases. We centered additive variable values and visualized as a sequence
logo (**Figure 6C**). First, as expected, positions -1 and +1 were the major predictors,
however the influence of the -8 A did not appear as strong as in our previous preference
analysis. We suspected this might be because the -9/-8 interaction contribution was not
included. After adding the -9/-8 interaction term, we observed emergence of the position
-8 as an influential predictor (**Figure 6D, E**), which also emphasizes the contribution of
the -9/-8 interaction. Second, and importantly, modeling confirmed the +2A preference
observed in previous motif enrichment analysis using only the most efficient -8 TSS
variants (**Figure S2D**). The impact of position +2 is also evident when performing
Principal Component Analysis (PCA) to variables of models trained with WT or Pol II
mutant datasets (**Figure S5C**). The fact that sequences at position +2 are the top
contributing variables in 2nd principal component that distinguished G1097D and
E1103G is in agreement with differentially altered +2 sequence enrichment by Pol II
mutants (**Figure S3I**). These results suggest position +2 preference is altered by Pol II
activity changes, and that this position might work directly with the Pol II active site.

**Sequence defines TSS efficiency within a wider promoter context during initiation
by scanning**

To evaluate the extent to which DNA sequence around a TSS contributes to TSS
efficiency at genomic promoters, we compared the difference between observed and
model predicted efficiencies for all positions within known promoter windows or within
specific subgroups of known promoters (**Figure 6F-H, Figure S5D**). As expected, we
found most promoter positions showed low or no observed efficiency and were over-
predicted by sequence alone (**Figure 6F, Figure S5D**). This is because TSSs need to
be specified by a core promoter followed by scanning over a limited distance. Therefore,

features beyond TSS sequence, such as distance from the site of PIC assembly,
determine initiation in the genome. We therefore extracted observed median TSSs as
representatives of TSSs in contexts supporting efficient initiation to ask how our
sequence-based predictor functioned on genomic TSSs (**Figure 6G**). We also
separated median TSSs by promoter classes based on Taf1 enrichment [68], a proxy for
the two main types of promoters in yeast, and by promoter expression levels (**Figure
6H**). We observed good prediction performance at a wide range of TSSs indicating
sequence determinants identified in our limited promoter context contribute to TSS
efficiency in genomic promoter contexts. We observed increased performance for more
highly expressed promoters (Pearson r increased from 0.37-0.54 to 0.46-0.64) and
Taf1-depleted promoters (Pearson r increased from 0.37-0.50 to 0.54-0.64) (**Figure
6H**). Conceivably, highly expressed promoters may have evolved TSSs at optimal
distances from core promoters, and therefore may be similarly sensitive to sequence
effects. Alternatively, the use in our libraries of *GAL1* UAS and *SNR37* core promoter
elements, where both *GAL1* and *SNR37* are highly-expressed, Taf1-depleted
promoters, may share sequence sensitivities for TSSs from related promoters.

Discussion

Individual TSS sequences are critically important for initiation output but this had not
previously been systematically determined for eukaryotic promoters. Changes to
initiation factor activity might also alter initiation preferences due to functional if not
physical interactions with DNA sequence. Studies examining TSS selection have been
based on existing gene promoters *in vivo* and are subject to unknown biases or
confounding variables of different promoter contexts. Here we developed and employed
Pol II MASTER to systematically investigate ~1 million TSS sequences in wildtype or
Pol II mutant cells. This system allowed us to specifically and comprehensively study
TSS efficiencies in initiation by promoter scanning, without confounding effects from
other architectural features, such as variability in core promoter-TSS distances,
differences in promoter identities or chromatin configurations that may obfuscate
analyses of genomic TSSs. We find sequence variation at different positions around
TSS considerably tunes initiation efficiency in a predictable way and these contributions
are important for initiation efficiency at genomic promoters.

Combining results from this study and others, we suggest how TSS sequence
contribution for TSS selection works (**Figure 7**). We propose that two major sequence
position groups contribute to TSS selection: bases around the TSS and bases around
position -8. First, in promoter scanning, the TSS and adjacent bases interact with Pol II
active site, the 1st NTP or each other to facilitate stable binding of 1st NTP and
potentially the 2nd NTP to stimulate RNA synthesis. Sensitivity of initiation efficiency to
NTP levels for first through fifth transcription positions in vivo supports this idea. This
function of downstream positions would be in contrast to the concept of the initiator or
downstream elements functioning as part of the TFIID or PIC binding site, as has been
proposed for higher eukaryotes (see [69,70] and references therein). As the universal
initiating element, $Y_{-1}R_{+1}$ has been established to facilitate stable binding of the NTPs by
RNA polymerases via base stacking between R_{-1} from template DNA and the 1st purine
NTP [43,44]. Positions upstream of the TSS, such as positions -4 to -2, might contribute to

stabilize template DNA via base stacking or physical interaction with initiation factors
[44,71].

Our observation that Pol II mutants showed selective effects on base at position +1
(**Figure 3**) supports that position +1 is interpreted by the Pol II active site. Meanwhile,
the observation that Pol II initiation is sensitive to NTP pools (**Figure 4C, Figure S4A**)
further suggests a mechanism for cellular state to regulate initiation via alteration of
NTP levels for the first position and several subsequent transcript positions levels. For
GOF Pol II mutants, we propose that differential preference for ATP versus GTP is a
direct consequence of Pol II active site changes. Pol II has been observed to bypass
cyclobutane pyrimidine dimers (CPD) in vitro through addition of an untemplated A
across from CPD lesions and a GOF polymerase appears to promote this activity,
suggestion a potential increase in selectivity for ATP in some situations [72]. In contrast to
effects observed for GOF mutants we proposed that tested LOF mutants may have
selective effects on TSS efficiency for +1A versus +1G sites due to indirect effects on
NTP levels in cells. Such defects might result from altered synthesis of nucleotide
synthesis-related genes, a number of which are themselves sensitive to Pol II activity
[23,67,73-76]. We observed the LOF mutant F1086S indeed alters all NTP levels (**Figure
4A**). Specifically, the LOF mutant increased GTP level, which is consistent with previous
observations of GTP-related genes, *IMD2*, *IMD3* and *IMD4*. In Pol II LOF mutants, *IMD2*
is constitutively active [23,74], and overall expression at the mRNA level for *IMD* genes is
increased [23].

While it is unsurprising that levels of NTP substrates would alter initiation efficiency, it is
striking that differences can be observed genome wide due to a modest reduction or
increase in GTP, CTP, or UTP levels. These changes suggest that environmental
control of initiation may be used more broadly than anticipated and that distinct
transcripts will be differentially affected upon perturbation of NTP levels. Direct sensing
and regulation of promoter activity by NTP levels has been proposed for the *IMD2*
promoter in yeast [67,77]. At this promoter, under GTP replete conditions upstream +1G
TSSs are used to generate a non-productive set of transcripts that terminate before
transcription of the *IMD2* reading frame. Upon GTP limitation (or catalytic defects in Pol
II), productive initiation is observed at a downstream +1A TSS. Our results here argue
that nucleotide sequence content is likely sufficient to confer the observed regulation as
we find that most +1G TSSs in the genome are reduced for initiation efficiency upon
reduction in GTP levels. For select TSSs from *IMD2* that they tested, the Brow lab had
previously found that Gs at the first or second transcript positions could confer MPA
sensitivity to a TSS [67]. Here, we find this effect genome-wide and that it extends to
transcript +5 position for GTP and also find evidence for effects of UTP and CTP levels.

The positions +2 to +4 downstream of TSS might contribute to establish initial RNA and
NTP stability in Pol II active site as well (**Figure S2D**). We observed Pol II mutant
effects on those preferences (**Figure S3I, Figure S5C**), suggesting these positions
might function via directly interacting with the Pol II active site, as observed in
transcription structures in other species [43,44]. First, the T7-like single-subunit RNAP
family shows a base-specific interaction between +2 NTP and a residue in the middle of
the O helix, which was suggested to enhance formation of the first phosphodiester bond

[43]. Second, a structure of *de novo* transcription initiation complex in bacterial RNA
Polymerase showed multiple interactions between the 2nd NTP and its β' and β
subunits, whose eukaryotic counterparts are Rpb1 and Rpb2 [44]. Alternatively, the A/G-
rich $A(/G)_{+1}A(/G)_{+2}G_{+3}G_{+4}$ motif might be related to translocation. A study from the
Landick lab showed that when A/G comprised the RNA 3' end, the RNAP active site
favored the post-translocated state [78]. If A/G similarly affects translocation state within
first four bases in Pol II initiation, synthesis of the first few bases might be promoted.
Our results detecting sensitivity in initiation efficiency based on levels of NTPs specified
at positions downstream from +1 (**Figure 4, S4**) suggest potential abortive initiation
under conditions that slow phosphodiester bond formation at positions early in transcript
synthesis. Abortive initiation by yeast Pol II in vitro under NTP replete conditions is very
low in specific artificial initiation conditions [79]. However, these conditions did not involve
initiation by a scanning PIC with all GTFs or reduced NTP levels. Revisiting the
biochemical properties of Pol II during bona fide initiation by promoter scanning could
be valuable and our studies make specific predictions that can be tested.

The GTF TFIIB has been specifically implicated in the preference for bases near
position -8, where the TFIIB B-reader domain has been observed to directly interact in a
structure of a yeast Pol II-TFIIB complex [80]. Here, it is attractive to envision TFIIB
functioning as an anchor point to pause the scanning process to promote Pol II initiation
at a fixed distance downstream. Several observations support this proposed function.
First, we detected sequence interaction between positions -8 and -7 (**Figure 2G, Figure
S2F**). This is in line with the direct contact of -7T and -8T on the template strand and
TFIIB B-reader R64 and D69 observed in Pol II-TFIIB complex structure [80], which has
been proposed to hold TSSs in Pol II active site longer during scanning. Second, we
observed a strong interaction between positions -9 and -8, where the presence of an A
at either position suppressed the preference of A at the other position (**Figure 2F,
Figure S2E**). This -9/-8 interaction was also evident when examining genomic median
TSSs (**Figure 5C**). Taken together, we speculate that Ts around position -8 or -9 on the
template strand and TFIIB may pause the scanning process to facilitate the usage of
TSSs positioned 8 to 9 bases downstream. Moreover, we have shown that Pol II
catalytic mutants alter TSS efficiencies across all TSS sequences, without showing
alteration in preference for -8A (**Figure 3, Figure S3F**), suggesting that Pol II catalytic
activity is not responsible for -8 preference.

Whether or how DNA sequence surrounding TSSs is involved in other promoter
properties is another question. A bases at positions -7 to -5 were observed to be neutral
in our promoter libraries (**Figure 2E**), in contrast to A-enrichment observed for highly
expressed and focused genomic TSSs (**Figure 5A, B**) [1,16,21,34,45]. We speculate that
observed A-richness around TSS functions through other evolved promoter properties.
First, observed A-richness between positions -10 to -3, together with T-richness at
further upstream core promoter region [45,46], provides an easily meltable region for DNA
unwinding, perhaps facilitating transcription initiation in specific contexts. Second,
higher A/T content may function to lower nucleosome occupancy [81], because
appropriately periodic G/C dinucleotides promote nucleosome occupancy [82-85].
However, the base composition switch in highly expressed promoters from T- to A-
preponderance [21,45,86] indicates A-richness may have other roles depending on the

characteristics of the sequence itself. <u>Third</u>, A-richness may be left over from the
evolution of promoter scanning. A recent study of transcription initiation mechanism
investigated 12 yeast species and proposed that during evolution a A-rich region
upstream of TSS appeared first, then the specific -8A preference occurred [47]. Therefore,
the observed A preference at upstream positions of highly used and focused TSSs may
be leftovers of A-enrichment in those promoters during scanning evolution in addition to
promoter roles beyond TSS selection per se.

Our studies highlight the strength of approaches to minimize contextual factors by
isolating specific promoter attributes for study in high-throughput. Here we have
employed Pol II MASTER to the DNA sequence determinants of initiation efficiency
during Pol II scanning. It will be valuable to apply this systematic analysis to other
promoter architectural factors determining Pol II initiation, such as UAS identity, core
promoter-TSS distance and sequence composition within the scanning region. In
addition, we have found sequences downstream of TSSs contribute to TSS efficiency.
Therefore, expanding the randomized region is needed to refine our understanding of
sequence preference and potential longer range sequence interactions. Furthermore,
applying Pol II MASTER across initiation mutants and promoter variants will reveal
factor-sequence relationships and may allow initiation potential to be determined from
DNA sequence and genome location alone.

Methods

Yeast strains, plasmids, oligonucleotides and media

Yeast strains, plasmids and oligonucleotide sequences are described in **Table S1**. All oligonucleotides were obtained from IDT. Yeast strains used in this study were constructed as previously [21,23,73,87]. Briefly, plasmids containing *rpb1* mutants (G1097D, E1103G, F1086S, and H1085Q) were introduced by transformation into yeast strain CKY749 containing a chromosomal deletion of *RPO21/RPB1* but with a wild type *RPB1 URA3* plasmid, which was subsequently lost by plasmid shuffling. All plasmids and yeast strains are available upon request. Yeast media are following standard protocols [88]. YPD solid medium is made of yeast extract (1% w/v; BD), peptone (2% w/v; BD, 211677), bacto-agar (2% w/v; BD, 214010) and dextrose (2% w/v; VWR, VWRBK876) supplemented with adenine (0.15 mM; Sigma-Aldrich, A9126) and L-tryptophan (0.4 mM; Sigma-Aldrich T0254). Minimal media plates are synthetic complete ("SC") with amino-acids dropped out as appropriate as described in [88] with minor alterations as described in [23]: per standard batch formulation, adenine hemisulfate (Sigma-Aldrich, A9126) was 2 g, L-Leucine (Sigma-Aldrich, L8000) was 4 g, myo-inositol was 0.1 g, para-aminobenzoic acid (PABA) was 0.2 g.

Construction and transformation of plasmid libraries

A 9 nt randomized TSS region and 20 nt randomized barcodes, with 4 fixed bases inserted between every 4 nt (NNNNANNNNCNNNNTNNNNGNNNN), were separately synthesized by IDT as oligo pools with specific randomized positions using "hand mixing" for N positions to ensure even randomization and avoid bias during machine mixing of precursors during oligo synthesis. Together with other components including the *GAL1* UAS, *SNR37* core promoter, *SNR37* TSS region ("flux detector"), *GFP* ORF, and the *CYC1* terminator, template libraries were constructed by PCR sewing and cloned into pRSII413 (a gift from Steven Haase, Addgene plasmid #35450; http://n2t.net/addgene:35450; RRID:Addgene_35450) [89] by ligation (**Figure S1A**). Ligation products were transformed into *Escherichia coli* TOP10F' cells (Invitrogen) and grown on LB plates supplemented with carbenicillin (100 µg/ml) at high density. 200,000-500,000 colonies were collected from each library to maximize variant representation. Plasmid libraries were isolated from cells pellets using ZymoPURE II Plasmid Maxiprep Kit (Zymo Research, D4203) per manufacturer's instructions. Plasmid library pools were transformed into yeast strains with wildtype and mutated Pol II using chemical transformation and electroporation, respectively. For Pol II WT libraries, 500 ng plasmid pool per reaction was transformed following yeast high efficiency transformation protocol described in [90]. For Pol II mutant libraries, 2 µg plasmid pool per reaction was electroporated into Pol II mutant strains following yeast electroporation transformation protocol described in [91], with 50 µg single-stranded carrier DNA added. Transformants were grown on selective SC-His plates with 2% glucose as carbon source at high density. Three biological replicates were performed for each library and on average over two million colonies were collected for each replicate. Transformants scraped from densely plated transformation plates were inoculated into fresh SC-His medium with 2% raffinose (Amresco, J392) at 0.25 x 10^7 cells/ml and grown until 0.5-0.8

x 10^7 cells/ml, as determined by cell counting. Subsequently, galactose (Amresco,
0637) was added for three hours (4% final concentration) to induce library expression.
For MPA treated WT libraries, MPA (in 100% ethanol; Sigma-Aldrich, M3536) was
added for 20 min (20 µg/ml final concentration) prior to the three-hour galactose
induction, with same volume of 100% ethanol as solvent control. 50 ml and 5 ml culture
aliquots, for RNA and DNA extraction respectively, were harvested and then cell pellets
were stored at -80 °C for further processing as described below.

Generation of DNA amplicons for DNA-seq

Plasmid DNA from yeast cell pellets was isolated using YeaStar Genomic DNA Kit
(Zymo Research, D2002) per manufacturer's instructions. Amplicon pools containing
the TSS and barcode regions were generated using plasmid DNA from *E.coli* or yeast
by Micellula DNA Emulsion & Purification (ePCR) Kit (EURx/CHIMERx, 3600) per
manufacturer's instructions. To minimize amplification bias, each sample was amplified
in a 15-cycle ePCR reaction, purified and subject to an additional 10-cycle scale-up
ePCR reaction. To create the necessary sequence template diversity for Illumina
sequencing, 18-25 bp and 1-7 bp "stuffer" sequences were added to 5'- and 3'-ends,
respectively, during amplicon preparation. Amplicon pools were subject to Illumina
NovaSeq 6000 (150 PE) sequencing, and on average 20 M paired-end reads were
obtained from each replicate of a sample, with high reproducibility and minimal
perturbation of the variant distribution with each library (**Table S2**).

Sample preparation for TSS-seq

Total RNA was extracted by a phenol-chloroform method [92], followed by RNA
purification (RNeasy Mini kit, QIAGEN, 74104) with on-column DNase digestion
(RNase-Free DNase Set, QIAGEN, 79254) to remove DNA. TSS-seq was done using
procedures described in [93]. To prepare RNAs for the cDNA library construction, samples
were sequentially treated with Terminator 5'-Phosphate-Dependent Exonuclease
(Lucigen), Quick CIP (calf-intestine alkaline phosphatase, NEB) and Cap-Clip™ Acid
Pyrophosphatase (CellScript) to remove 5' monophosphate RNA and convert 5'
triphosphate or capped RNAs to 5' monophosphate RNAs. Next, RNA prepared with
enzymatic treatments was ligated to the 5'-adapter (s1206-N15, 5'-
GUUCAGAGUUCUACAGUCCGACGAUCNNNNNNNNNNNNNNN-3') that contains
Illumina adapter sequence and a 15 nt randomized 3'-end to reduce ligation bias and
serve as a Unique Molecular Identifier (UMI). Next, cDNA was constructed by reverse
transcription using RT primer CKO2191-s128A (5'-
CCTTGGCACCCGAGAATTCCAAGTGAATAATTCTTCACCTTTA-3') followed by
emulsion PCR amplification for 20-22 cycles using Illumina PCR primers (RP1 and
RPI3-30). Final DNA was gel size selected for 180-250 bp lengths and sequenced by
Illumina NextSeq 500 (150 SE) or NovaSeq 6000 (200 SE) (**Table S2**) using custom
primer s1115 (5'-CTACACGTTCAGAGTTCTACAGTCCGACGATC-3') to avoid
potentially confounding effects of misannealing of the default pooled Illumina
sequencing primers to the two randomized sequence regions.

Primer extension assay

Primer extension assays were performed on the same batch of total RNA extracted for
TSS-seq as described in [94] with modifications described in [23]. For each reaction, 30 µg
total RNA was used. An RNA sample without library transformed was used as "no GFP"
control. A sample containing same amount of nuclease-free water was used as "no
RNA" control. A primer (CKO2191) complementary to the 6th to 27th bases of *GFP* ORF,
which is the same annealing region for reverse transcription of TSS-seq sample
preparation, was labeled with ^{32}P γ-ATP (PerkinElmer, BLU502Z250UC) and T4
polynucleotide kinase (Thermo Scientific, EK0031). M-MuLV Reverse Transcriptase
(NEB, M0253L), RNase inhibitor (NEB, M0307L), dNTPs (GE) and DTT were added to
mix of RNA and labelled primer for reverse transcription reaction. Before loading to
sequencing gel, RNase A (Thermo Scientific, EN0531) was added to remove RNA. The
products were analyzed by 8% acrylamide/bis-acrylamide (19:1 ratio, Bio-Rad,
1610145) gel containing 1x TBE and 7M Urea. Primer extension gels were visualized by
a Molecular Imager PharosFX™ Plus System (Bio-Rad) and quantified by Image Lab
(5.2).

Determination of NTP levels

For each genotype or treatment, six biological replicates were performed. Cells from
saturated overnight cultures were used to inoculate fresh SC medium containing 2%
raffinose at 0.25 x 10^7 cells/ml and grown to a density of 0.5-0.8 x 10^7 cells/ml, as
determined by cell counting. For each WT replicate, two 15 ml cultures were aliquoted
and treated with 20 ug/ml MPA or 100% ethanol for 20 min, respectively. Subsequently,
three-hour 4% galactose treatment was performed for all samples. About 1 x 10^7 cells
were harvested, and then cell pellets were snap frozen using liquid nitrogen and
immediately stored at -80 °C for further processing as described below.

Metabolic quenching and nucleotide phosphate metabolite pool extraction were
performed by adding 1 ml ice-cold 2:2:1 acetonitrile:methanol:water with 0.1% formic
acid with 1x MS-Stop phosphatase inhibitor (Sigma-Aldrich). Samples were spiked with
10 µl 100 mM mix of adenosine-^{15}N$_5$-5′-triphosphate, guanosine-^{15}N$_5$-5′-triphosphate,
thymidine-^{13}C$_{10}$,^{15}N$_2$-5′-triphosphate, cytosine-^{2}H$_{14}$-5′-triphosphate, and uridine-^{15}N$_2$-5′-
triphosphate (Sigma-Aldrich). Samples were homogenized via sonication at room
temperature, and the supernatant was then cleared of protein by centrifugation at
16,000 xg. The protein pellet was resuspended in RIPA buffer for protein normalization.
The cleared supernatant was dried down under nitrogen gas and resuspended in 100 µl
7.5 mM ammonium acetate / 0.05% ammonium hydroxide. 2µl of cleared supernatant
was subjected to online LC-MS analysis. Purified adenosine-5′-triphosphate, guanosine-
5′-triphosphate, thymidine-5′-triphosphate, cytosine-5′-triphosphate, and uridine-5′-
triphosphate (Sigma-Aldrich) were serially diluted from 50 pmol/µl to 0.39 pmol/µl to
generate calibration curves.

Analyses were performed by untargeted LC-HRMS. Briefly, samples were injected via a
Thermo Vanquish UHPLC and separated over a reversed phase Phenomenex Kinetix-
Polar C18 column (2.1 x 100 mm, 1.7 µm particle size) maintained at 55°C. For the
22.5-minute LC gradient, the mobile phase consisted of the following: solvent A
(water/7.5 mM ammonium acetate/0.05% ammonium hydroxide) and solvent B

(acetonitrile/0.05% ammonium hydroxide). The gradient was the following: 0-.1 min 2%
B, increase to 70% B over 12 minutes, increase to 98% B over 0.1 min, hold at 98% B
for 5 minutes, re-equilibrate at 2% B for five minutes. The Thermo IDX tribrid mass
spectrometer was operated in positive ion mode, scanning in ddMS^2 mode (2 μscans)
from 200 to 800 m/z at 120,000 resolution with an AGC target of 2e5 for full scan, 2e4
for ms^2 scans using HCD fragmentation at stepped 15,35,50 collision energies. Source
ionization setting was 3.0 kV spray voltage for positive mode. Source gas parameters
were 35 sheath gas, 12 auxiliary gas at 320°C, and 8 sweep gas. Calibration was
performed prior to analysis using the Pierce^{TM} FlexMix Ion Calibration Solutions
(Thermo Fisher Scientific). Integrated peak areas were then extracted manually using
Quan Browser (Thermo Fisher Xcalibur ver. 2.7). Calibration curves using purified
standards were then used to convert peak area ratios to concentration.

Computational analyses

Data and statistical analyses were performed in Python (3.8.5) and R (4.0.0)
environments. Additional packages usages are reported throughout the methods
description. Source code is provided at https://github.com/Kaplan-Lab-
Pitt/PolII_MASTER-TSS_sequence. Raw sequencing data have been deposited on the
NCBI SRA (Sequence Read Archive) under the BioProject accession number
PRJNA766624. Processed data have been deposited on the GEO (Gene Expression
Omnibus) under the accession number GSE185290. Visualizations were compiled in
Adobe Illustrator 2021.

DNA-seq analysis. High-throughput sequencing of template DNA amplicon was used
to assign each 9 nt randomized TSS sequence to a corresponding 24 nt barcode. First,
paired-end reads were merged using PEAR (0.9.11) [95]. Next, we considered only those
reads that contained a perfect match to three sequence regions common to all variants:
27 nt sequence upstream of the TSS region, 24 nt sequence between TSS region and
barcode, and 27 nt sequence downstream of barcode (5'-
TTCAAATTTTTCTTTTGATTTTTTTTCNNNNNNNNNNACATTTTCAAAAGGCTAACATC
AGNNNNANNNNCNNNNTNNNNGNNNNATGTCTAAAGGTGAAGAATTATTCACT-3',
randomized TSS and barcode regions are underlined). From these reads, 9 nt TSS
region and 24 nt barcode were extracted, followed by individually error correction using
UMI-tools (1.0.0) [96]. Next, for barcodes linking to multiple TSS variants, only barcodes
for which >= 90% of the sequencing reads containing a specified barcode also
contained a shared, exact 9 nt TSS region were kept. To generate a master pool of
TSS-barcode linkages for all TSS-seq samples, for each library ("AYR", "BYR", "ARY"),
TSS-barcode linkages that existed in at least two out of four samples (one *E.coli* sample
plus three WT yeast replicates) and in which >= 5 reads existed were kept and pooled.
Two types of processed data are available in GEO database, with accession numbers
listed in (**Table S2**): tables containing TSS-barcode linkages and corresponding DNA-
seq read counts for each sample, tables of the master pool containing kept TSS-
barcode linkages and corresponding DNA-seq read count in all related samples.

TSS-seq analysis for libraries. High-throughput sequencing of RNA samples was
used to link RNA products to barcodes, therefore assign TSS usage to corresponding

DNA templates. For TSS identification and subsequent preference analysis, we
considered only those reads that contained a perfect match to a 27 nt sequence region
downstream of the barcode, as well as expected length of 5'-end: 5'-[15 nt 5'-UMI]-[>1
nt upstream of barcode region, designated as "RNA 5'-end"]-[24 nt barcode]-[the first 27
nt of GFP ORF, ATGTCTAAAGGTGAAGAATTATTCACT]-3'. Those reads carrying one
or more mismatches from expected sequences were used for analysis of relative
expression of promoters. Next, 15 nt 5'-UMIs, "RNA 5'-end" with varying length, and 24
nt barcode were extracted and individually corrected by UMI-tools (1.0.0). Deduplication
was performed based on 5'-UMIs, meaning reads contained a shared UMI-"RNA 5'-
end"-barcode linkage were counted as one deduplicated read for further analysis. Next,
the identity of the 24 nt barcode was used to determine the template sequences of
randomized TSS region. Then, reads with "RNA 5'-end" sequence perfectly matched to
corresponding template sequence were kept and used for downstream analysis. A TSS-
seq count table containing TSS usage distribution of each TSS promoter variant was
generated. In the count table, each row represents one TSS promoter variant, and each
column represents one position between positions -68 to +25 relative to "designed" +1
TSS. The number in each cell represents TSS-seq reads generated from a particular
position, with perfectly match to DNA template. After investigating reproducibility
(**Figure S3B-C, S4A-C**), count tables generated from three biological replicates were
merged into one by aggregating read counts at each position. Promoter variants with >=
5 TSS-seq reads in each replicate and whose Coefficient of Variation (CV) of TSS-seq
reads is less than 0.5 were kept. Two types of processed data are available in GEO
database, with accession numbers listed in **Table S2**: tables containing "RNA 5'-end"-
Barcode linkages and corresponding deduplicated TSS-seq read counts for individual
sample, TSS-seq count tables for individual samples and after aggregating replicates.
As an example, a TSS-seq count table including positions -10 to +25 relative to
"designed" +1 TSS after aggregating replicates of WT libraries is shown as **Table S4**.

TSS efficiency calculation. TSS efficiency for each position was calculated by dividing
reads count at a particular position by the reads at or downstream of this TSS. TSS
positions with >= 20% efficiency but with <= 5 reads left for this and following positions
were filtered out, as well as their downstream positions.

Relative expression analysis for libraries. The relative expression from each library
promoter variant was defined as the ratio of normalized TSS-seq reads generated from
a particular promoter variant to normalized DNA-seq reads containing the variant.

Sequence preference analysis. For sequence preference at each position, all TSS
variants were subgrouped based on the bases at a particular position. TSS efficiencies
of TSS variants were visualized as scatter plots using GraphPad Prism 9. Kruskal-
Wallis with Dunn's test was performed to test sequence preference in GraphPad Prism
9. Next, TSS efficiency medians of each subgroup were calculated and centered to
calculate "relative efficiency" at each position. The relative efficiencies were visualized
as sequence logos using Logomaker (0.8) [97]. In motif enrichment analysis, surrounding
sequences relative to examined TSSs were extracted and visualized as sequence logos
using WebLogo 3 [98]. Heatmaps, scatter plots and density plots for comparing Pol II WT

and mutants were generated by Morpheus
(https://software.broadinstitute.org/morpheus) or ggplot2 (3.3.3) R package.

Interaction analysis. The interaction between positions is defined as different bases
existing at one position resulting in different sequence preferences at another position.
For any two positions, all TSS variants were subgrouped based on bases at both
positions. Median values of TSS efficiency distribution of each subgroup were
calculated and centered twice to calculate "centered relative efficiency". The centered
relative efficiencies were visualized as heatmaps using Seaborn (0.11.0). Interactions
related to positions between -11 to -9 were calculated using datasets of designed +4
TSS deriving from "AYR", "BYR" and "ARY" libraries. Other interactions were calculated
using datasets of designed +1 TSS deriving from "AYR" and "BYR" libraries.

TSS-seq analysis for genomic TSSs. Genomic TSS-seq datasets from our lab's
previous study were used for comparison of model to in vivo TSS usage [27]. Quality
control, read trimming and mapping were performed as described in [21,27] to generate a
TSS count table that contains TSS-seq reads at each individual position within known
promoter windows ("median" TSS, 250 nt upstream and 150 nt downstream from
median TSS position). TSS efficiency calculation and subsequent sequence preference
analyses were performed as that for Pol II MASTER libraries. EtOH and MPA treatment
for genomic TSSs are described in Results and these libraries were analyzed as TSS-
seq was above. TSSs within a promoter were analyzed by their positions within the
cumulative distribution of usage from upstream to downstream, with 50% of the
distribution representing the median TSS position. To limit potential effects of marginal
TSSs, analyses of MPA treatment were limited to TSSs showing ≥2% efficiency in the
EtOH condition within the 25%-75% of the TSS distribution.

Prediction of TSS efficiency. To prepare datasets for modeling, positions of designed
-8 to +2 and +4 TSSs of each promoter variant that have valid TSS efficiency were
compiled as sequence variants. For each TSS variant, sequences at -11 to +9 positions
relative to TSS, together with corresponding TSS efficiency, were extracted. 80% of
dataset were randomly partitioned as training set and the rest 20% as testing set. To
select robust features, a forward stepwise strategy with a 5-fold Cross-Validation (CV)
was employed in two major stages, for additive terms and for interactions. Starting with
no variable in the model, logistic regression models with one additional variable (the
sequence at a particular position) were trained to predict TSS efficiency on training set
by train() of caret (6.0.86) R package [99], with a 5-fold CV. The R^2, representing the
proportion of variance explained, was calculated to indicate the performance of each
model. The variable that provides the highest increased R^2 for model was added into
the model for next round of variable selection. This process was repeated until the
increased R^2 is less than 0.01. After identifying the most influential additive variables,
same process was repeated for investigating robust interactions between selected
additive variables. Next, a final model with selected robust features. including additive
variables and interactions, was constructed on entire training set using glm() and
investigated on testing set. Comparison between predicted and measured efficiencies
was visualized as scatter plots using LSD (4.1.0) R package. Model parameters were
extracted and used to further calculation. Visualizations were done in Logomaker (0.8)

and Seaborn (0.11.0) in Python. Principal Component Analysis (PCA) was performed
using prcomp() in R.

Supplementary Tables

Supplementary Table 1. Oligonucleotides, yeast strains, and plasmids.

Supplementary Table 2. Statistics and data accession information for DNA-seq and TSS-seq.

Supplementary Table 3. Summary of libraries.

Supplementary Table 4. TSS-seq count table of WT libraries.

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Authors' contributions

Y.Z. designed the project, performed experiments, analyzed data, made figures, drafted and revised the manuscript. I.V. generated libraries for TSS-seq. S-H.Z. analyzed data and discussed analysis. B.E.N. provided funding and methodology of TSS-seq, and revised the manuscript. C.D.K. conceived and designed the project, guided analyses and interpretation of data, provided funding, revised the manuscript.

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Data availability

Raw sequencing data generated in this study are available in the NCBI BioProject, under the accession number PRJNA766624. Processed data are available in GEO, under the accession number GSE185290.

Code availability

Code for analyses in this study is provided at https://github.com/Kaplan-Lab-Pitt/PolII_MASTER-TSS_sequence.

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1183 **Figures and Figure Legends**

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Figure 1. A high-throughput system for studying transcription TSS selection. (A and B) Pol II initiation in yeast proceeds by a promoter scanning mechanism. Yeast Pol II initiation usually occurs at multiple TSSs ranging from 40 to 120 bp downstream of a core promoter region comprising the PIC assembly position (the TATA-box for TATA-containing promoters). After the PIC assembles upstream (core promoter) scanning will proceed over a scanning region towards positions where TSS selection occurs (TSS region). In addition to sequence control of initiation through suitable TSS sequences, the probability of initiation across promoter positions is also controlled by multiple architectural features shown in **(A)**. These include the inhibition of initiation near a core promoter that diminishes as scanning proceeds ("distance restriction"), biochemical restrictions on how far scanning can proceed that increase as scanning proceeds ("scanning processivity"), and "Pol II flux", which represents the decrease in amount of scanning Pol II as scanning proceeds due to conversion of scanning Pol II to transcribing Pol II upon initiation. **(C)** Construction of promoter libraries that control TSS sequence context. Top panel shows schematic of the *SNR37* promoter and its TSS usage distribution based on TSS-seq [21]. *SNR37* has a focused and highly efficient TSS region. Bottom panel shows schematic of the Pol II MASTER libraries used in this study. A duplication of the *SNR37* TSS region was inserted before native TSS region, and the -8 to +1 positions relative to native *SNR37* +1 TSS (black box) were replaced by a 9 nt randomized region to provide almost all possible sequences. The second *SNR37* TSS region functions as a "Flux Detector" (FD) to capture Pol II flux that fails to initiate within the randomized region and allow determination of initiation efficiency within the randomized region. A barcode region (purple box) allows RNA products to be assigned to respective promoter variants. The *GAL1* UAS allows control of library expression, and the *GFP* ORF and *CYC1* terminator support termination and stabilize RNA products. **(D)** TSS usage distributions at TSS and FD regions for different promoter variant "AYR", "BYR", and "ARY" libraries are shown on the left. TSS usages from designed +1 TSS and positions upstream are in red and grey, respectively. TSS usage from the FD region is in brown. The definition of "TSS efficiency" and overall TSS efficiency calculations for the aggregate +1 TSS of all in different libraries are shown in the middle. Example correlation plots of TSS efficiency calculations for +1 TSSs from individual promoter variants in Pol II MASTER libraries between representative biological replicates are show on right. Pearson r and number of compared variants are shown.

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Figure S1. High level of reproducibility and coverage depth of library variants. (A)
Schematic representation of experimental approach. Promoter libraries with almost all
possible sequences within a 9 nt randomized region in a promoter context designed for
specific functionalities were constructed on plasmids. Libraries were of three types,
designated "AYR", "BYR", and "ARY" based on compositions of their randomized
regions. Plasmids were amplified in *E. coli* and transformed into yeast strains with wild
type or mutated Pol II. DNA and RNA were extracted from yeast pellets and prepared
for DNA-seq and TSS-seq. **(B)** Base frequencies at positions within the randomized
region of promoter variants demonstrate unbiased synthesis of randomized regions.
Bars are mean +/- standard deviation of the mean for promoter variants in WT and four

Pol II mutants. **(C)** Heatmap illustrating hierarchical clustering of Pearson correlation coefficients of reads per promoter variant for libraries amplified in *E. coli* and three biological replicates of libraries transformed into yeast. **(D)** Example correlation plots of DNA reads count of promoter variants for *E. coli* and yeast WT biological replicates. Pearson r and number of compared variants are shown. **(E)** Bulk primer extension for RNA produced from promoter variant libraries and quantification for biological replicates transformed into WT yeast. "No GFP" control used an RNA sample without a library transformed. "No RNA" control used a sample of nuclease-free water. Dots represent three biological replicates. Bars are mean +/- standard deviation of the mean. **(F)** TSS usage distribution based on insert length of TSS-seq reads generated from transformed libraries. Dots represent three biological replicates. Bars are mean +/- standard deviation of mean. Distributions are similar to the distributions in **(E)**. Note that primer extension will blur usage into adjacent upstream position due to some level of non-templated addition of C to RNA 5' ends. **(G)** Heat scatter plots of Coefficient of Variation (CV, y axis) versus total RNA sequencing reads per promoter variant in each of three Pol II MASTER libraries. A cutoff of CV=0.5 was used to filter out variants with higher variance.

Figure 2. TSS and surrounding sequences modulate initiation efficiency over a wide range. (A) Comparison of single base mutation effects on TSS efficiency measured by Pol II MASTER and primer extension. (Left) Sequences of *SNR14* and

SNR37 TSS regions (in black boxes, including positions between -8 to +1 relative to
TSS) and all possible single substitutions of *SNR14* TSS region. Single substitutions
included in both a prior *SNR14* mutagenesis study [22] and Pol II MASTER libraries are in
blue while those lacking in our study are in brown. Additional substitutions analyzed
here are in gray. (Middle) High correlation of TSS efficiency measured by Pol II
MASTER and primer extension. Mutation classes color coded as on left. (Right) Range
of effects of single base substitutions on TSS efficiency for *SNR14*- and *SNR37*-related
sequences. Mutation classes color coded as on left. Single substitutions absent from
Pol II MASTER because of library design (R_{-1} and Y_{+1}) were expected to have super low
TSS efficiencies. Double substitutions of *SNR14* and *SNR37* TSS region included in Pol
II MASTER "ARY" library are shown as orange inverted triangles and show almost no
efficiency. **(B)** Pol II initiation shows a strong preference for A_{-8} and $C_{-1}A_{+1}$ containing
variants. All promoter variants were divided into 20 groups defined by bases at positions
-8, -1 and +1 relative to the designed +1 TSS, and their +1 TSS efficiencies were plotted
as spots. Lines represent median efficiencies of each group. **(C)** Definition of "designed"
+1 TSS (designated as +1) and positions relative to this TSS (blue TSS arrow and
sequence). TSS usage generated from upstream or downstream of "designed" +1 TSS
(green and brown TSS arrows and sequences, respectively) allows us to study
sequence preferences at positions -11 to -9 and +2 to +9 relative to site of initiation. **(D)**
Schematic illustrating how "relative efficiency" is calculated and visualized as a
sequence logo in **E**. At a particular position relative to a particular TSS, first, all variants
were divided into the four base subgroups defined by base at this position. Next,
median values of each base group were extracted and centered based on the mean of
all median values. The centered median values were defined as "relative efficiencies",
representing preferences for bases at this particular position. Finally, relative
efficiencies of bases were visualized as sequence logos. Positive and negative values
indicate relatively preferred or less preferred bases, respectively. **(E)** Pol II initiation
shows distinct sequence preference at positions around the TSS. Top panel was
generated using datasets of designed +4 TSS deriving from "AYR", "BYR" and "ARY"
libraries. Middle panel was generated using datasets of designed +1 TSS deriving from
"AYR" and "BYR" libraries. Bottom panel was generated using datasets of designed +1
TSS deriving from "AYR" and "ARY" libraries. Positions that contain fixed or not
completely randomized bases are shown in grey. **(F)** Schematic illustrating how
sequence interaction between positions is calculated and visualized as a heat map in **G**.
Using -9/-8 positions as an example, relative efficiencies at position -8 were calculated
when different bases were present at position -9. Next, relative efficiencies of each base
were centered based on the mean of all relative efficiencies of a particular base. After
centralization, negative and positive values indicate negative and positive interactions.
Interaction scores for two sequence positions are read at the intersection of the *x* and *y*
axes labeled by base and position. That means, for example, -9.3 at -9A/-8A and 5.9 at
-9A/-8G intersections indicate having a -9A decreases A preference and increases G
preference at position -8, respectively. Finally, the centered relative efficiencies matrix
was visualized as a heat map to represent the interaction between examined positions.
(G) Sequence interactions are mainly observed at neighboring positions. Red and blue
indicate positive and negative interactions, respectively. Missing values are shown in
grey. Interactions related to positions -11 to -9 were calculated using datasets of

designed +4 TSS deriving from "AYR", "BYR" and "ARY" libraries. Other interactions

were calculated using datasets of designed +1 TSS deriving from "AYR" and "BYR"

libraries.

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Figure S2. Surrounding sequence of TSSs modulates initiation efficiency. (A) +1
TSS efficiency of all -7 to -2 sequences within each $N_{-8}N_{-1}N_{+1}$ motif in WT, rank ordered
by efficiency of their $A_{-8}C_{-1}A_{+1}$ version, is shown as a heat map. The x-axis is ordered
based on median efficiency values for each $N_{-8}N_{-1}N_{+1}$ motif group, as shown in **B**.
Statistical analyses by Spearman's rank correlation test between $A_{-8}C_{-1}A_{+1}$ group and all
groups are shown beneath the heat map. **(B)** Efficiency distributions of designed +1
TSSs grouped by base identities between -8 and +1 positions. Statistical analyses by
Kruskal-Wallis with Dunn's multiple comparisons test for base preference at individual
positions relative to +1 TSS are shown beneath data plots. Lines represent median
values of subgroups. ****, P ≤ 0.0001; ***, P ≤ 0.001; **, P ≤ 0.01; *, P ≤ 0.05. **(C)**
Histogram showing the distribution of measured efficiencies for all designed -8 to +4
TSSs of all promoter variants from "AYR", "BYR" and "ARY" libraries in WT. Dashed
lines mark the 5% efficiency cutoff used to designate a TSS as active. Total number of
TSS sequences is shown. **(D)** $A_{+2}G_{+3}G_{+4}$ motif enrichment is apparent for the top 10%
most efficient designed -8 TSS. $A(/G)_{+2}G(/C)_{+3}G(/C)_{+4}$ motif enrichment was observed
for the top 10% most efficient -8 TSSs but not for the next 10% most efficient TSSs.
$A(/G)_{+1}$ enrichment observed for top 20% most efficient TSSs is consistent with the +1R
preference of TSS. Numbers (N) of variants assessed are shown. Sequence logos were

generated using WebLogo 3. Bars represent an approximate Bayesian 95% confidence

interval. **(E)** An A at position -9 results in different sequence preferences at position -8.

The dataset of designed +4 TSSs deriving from "AYR", "BYR" and "ARY" libraries was

used to detect the -9/-8 interaction. All variants were divided into 16 subgroups defined

by bases at positions -9 and -8 relative to designed +4 TSS, and then their TSS

efficiencies were plotted. Lines represent median values of subgroups. **(F)** An A at

position -8 results in different sequence preferences at position -7. The dataset of

designed +1 TSSs deriving from "AYR" and "BYR" libraries was used to detect -8/-7

interaction. Calculations same as -9/-8 interaction described in **E**.

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Figure 3. Pol II mutants alter TSS efficiency for all possible TSS motifs while showing selective effects for base at +1. (A) Pol II mutants alter TSS efficiencies across all motifs, corresponding to their direction of change to Pol II catalytic activity *in vitro*. TSS efficiency changes for each TSS variant were first determined by subtracting WT efficiency from Pol II mutant efficiency. The medians of efficiency changes for variant groups with indicated bases at each position relative to TSS were then calculated and illustrated in a heat map. Positive (red) values indicate Pol II mutants increased overall efficiency while negative (blue) values indicate decreased overall efficiency. **(B)** WT TSS efficiency for TSS variants divided into motif groups are plotted for mutant TSS efficiencies for the same TSS groups. The eight possible groups of TSSs for $A/B_{-8}C/T_{-1}A/G_{+1}$ motifs were plotted and curve fit. Histograms show density of variants within each -8/-1/+1 subgroups. As to position -8, A_{-8} containing motifs show higher efficiency than B_{-8} containing motifs in both Pol II GOF (G1097D and E1103G) and LOF (F1086S and H1085Q) mutants (A_{-8} motifs: maroon and blue vs B_{-8} motifs: light coral and light blue). This is consistent with the proposed function of -8A to retain TSSs longer in the Pol II active site during scanning. This means that -8A may boost the positive effects of GOF mutants, therefore Pol II GOF mutants showed greater effects on A_{-8} motifs compared to B_{-8} motifs. In contrast, -8A compensates for active site defects of LOF mutants, therefore Pol II LOF mutants showed reduced effects on A_{-8} motifs compared to B_{-8} motifs. Both GOF and LOF mutants show reduced effects on G_{+1} motifs relative to A_{+1} motifs (G_{+1} motifs: light coral and maroon vs A_{+1} motifs: light blue and blue).

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Figure S3. (A) Histograms showing the distribution of measured efficiencies for all designed -8 to +4 TSSs of all promoter variants deriving from "AYR", "BYR" and "ARY" libraries in Pol II mutants. Dashed lines mark the 5% efficiency cutoff used to designate a TSS as active. Total numbers of TSS sequences are shown. **(B)** TSS usage distributions at designed -10 to +25 TSSs in different promoter variant "AYR", "BYR", and "ARY" libraries. Pol II GOF and LOF mutants shifted TSS usage upstream or downstream relative to WT, respectively. Dots represent three biological replicates. Bars are mean +/- standard deviation of the mean. **(C)** Hierarchical clustering of Pearson correlation coefficients of TSS efficiencies for the major TSSs (designed +1 TSS for "AYR" and "BYR" libraries, +2 TSS for "ARY" library) for three biological replicates for WT or mutant Pol II illustrated as a heat map. **(D)** Example correlation plots of TSS efficiency of major TSSs of promoter variants between representative biological replicates. Pearson r and number of compared variants are shown. **(E)** Plots

of the coefficient of variation (CV) versus the total RNA reads for three yeast replicates
in Pol II mutants. The red dashed lines mark the CV=0.5 cutoff, which was chosen as an
arbitrary cutoff for variants showing reasonable reproducibility across three biological
replicates. G1097D replicates contain outliers because these slow growing strains are
susceptible to genetic suppressors. These outliers are filtered out using our CV cutoff
because of high CVs caused by suppressor only existing in one of three biological
replicates. **(F)** TSS efficiency distributions of designed +1 TSSs of Pol II mutants for
base subgroups at individual positions relative to +1. Identical analysis as in **Figure**
S2B for WT was performed for Pol II mutant libraries. **(G)** Pol II GOF G1097D showed
greater increase in efficiency than GOF allele E1103G at upstream TSSs (designed -32
and -8 TSSs), while E1103G showed stronger effects at designed +1 TSS than
G1097D. This may indicate that where Pol II is in the scanning process may affect
efficiency. **(H)** Pol II initiation sequence preference in Pol II mutants. Identical analysis
as in **Figure 2E** for WT was performed for Pol II mutant libraries. Sequence logos were
generated using multiple datasets and the dashed lines indicate divisions in datasets
used to generate them. Specifically, preferences at positions -11 to -9 were generated
using datasets of designed +4 TSS deriving from "AYR", "BYR" and "ARY" libraries.
Preferences at positions -8 to -2 were generated using datasets of designed +1 TSS
deriving from "AYR" and "BYR" libraries. Preferences at positions -1 and +1 were
generated using datasets of designed +1 TSS deriving from "AYR" and "ARY" libraries.
(I) Motif enrichment for top the 10% most efficient -8 TSSs for Pol II mutants. Identical
motif enrichment analysis as in **Figure S2D** top panel for WT was performed to Pol II
mutant libraries. Numbers (N) of variants assessed are indicated. Bars represent an
approximate Bayesian 95% confidence interval.
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Figure 4. Pol II initiation is sensitive to NTP pools. (A) NTP levels measured in WT, Pol II E1103G (GOF) and Pol II F1086S (LOF) mutants. Statistical analyses by Ordinary one-way ANOVA with Dunnett's multiple comparisons test for each NTP leave are shown. ****, P ≤ 0.0001; ***, P ≤ 0.001; **, P ≤ 0.01; *, P ≤ 0.05. **(B)** NTP levels measured in WT, WT treated with 100% ethanol (the solvent for MPA) and WT treated with 20 ug/ml MPA. Statistical analyses by RM one-way ANOVA with Dunnett's multiple comparisons test for each NTP leave are shown. ****, P ≤ 0.0001; ***, P ≤ 0.001; **, P ≤ 0.01; *, P ≤ 0.05. **(C)** MPA selectively decreases TSS efficiencies of +1G TSSs (right) but shows no effects on +1A TSSs (left) in Pol II MASTER libraries. **(D)** MPA alters TSS efficiencies at genomic TSS positions. TSS efficiency change shown as difference between efficiency in EtOH and MPA treatment for all TSSs ≥ 2% efficiency in EtOH conditions within the 25%-75% of TSS distributions (see Methods). Bars on violin plots indicate medians computed based on density estimates. **(E)** Comparison of TSSs of any composition for first six nucleotides (NNNNNN) with those that lack G (HHHHHH, H=not G) or subsets of TSSs with A,C,G,T at positions +1 to +6. Bars on violin plots are as in **(D)**. All C,G,T datasets for all positions 1-5 are distinct from A as a baseline comparison (Wilcoxon rank sum test with continuity correction). Position 6 each base is not significantly different from A.

Figure S4. High of reproducibility of TSS usage and efficiency upon MPA treatment. (A) TSS usage distributions at designed -10 to +25 TSSs in WT "NYR" library (mixed AYR and BYR libraries) treated with 100% ethanol or with 20 ug/ml MPA.

MPA treatment shifted TSS usage downstream relative to EtOH treatment. Dots
represent three biological replicates. Bars are mean +/- standard deviation of the mean.
(B) Hierarchical clustering of Pearson correlation coefficients of TSS efficiencies for
designed +1 TSS for three biological replicates for MPA or EtOH treatment, illustrated
as a heat map. **(C)** Hierarchical clustering of Pearson correlation coefficients of TSS
efficiencies for all genome positions within defined promoter windows with >=3 reads in
each replicate, illustrated as a heat map. **(D)** Correlation plots for combined biological
replicates for TSS efficiency upon MPA treatment (*y* axes) versus EtOH treatment (*x*
axes) for all TSSs ≥2% efficiency in the 25%-75% of the distribution for a curated set of
5979 yeast promoters (see Methods). TSSs are separated into groups depending on
base identity at positions -3 (control) or positions +1 to +6.

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Figure 5. Learned initiation preferences are predictive of TSS efficiencies at
genomic promoters. (A) Sequence preferences determined from TSSs representing
the median of the distribution of usage from a set of 5979 yeast promoters ("median
TSSs") are congruent with library determined TSS preferences, except there is
preference for A at positions -7 to -5 learned from genomic data. Sequence context and
TSS efficiency of median TSSs were extracted from genomic TSS-seq data
(GSE182792) [27]. Calculation and visualization were as performed for promoter variant
libraries. The number (N) of genomic median TSSs examined is shown. Statistical
analyses by Spearman's rank correlation test between relative efficiency at individual
positions learned from promoter variant libraries and genomic median TSSs are shown
beneath the sequence logo. **(B)** A-richness upstream and downstream of TSS is
observed for highly expressed median TSSs. The number (N) of analyzed median TSSs
is shown. Bars represent an approximate Bayesian 95% confidence interval. **(C)** Having
an A at either position -9 or -8 reduces enrichment of A at the other position. Top 20%
expressed median TSSs were divided into subgroups based on bases at position -9 or -
8. Motif enrichment analysis was individually performed to subgroups. Numbers (N) of
median TSSs within each subgroup are shown. Bars represent an approximate
Bayesian 95% confidence interval.

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Figure 6. Logistic regression model of DNA sequence contribution to TSS efficiency. (A) Regression modeling identifies key DNA sequences and interactions contributing to TSS efficiency. Nine additive parameters plus one interaction were selected for the final model. Dots represents average R^2 obtained in a 5-fold Cross-Validation (CV) strategy for logistic regression models using different numbers of features. The black line with SD error bars represents models with the best performance under a certain number of predictors. Note that model with as few as three additive parameters could explain 74.10% of TSS efficiency variation in our WT libraries. **(B** to **E)** Good performance of model including sequences at nine positions and the -9/-8 interaction indicates that TSS efficiency in our libraries is mainly regulated by the included features. (**B**) A scatterplot of comparison of measured and predicted efficiency of test sets, with a 5% efficiency cut-off. Model performance R^2 on entire test and number (N) of data points shown in plot are shown. (**C**) A sequence logo of centered additive parameters. The coefficients for bases at a particular position was centered and visualized as a sequence logo. (**D**) A sequence logo showing learned preference at position -8 when different bases existing at position -9, with -9/-8 interaction included. The -9/-8 interaction parameters were added to corresponding additive coefficients for bases at position -8. The additive plus interaction parameters were then centered and visualized as sequence logos. (**E**) A heat map of centered parameters for -9/-8 interaction illustrating how bases at one position affect preference at another position. **(F** to **G)** Efficiency prediction for positions within known promoter windows in WT shows overall over-prediction. Scatterplots of comparison of measured and predicted TSS

efficiencies of all positions (with a 5% efficiency cut-off) (**F**) or median TSSs (**G**) within

5979 known genomic promoter windows [21] with available measured efficiency. **(H)**

Model shows better performance on Taf1-depleted promoters and promoters with

medium to high expression. Scatterplots of comparison of measured and predicted TSS

efficiency of median TSSs subgrouped by promoter classes and expression levels.

Expression levels of genomic promoters are defined based on their total TSS-seq reads

in the promoter window in the examined datasets: low, [0, 200); medium, [200, 1000);

high, [1000, max). Pearson r and number (N) of compared variants are shown.

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Figure S5. (A) Overview of the TSS efficiency modeling process. (1) Variants including
designed -8 to +2 and +4 TSSs deriving from "AYR", "BYR" and "ARY" libraries with
available TSS efficiency were pooled for modeling. (2) Sequences at positions -11 to +9
relative to TSS of each variant were extracted. (3) To identify robust features, a forward
stepwise selection strategy coupled with a 5-fold cross-validation for logistic regression
was used. Data were randomly split into training (80%) and test (20%) sets. The training
set was used for a stepwise regression approach that starts from a model with a
constant term only and adds variables that improve the model the most one at a time,
until a stopping criterion is met. In stage I, additive terms (sequences at positions -11 to
+9) were tested. In stage II, interactions between positions selected in stage I were
tested. Model performance was evaluated with R^2. The stopping criterion for adding
additional variables was an increase $R^2 < 0.01$. (4) A logistic regression model
containing selected robust features was trained with training set and then evaluated with
the test set. **(B)** A scatterplot of comparison of measured efficiencies and predicted
efficiencies within the test sets. Model performance R^2 on entire test set and number of
data points shown in plot are shown. **(C)** PCA analysis for parameters of models trained
by the individual replicates of WT and Pol II mutant. Parameters for replicates cluster
more closely than parameters for different Pol II mutants, indicating that modeling
captured features of different Pol II groups and models are not overfit. The top 15
contributing variables are shown. GOF and LOF mutants were separated from WT by
the 1st principal component. GOF G1097D and E1103G were further distinguished by
2nd principal component by additional position +2 information, which is consistent with
results in **Figure S3I**, where G1097D and E1103G differentially altered +2 sequence
enrichment. **(D)** A scatterplot of comparison of measured and predicted TSS efficiencies
of all positions within 5979 known genomic promoter windows [21] with available
measured efficiency. Pearson r and number (N) of compared variants are shown. Most
promoter positions (82%, 1,678,406 out of 2,047,205) showed no observed efficiency,

which is expected because TSSs need to be specified by a core promoter and scanning
occurs over some distance downstream.

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Figure 7. Model for TSS sequence preference regulated by multiple mechanisms.
Top panels show determined contribution of sequence at positions around TSS and
proposed mechanisms. Two major groups of positions around TSS contribute to TSS
selection: bases around TSS (actual initiating site) and bases around position -8. The
TSS and adjacent bases interact with Pol II active site, the 1st NTP or each other to
facilitate stable binding of 1st NTP thus stimulate RNA synthesis. -8 and -9 Ts on the
template strand with an additional interaction between -8 and -7 template strand
positions may serve as an anchor point interacting with TFIIB B-reader domain allowing
pausing of scanning and promotion of Pol II initiation at TSSs a fixed distance
downstream, as proposed [47]. These preferences are reflected as As if the analysis is on
the coding strand. Positions and interactions that were identified by regression modeling
as robust features are labelled in bold. Bottom panel shows other architectural features
involved in Pol II transcription initiation likely additionally contributing to TSS selection
and initiation efficiency that will be accessible to Pol II MASTER analysis.