

**Transferable deep learning enables identification of multiple types of RNA
modifications using nanopore direct RNA sequencing**

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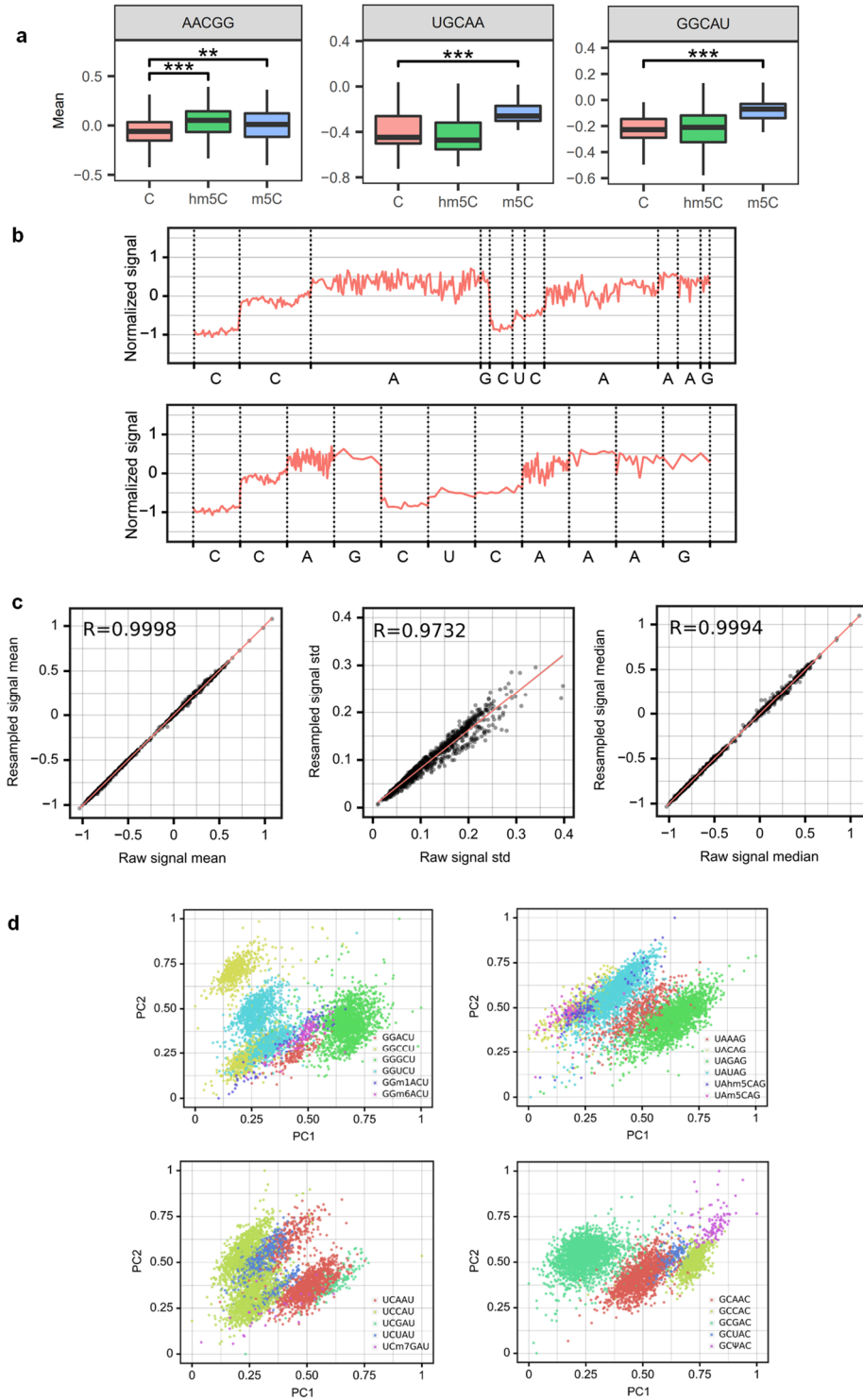
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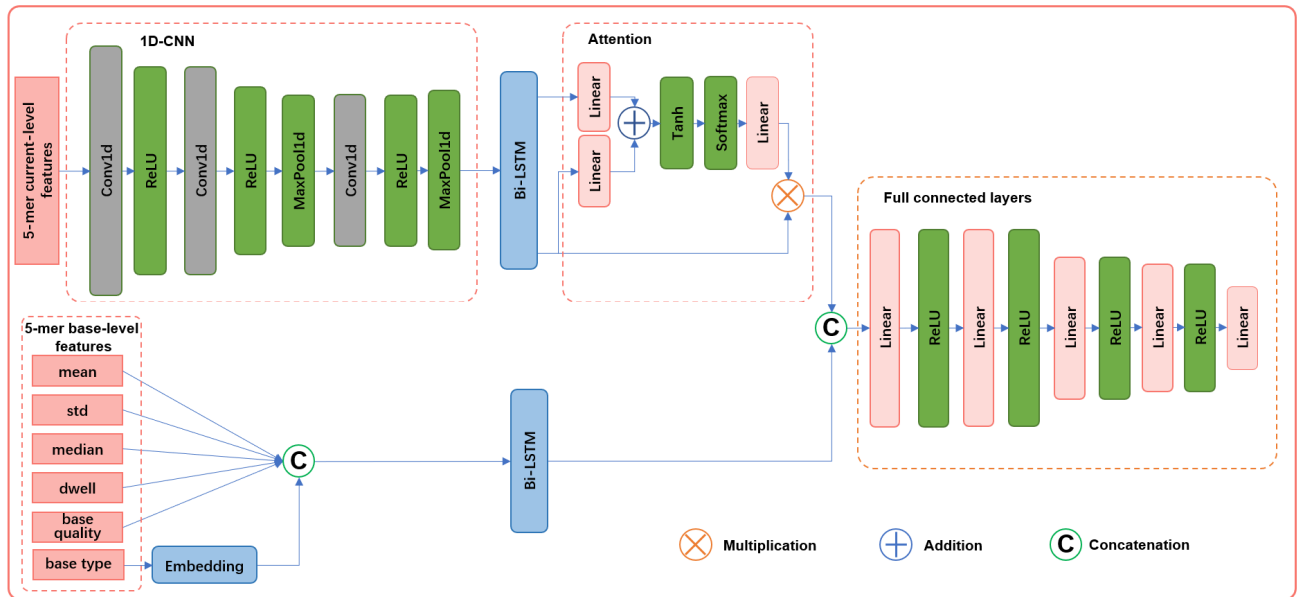
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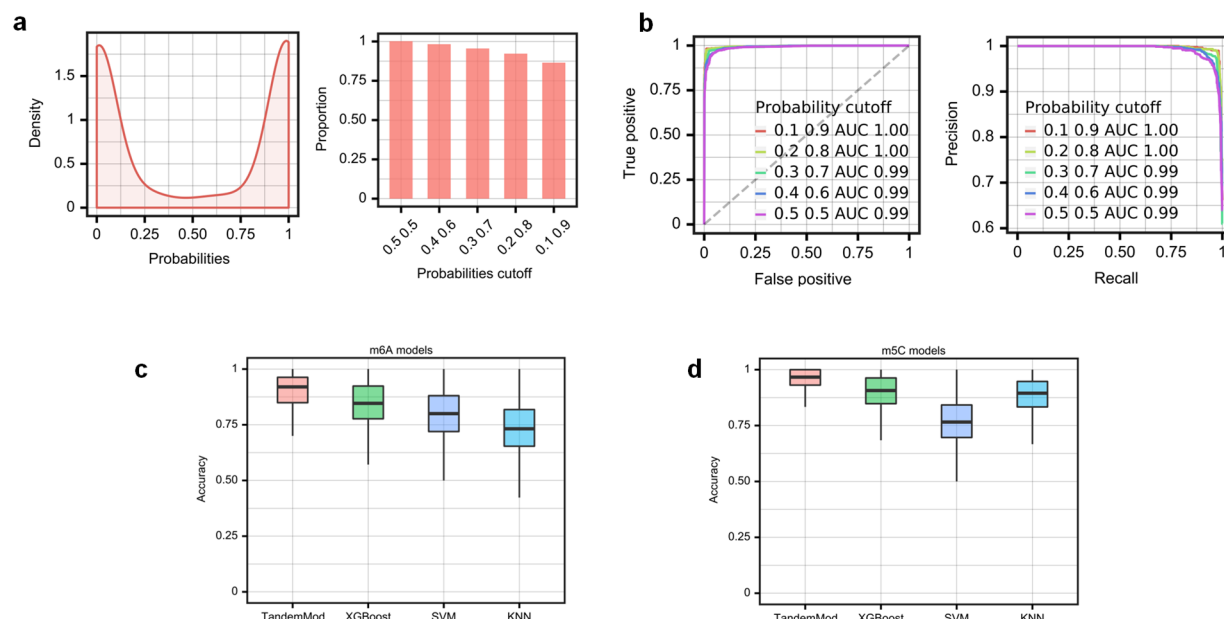
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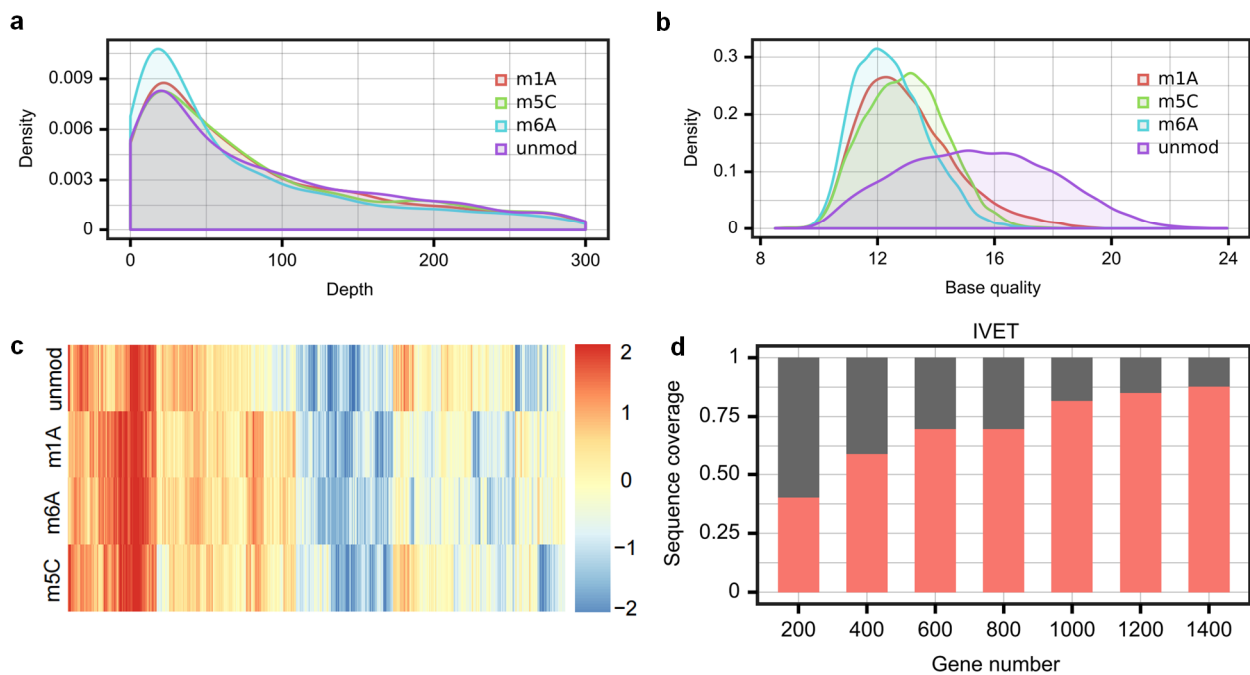
Supplementary Fig. 1: DRS features at base level and current level affected by 5-mer sequence contexts. **a**, Mean signatures of current signals at C, hm⁵C and m⁵C under different 5-mer sequence contexts. **b**, An example showing normalized current signals with raw signal width (top panel), and resampled current signals with equal width (bottom panel). **c**, Scatterplot of base-level features including mean, std and median between original signals and resampled signals. R value indicates the correlation coefficient. **d**, Scatter plots displaying the difference between the modified bases and the four canonical bases after PCA transform, with the two bases before and after the target base being the same.



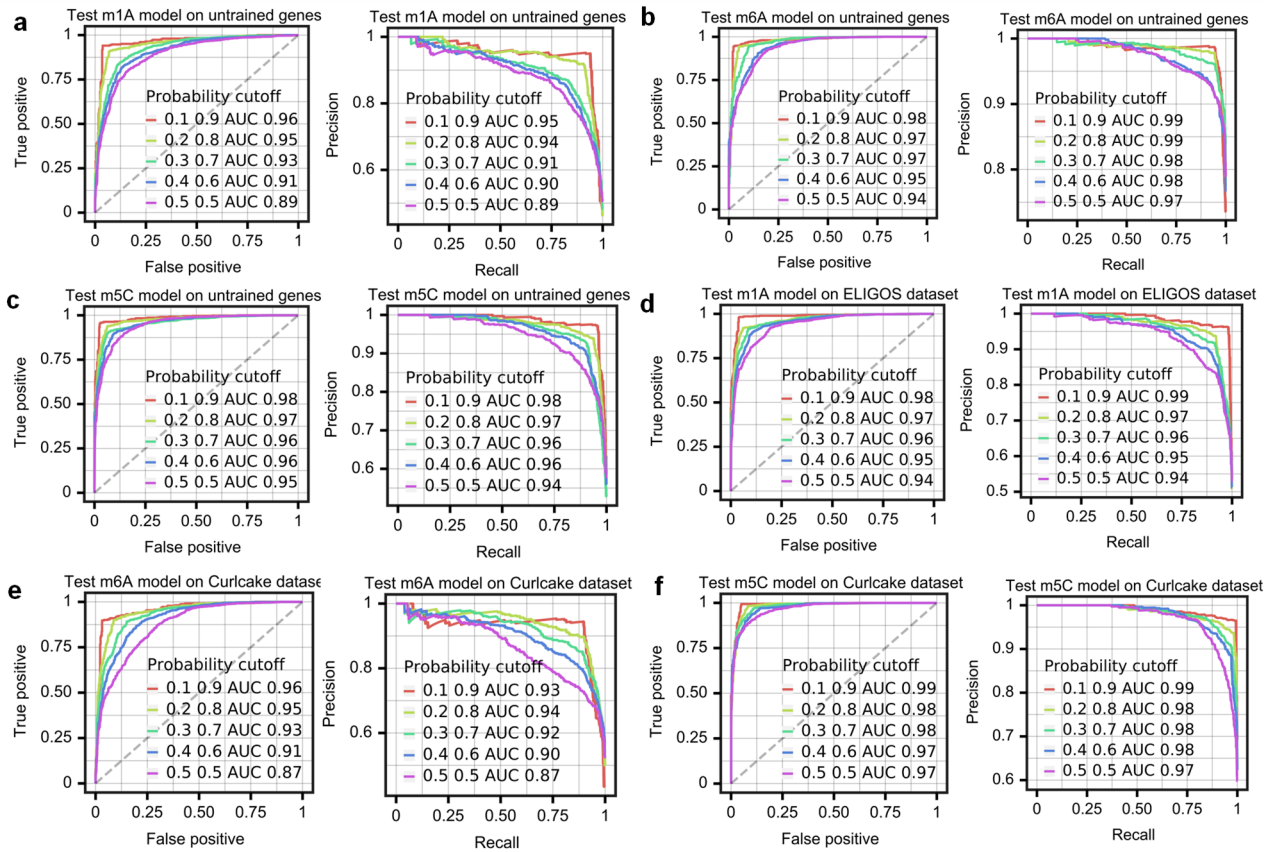
Supplementary Fig. 2: Detailed architecture of TandemMod model. TandemMod consists of 4 main components: a one-dimensional convolutional neural network(1D-CNN), a bi-directional long short-term memory (bi-LSTM) module, an attention mechanism and full-connected layers. The top layers of the model act as feature extractor that learns modification-related information and the bottom layers act as classifier that predict modification type.



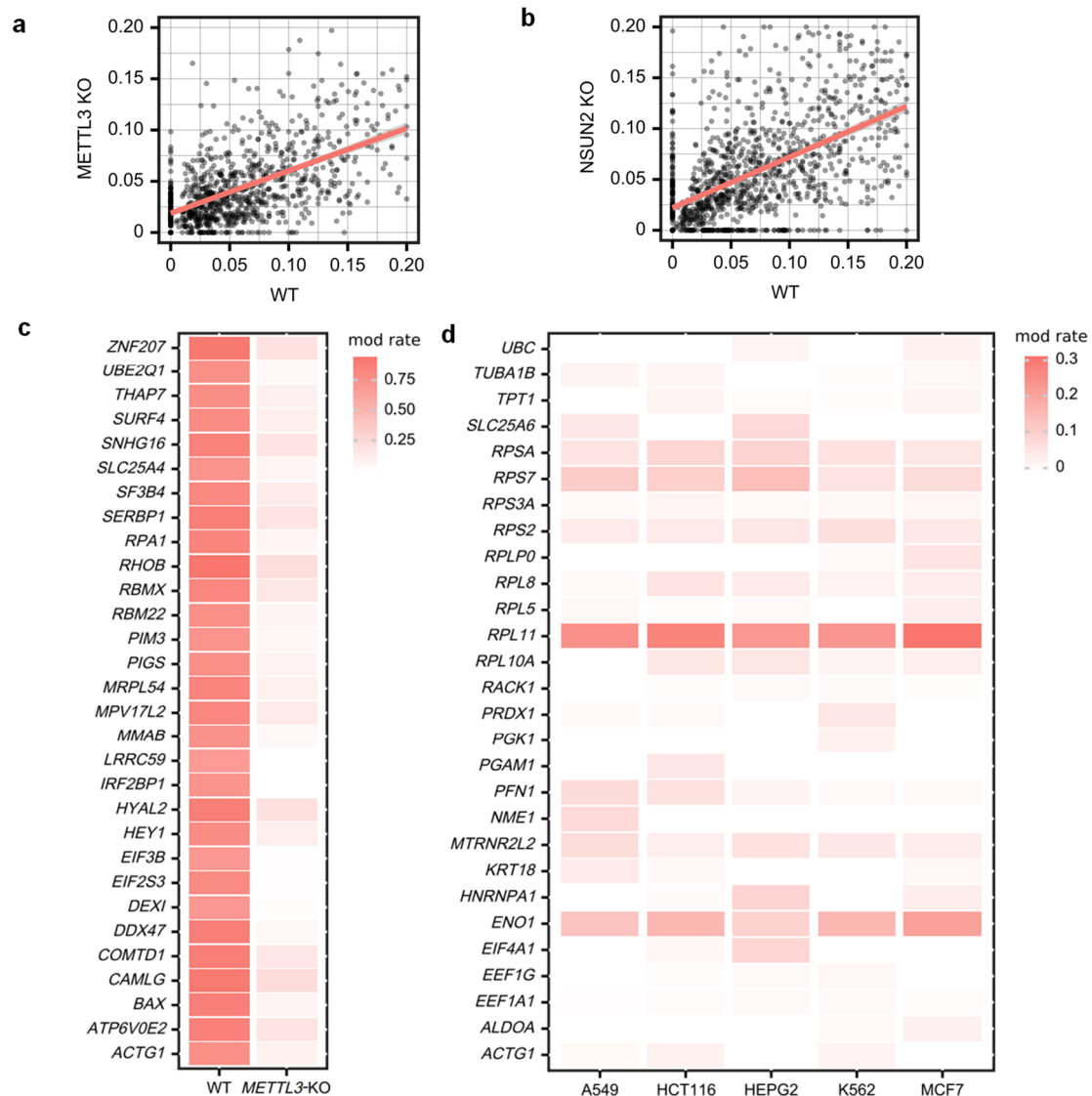
Supplementary Fig. 3: Performance evaluation of TandemMod. **a**, The distribution of predicted modification probabilities by the TandemMod model on the Curlcake testing set (left panel) and the relationship between the remaining reads and the cutoff probabilities (right panel). **b**, Performance for m⁵C identification model on Curlcake dataset using different cutoff values. **c**, Comparison of models trained on Curlcake m⁵C dataset and tested on Curlcake testing set. **d**, Comparison of models trained on Curlcake m⁶A dataset and tested on Curlcake testing set.



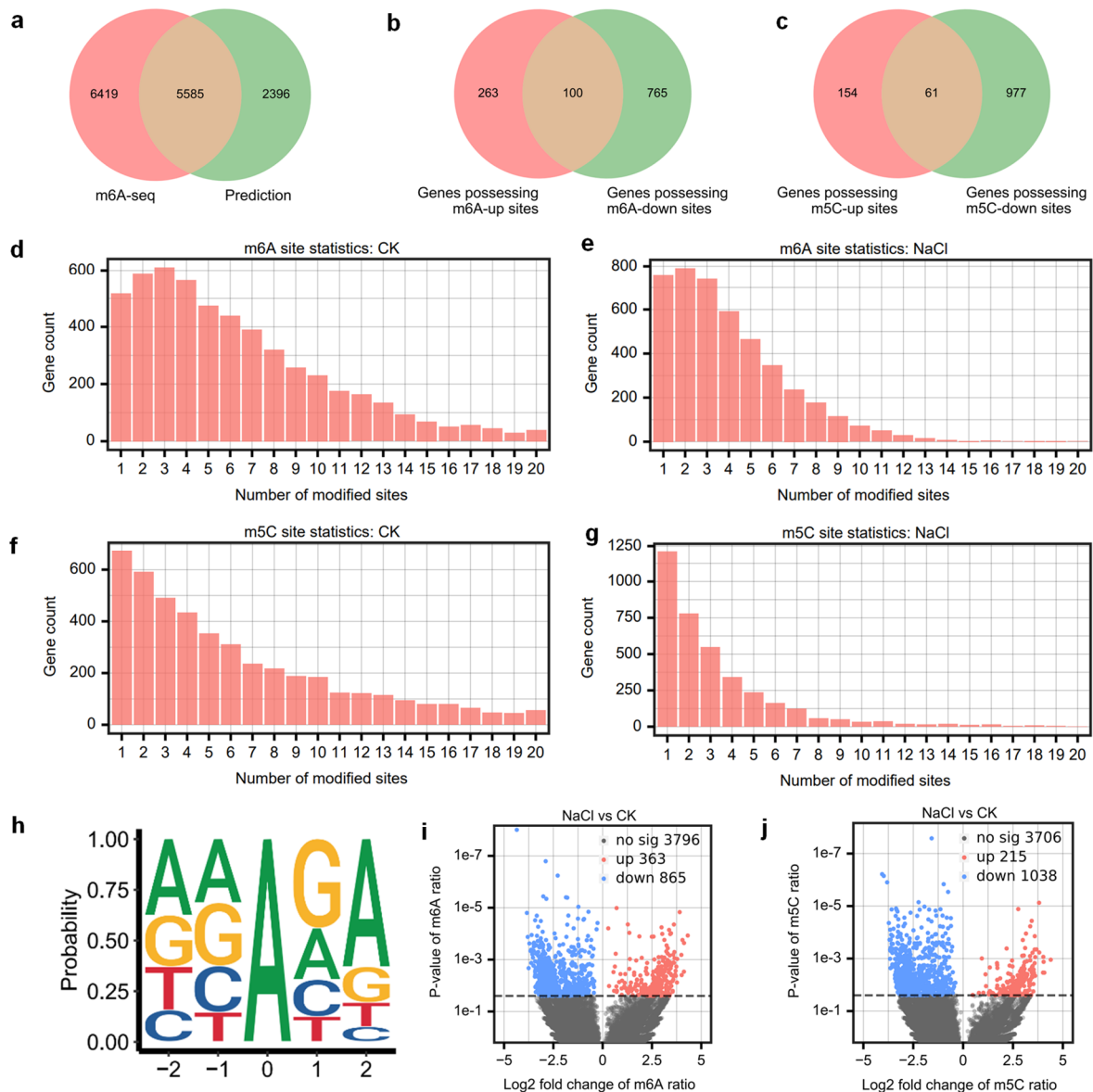
Supplementary Fig. 4: Statistics of direct RNA sequencing results from the IVET datasets. **a**, The sequencing depth of the four IVET datasets. **b**, The mean base quality of the four IVET datasets. **c**, Heatmap displaying the transcribed mRNA abundance across the four IVET samples. **d**, The barplot showing the 9-mer sequence coverage across different numbers of genes in the IVET dataset.



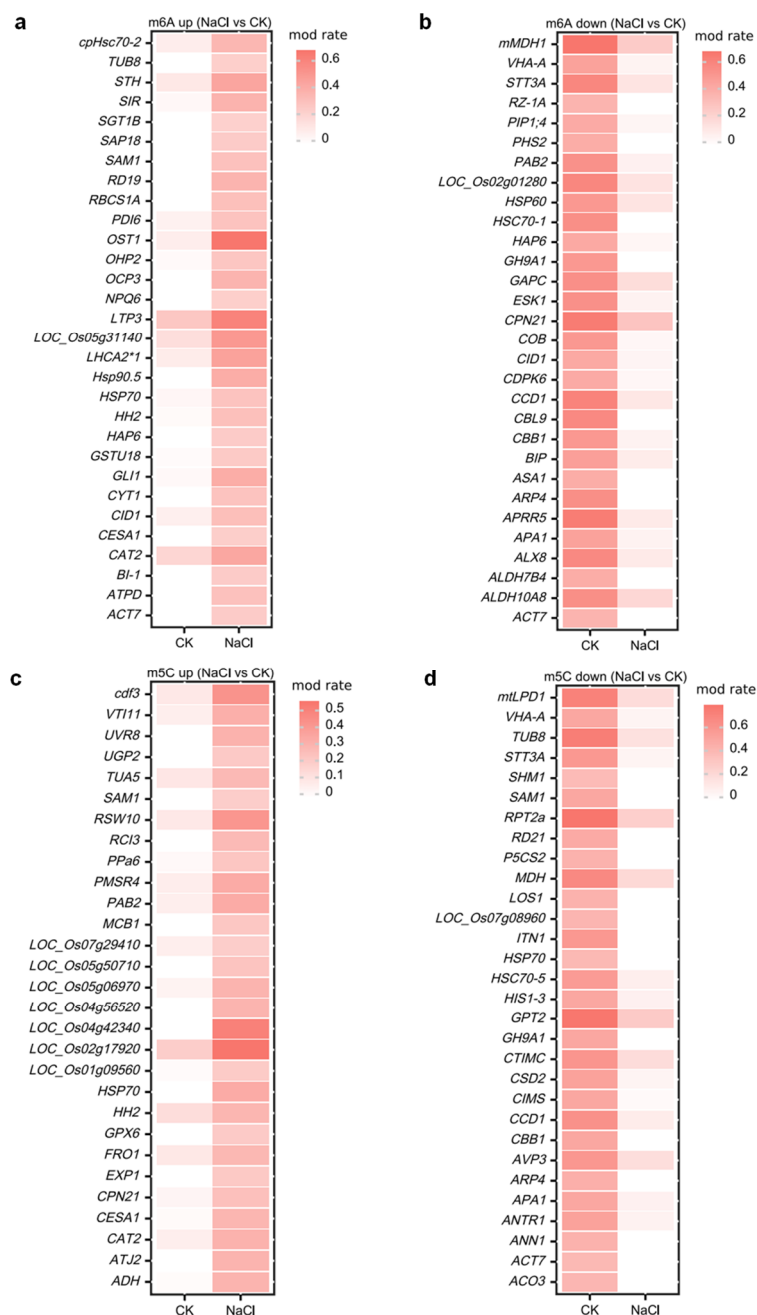
Supplementary Fig. 5: Performance evaluation of TandemMod models trained on IVET datasets. **a-c**, Performance evaluation of the m¹A model (**a**), m⁶A model (**b**) and m⁵C model (**c**) trained on the IVET training sets and tested on the IVET independent genes. **d**, Performance evaluation of the m¹A model trained on the IVET training set and tested on ELIGOS dataset. **e**, Performance evaluation of the m⁶A model trained on the IVET training set and tested on the Curlicake dataset. **f**, Performance evaluation of the m⁵C model trained on the IVET training set and tested on the Curlicake dataset.



Supplementary Fig. 6: Identification of m⁶A sites across different tissues in human. **a**, Scatter plot showing decreased m⁶A level in *METTL3*-KO HEK293T sample compared to WT sample. **b**, Scatter plot showing decreased m⁵C level in *NSUN2*-KO HeLa sample compared to WT sample. **c**, Heatmap of top 30 genes with differentially m⁶A-modified rates between *METTL3*-KO and WT samples. **d**, Heatmap showing m⁶A rates of randomly selected genes across 5 human cell lines.



Supplementary Fig. 7: Identification of m⁶A and m⁵C sites across different conditions in rice. **a**, Venn diagram showing gene with TandemMod-predicted m⁶A sites and genes detected by m⁶A-seq. **b**, Venn diagram of genes containing m⁶A sites with increased and decreased modification rate. **c**, Venn diagram showing of genes containing m⁵C sites with increased and decreased modification rate. **d-e**, Statistics of genes possessing different number of m⁶A-modified sites in CK (**d**) and NaCl-treated sample (**e**). **f-g**, Statistics of genes possessing different number of m⁵C-modified sites in CK (**f**) and NaCl-treated sample (**g**). **h**, Sequence motif of m⁶A-modified sites enriched in rice control sample in addition to DRACH. **i**, Volcano plot showing differentially m⁶A-modified genes. **j**, Volcano plot showing differentially m⁵C-modified genes. Chi-square test was performed and sites with Chi-square p-value less than 0.05 were considered as differentially modified.



Supplementary Fig. 8: Differentially m⁶A-modified and m⁵C-modified genes identified by TandemMod in NaCl-treated rice as compared to the control. a, Genes with increased m⁶A modification rate in NaCl-treated sample. b, Genes with decreased m⁶A modification rate in NaCl-treated sample. c, Genes with increased m⁵C modification rate in NaCl-treated sample. d, Genes with decreased m⁵C modification rate in NaCl-treated sample.