

Supplementary Information

Prediction of clustered RNA-binding protein motif sites in the mammalian genome

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Supplementary Notes: Specification of the HMM

Emission probability

The HMM underlying mCards consists of six states, among which three states represent RBP-bound motif sites at different positions of a cluster (S+, I+, +, shown in blue in Fig. 1C) and three other states represent background motif sites (S-, I-, -, shown in gray, Fig. 1C). Each motif site (e.g., YCA Y element) is an observation of one of the six states characterized by three types of features in the emission probability: clustering of neighboring sites (d), their accessibility (a) and conservation (c), as summarized in Table 1.

Clustering of motif sites is explicitly modeled by the distance of each site to the preceding site (denoted as d), except the very first site in an input sequence (states S+ and S-, Fig. 1C and Table 1), for which a dummy distribution was used (Table 1). The distance of the second and additional succeeding sites in a cluster (state +) followed the same distribution estimated from motif sites in CLIP tag clusters ($P(d|+)$, e.g., blue curve in the left panel, Fig. 1B), while the distance of any other states followed another distribution estimated from YCA Ys in background sequences ($P(d|s)$, $s=\{I+, I-, -\}$, gray curve in the left panel, e.g., Fig. 1B). We also censored the distribution for motif sites belonging to a RBP-bound cluster [$P(d|+)$, e.g., blue curve in the left panel, Fig. 1B], which imposed an implicit limit on the maximum spacing allowed for individual motif sites in an RBP-bound cluster. For this study, we required $d \leq 30$, a threshold determined empirically by examining validated Nova and Mbnl target exons (see below), which is much less restrictive than several previous studies (1,2). However, we also tried other thresholds (e.g., 20 and 50) and obtained qualitatively similar results.

Motif conservation (denoted as c) was quantified using multiple alignments of 20 mammalian species (3) by branch length scores (BLS) (4), which were previously demonstrated to be effective in predicting binding sites of brain- and muscle-specific splicing factors of the RBFOX family (5). Because the basal conservation level of sequences in different genomic regions varies dramatically, we modeled motif site conservation separately for sites in 5' untranslated regions (UTRs), coding sequences (CDS), introns and 3' UTRs.

In vitro selection (6,7), X-ray crystallographic data (8), CLIP data (9) and computational analysis (10,11) consistently suggested that RNA secondary structures can modulate the accessibility and function of RBP motif sites. The accessibility of a motif site (denoted as a) in local RNA secondary structures was measured by its probability of being located in single-stranded region, as predicted by the RNAplfold program (parameters: -u 4 for tetramer motif, and a default window size of 70 nt) in the ViennaRNA package (12). The probabilistic distributions of these features were estimated nonparametrically from motif sites in training CLIP tag clusters and background sequences, respectively, and represented by histograms (e.g., Fig. 1A and B). Since conservation and accessibility scores are continuous, these variables were discretized into a specified number of bins (e.g., 20 bins in this study).

Together, each motif site can be denoted by $\mathbf{x}_{i,l_i} = (d_i, a_i, c_i)$ ($i=1,2,\dots, N$), in which i is the index of the motif site, and l_i is an indicator that records the genomic location of the site (5' or 3' UTRs, CDS, or introns). For simplicity, these features are assumed to be independent with each other to estimate the overall emission probability:

$$e(s_i, \mathbf{x}_{i,l_i}) = P(\mathbf{x}_{i,l_i} | s_i) = P(d_i | s_i)P(c_i | s_i, l_i)P(a_i | s_i), \quad (1)$$

where s_i represents the state of the site.

Transition between states

Legitimate transitions between states $t(s_i, s_{i+1})$ are naturally determined by the definition of a motif cluster while the other transitions are not allowed to occur (Fig. 1 C). We denote the probability of staying at the “+” state and the “-” as $a[+]$ and $a[-]$, respectively, and they can be estimated from the average duration (number) of sites in each CLIP tag cluster ($\mu[+]$) and background sequences ($\mu[-]$), respectively, $a[+]=1-1/\mu[+]$, and $a[-]=1/\mu[-]$. Since the size of each background sequence was chosen arbitrarily, for this study we estimated $\mu[-]$ based on the relative frequency of YCA Ys in positive versus negative training set r , $\mu[-] = \mu[+]/r$. Transitions between different states can be calculated accordingly, as summarized in Fig. 1C.

Model training and prediction

The parameters of the model $\lambda = (t, e)$ (transition probabilities and emission probabilistic distributions) were estimated from the training data. During prediction, each input sequence (i.e., whole transcripts including introns with 10 kb extension on both sides for this study), which usually contains a sufficiently large number of motif sites, is processed independently. The model takes the motif sites (represented by features described above) in each sequence as input, and decodes the state of each site by the Viterbi algorithm to maximize the joint likelihood of the hidden states and the observed variables (13).

$$\begin{aligned} & P(s_1, s_2, \dots, s_N, \mathbf{x}_{1,l_1}, \mathbf{x}_{2,l_2}, \dots, \mathbf{x}_{N,l_N} | \lambda) \\ &= P(s_1, s_2, \dots, s_N | \lambda) P(\mathbf{x}_{1,l_1}, \mathbf{x}_{2,l_2}, \dots, \mathbf{x}_{N,l_N} | s_1, s_2, \dots, s_N, \lambda) \\ &= P(s_1) \prod_2^N P(s_i | s_{i-1}) \prod_1^n P(\mathbf{x}_{i,l_i} | s_i) \end{aligned} \quad (2)$$

RBP-bound clusters are defined to be a consecutive series of states $S^+, +, \dots, +$, or $I^+, +, \dots, +$, and ranked by the log-likelihood ratio:

$$\begin{aligned}
C &= \log \left\{ \frac{P(\mathbf{x}_{1,l_1}, \mathbf{x}_{2,l_2}, \dots, \mathbf{x}_{n,l_n} \mid s_1 = I+, s_2 = +, \dots, s_n = +)}{P(\mathbf{x}_{1,l_1}, \mathbf{x}_{2,l_2}, \dots, \mathbf{x}_{n,l_n} \mid s_1 = I-, s_2 = -, \dots, s_n = -)} \right\} \\
&= \log \left\{ P(\mathbf{x}_{1,l_1} \mid s_1 = I+) / P(\mathbf{x}_{1,l_1} \mid s_1 = I-) \right\} \\
&\quad + \sum_{i=2}^n \log \left\{ P(\mathbf{x}_{i,l_i} \mid s_i = +) / P(\mathbf{x}_{i,l_i} \mid s_i = -) \right\}
\end{aligned} \tag{3a}$$

for a cluster in which the first site is the very first site in the input sequence, or otherwise

$$\begin{aligned}
C &= \log \left\{ \frac{P(\mathbf{x}_{k,l_k}, \mathbf{x}_{k+1,l_{k+1}}, \dots, \mathbf{x}_{k+n-1,l_{k+n-1}} \mid s_k = I+, s_{k+1} = +, \dots, s_{k+n-1} = +)}{P(\mathbf{x}_{k,l_k}, \mathbf{x}_{k+1,l_{k+1}}, \dots, \mathbf{x}_{k+n-1,l_{k+n-1}} \mid s_k = I-, s_{k+1} = -, \dots, s_{k+n-1} = -)} \right\} \\
&= \log \left\{ P(\mathbf{x}_{k,l_k} \mid s_k = I+) / P(\mathbf{x}_{k,l_k} \mid s_k = I-) \right\} \\
&\quad + \sum_{i=k+1}^{k+n-1} \log \left\{ P(\mathbf{x}_{i,l_i} \mid s_i = +) / P(\mathbf{x}_{i,l_i} \mid s_i = -) \right\}
\end{aligned} \tag{3b}$$

where k (>1) is the first motif site in a cluster and n is number of motif sites of the cluster.

Supplementary references

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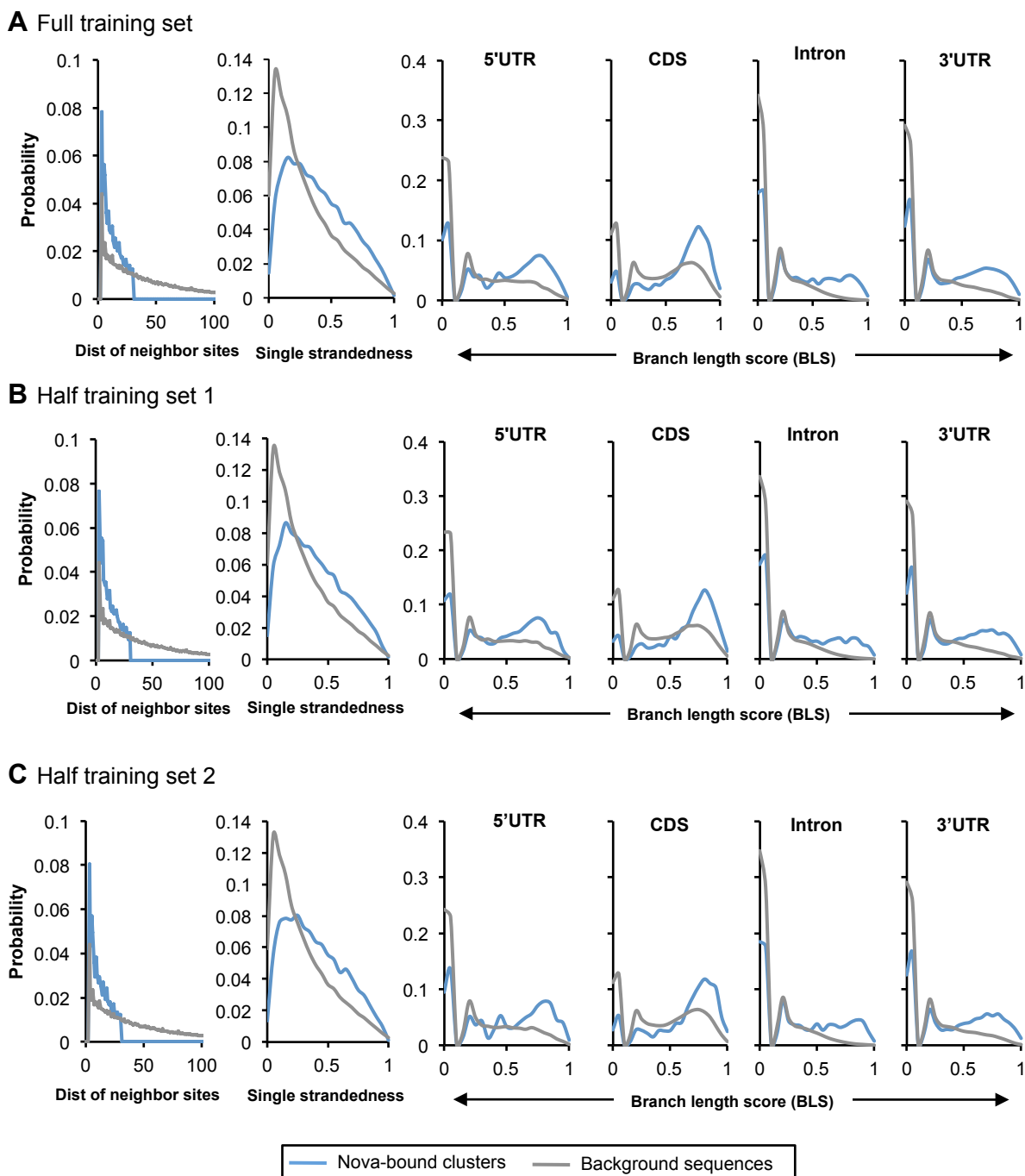


Fig. S1: (related to Fig. 1B in the main text) Emission probability distributions of spacing of Nova-binding motif sites, their accessibility and conservation.

A. Distributions estimated from the full training set, which is the same as shown in Fig. 1B in the main text. **B,C.** The whole transcriptome was split into two halves, and training data in each half was used to estimate the distributions. See Fig. 1B legends in the main text for more details.

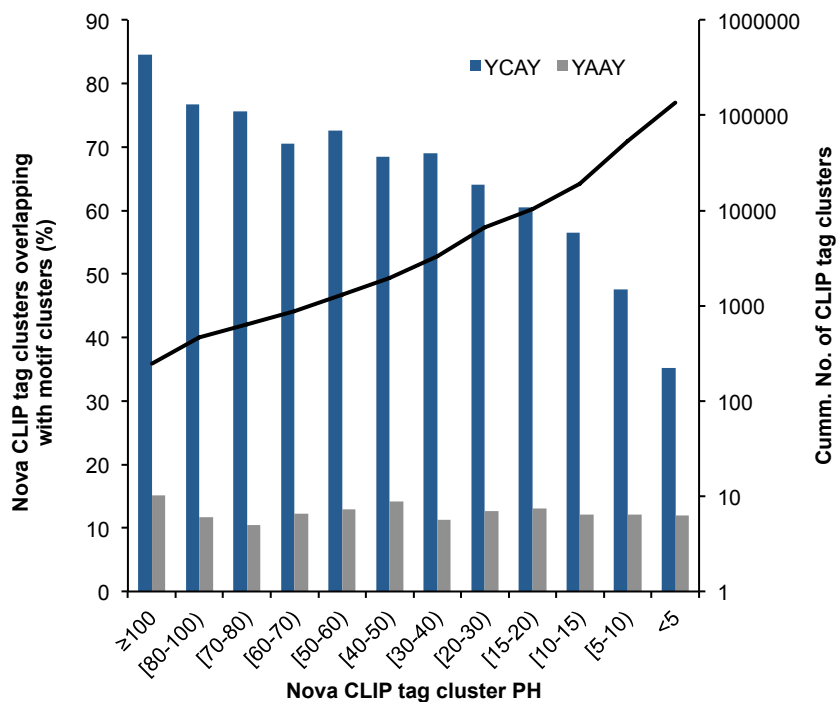


Fig. S2: The overlap between predicted YCAY clusters and CLIP tag clusters with varying peak heights (PH).

Non-repetitive CLIP tag clusters are binned according to PH and compared to YCAY clusters. For each bin, the proportion of CLIP tag cluster footprints (± 50 nt around peak) that overlap with YCAY clusters (blue bars), or control YAAY clusters (gray bars), is shown on the left axis. The cumulative number of non-repetitive Nova CLIP tag clusters is shown as the black curve (right axis).

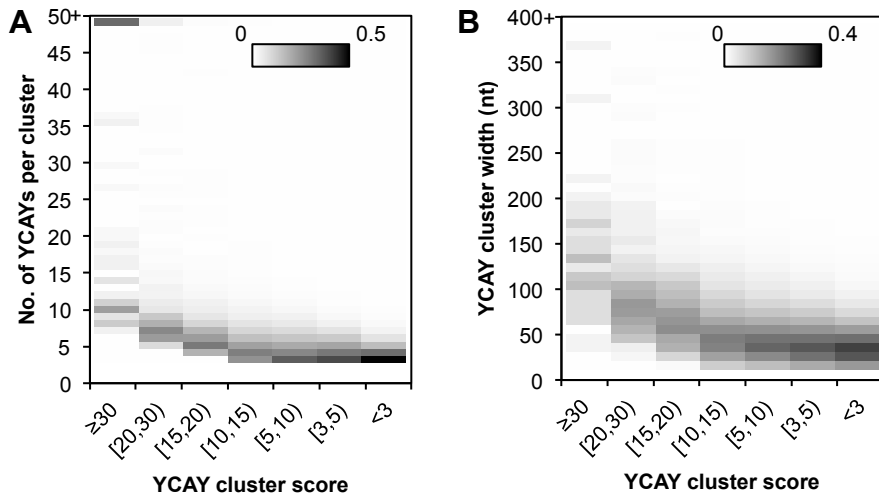


Fig. S3: General characteristics of predicted YCAY clusters.

A. YCAY clusters are divided into bins with different scores. The probability of clusters with a specific number of YCAY elements is calculated for each bin separately, and is represented in gray scale.

B. Similar to (A), but the probability of specific cluster width is shown.

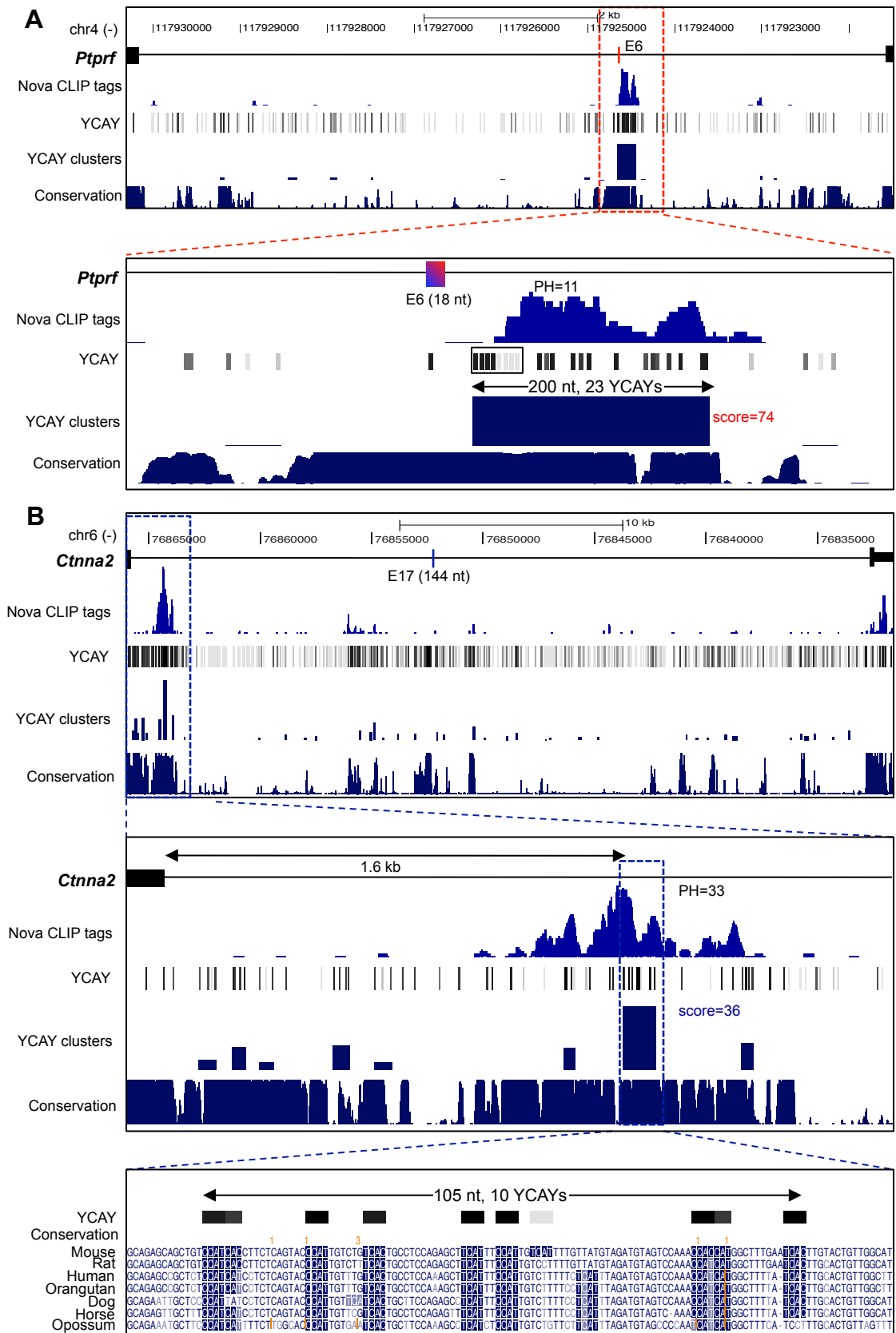


Fig. S4 (continued next page).

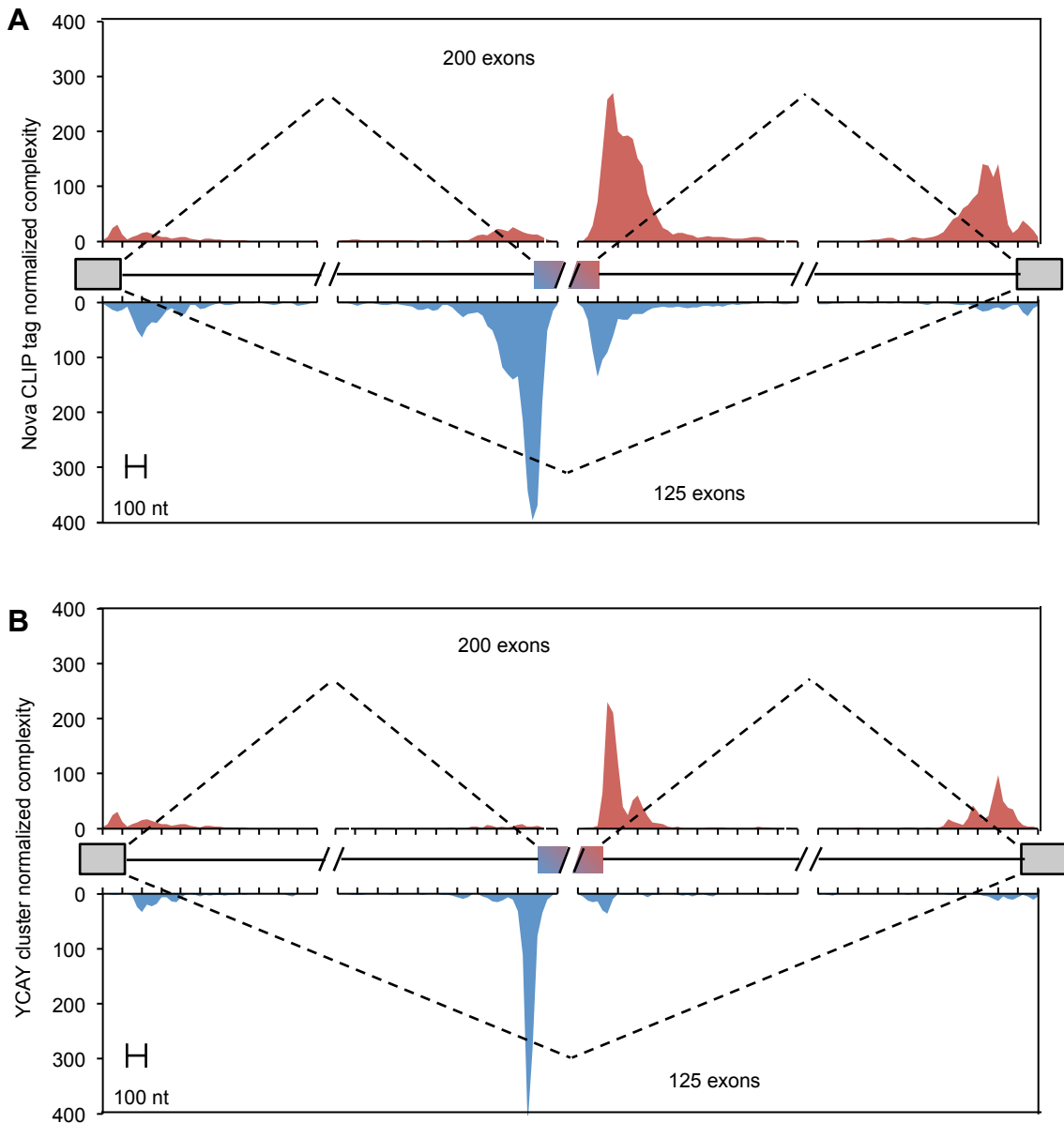


Fig. S5: The RNA map of Nova regulated alternative splicing.

A set of 325 non-redundant cassette exons regulated by Nova are shown.

A. The normalized complexity map of Nova CLIP tags for cassette exons activated (red) or repressed (blue) by Nova is shown. The number of exons in each group is indicated.

B. The normalized complexity map of YCAY clusters is displayed similar to panel (A).

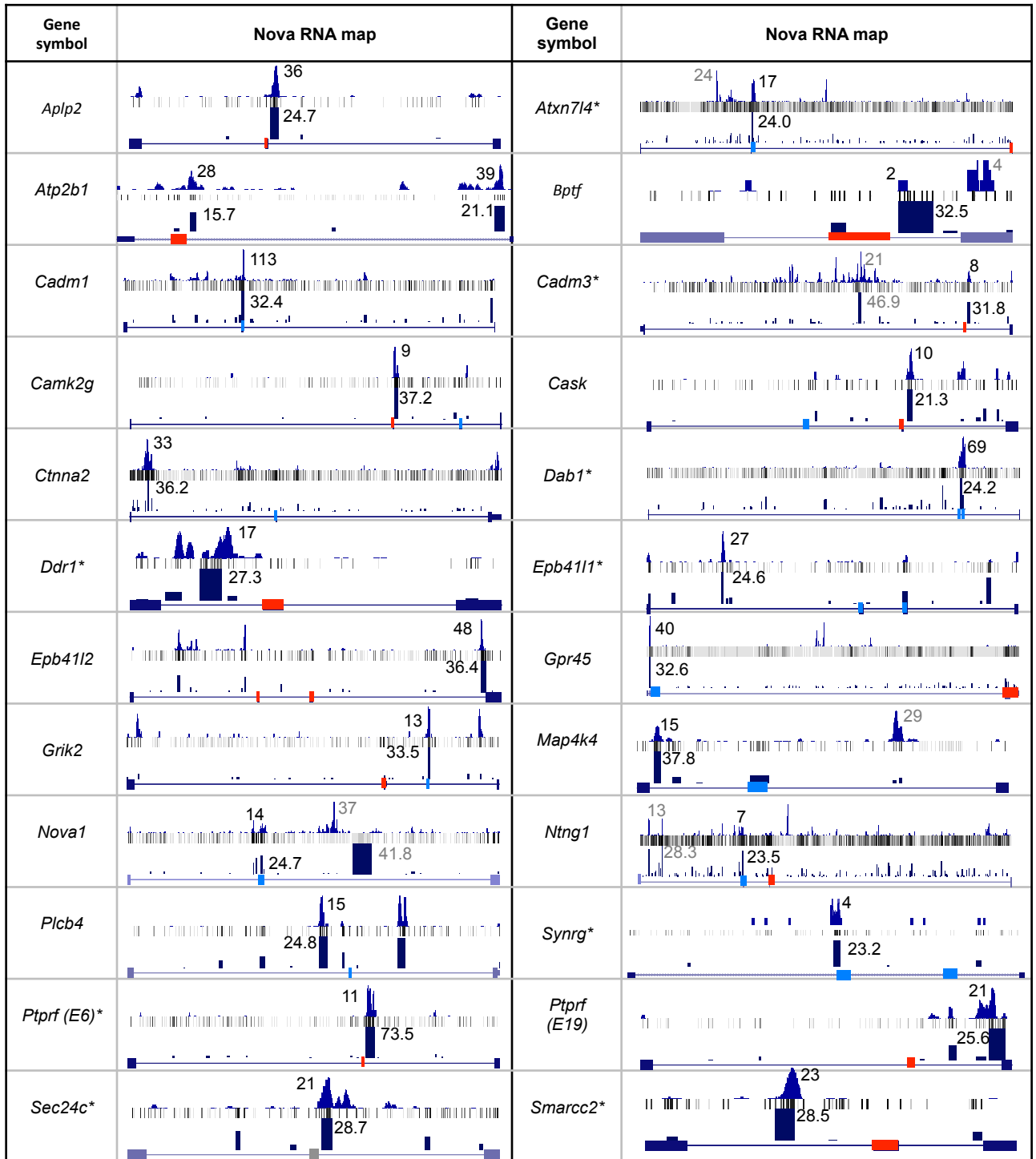


Fig. S6 legend (next page).

Fig. S6: Predicting Nova regulated alternative exons using YCAY clusters.

Exons analyzed in ref. (Ule et al. 2006 *Nature*, 444:580-586) and predicted based on mCarts YCAY cluster score (≥ 21.1) are shown. In each panel, the three tracks from top to bottom are distribution of CLIP tags, mCarts YCAY cluster scores, and gene structure in the alternatively spliced region. The highest-scoring YCAY clusters and major CLIP tag cluster peaks are indicated. Alternative exons with Nova-dependent inclusion and exclusion are shown in red, and blue, respectively. Exons above the “net score” threshold (≥ 2.7) (Ule et al. 2006 *Nature*, 444:580-586) are indicated by asterisk.

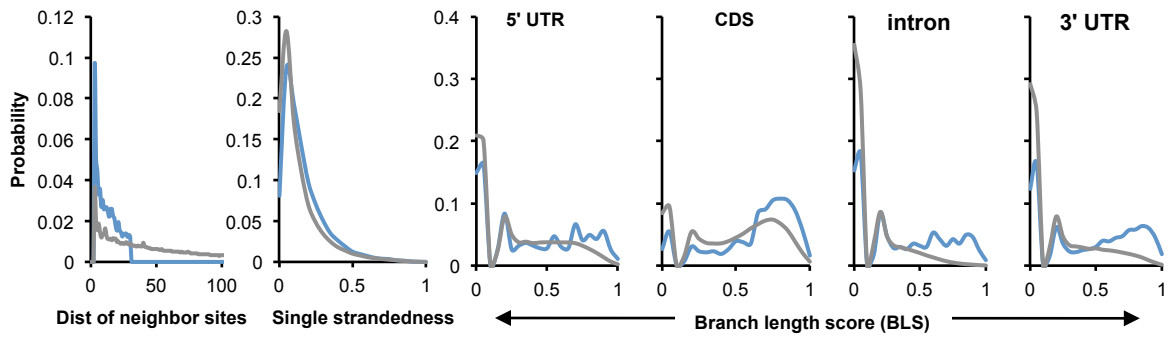


Fig. S7: Emission probability distributions of spacing of Mbni-binding motif sites, their accessibility and conservation.

See Fig. 1B legend in the main text for more details.

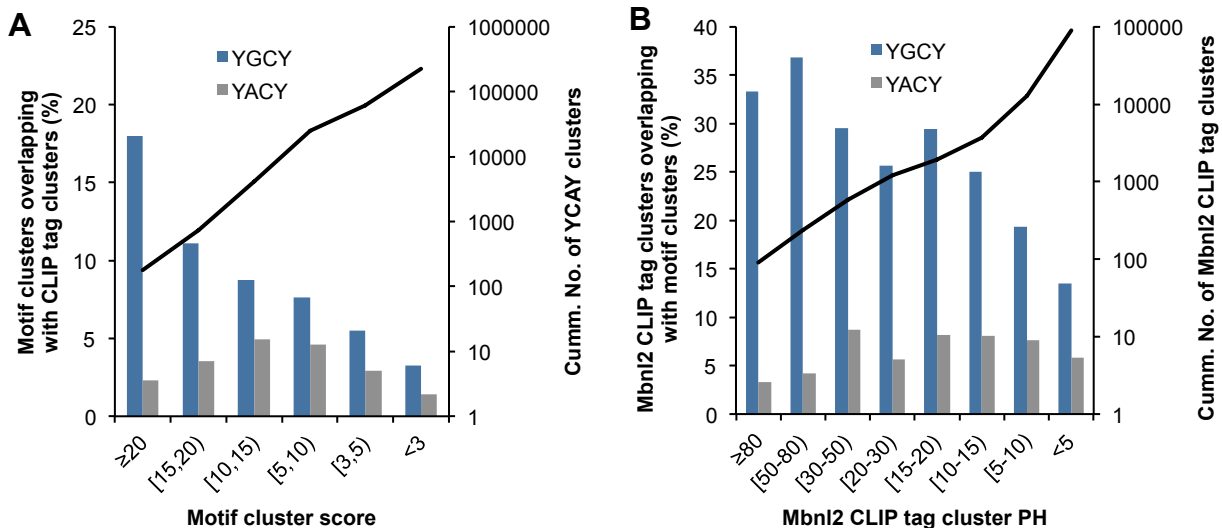


Fig. S8: The overlap between predicted YGCY clusters and Mbnl2 CLIP tag clusters depending on the stringency of each dataset.

A. The overlap between the footprints of Mbnl2 CLIP tag clusters and predicted YGCY clusters with varying scores. Non-repetitive YGCY clusters are binned into groups according to their scores. For each bin, the proportion of YGCY clusters overlapping with CLIP tag cluster footprints (± 50 nt around peak) is shown (blue bars, left axis). YACY clusters predicted by the same model are shown (gray bars, left axis) as a control. The cumulative number of non-repetitive YGCY clusters is shown as the black curve (right axis).

B. The overlap between predicted YGCY clusters and Mbnl2 CLIP tag clusters with varying peak heights (PH). Non-repetitive CLIP tag clusters are binned according to PH and compared to YGCY clusters. For each bin, the proportion of CLIP tag cluster footprints that overlap with YGCY clusters (blue bars), or control YACY clusters (gray bars), is shown on the left axis. The cumulative number of non-repetitive Mbnl2 CLIP tag clusters is shown as the black curve (right axis).

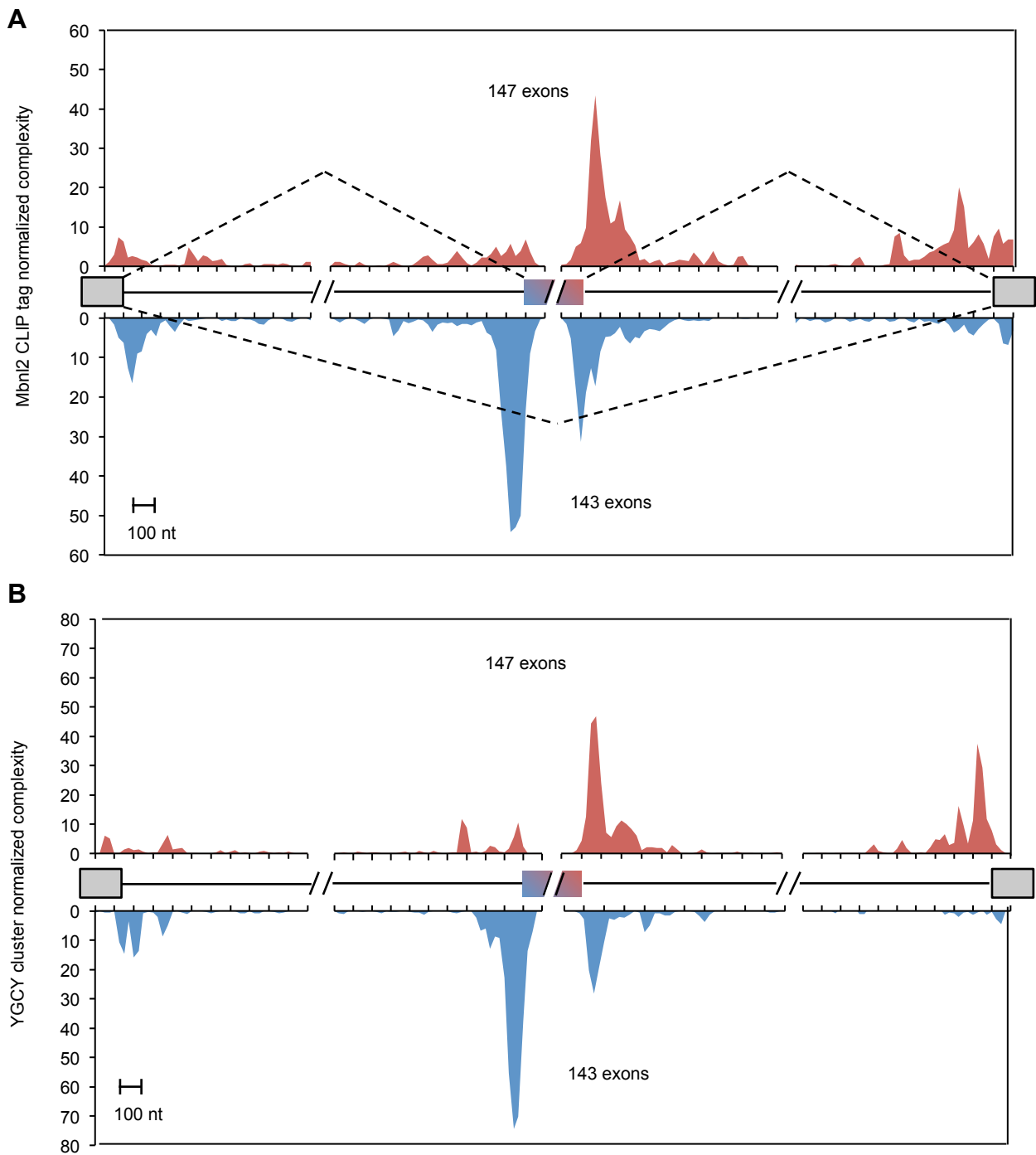


Fig. S9: The normalized complexity map of Mbn2-regulated alternative splicing.
See Fig. S5 legend for more details.

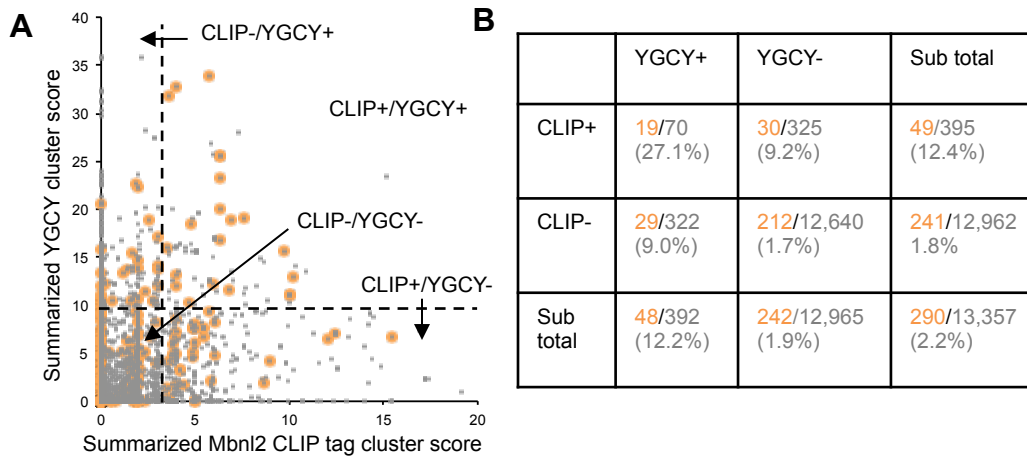


Fig. S10: Mbnl2-regulated alternative exons predicted from CLIP data and those predicted from YGCY clusters are complementary to each other.

A. Target exon scores predicted from CLIP data (x-axis) are plot against scores predicted from YGCY clusters (y-axis). Each dot is an exon. All cassette exons are shown in gray, and exons with Mbnl2-dependent splicing as determined by Affymetrix exon-junction microarray or RNA-Seq data are overlaid in orange. An arbitrary threshold of summarized CLIP tag cluster score (3.8) or YGCY cluster score (10) is indicated by the dotted lines.

B. Breakdown of exons according to their summarized CLIP tag cluster score or YGCY cluster score above or below the threshold. The number (orange) and percentage of exons with Mbnl2-dependent splicing in each category are also shown.