1   VIDEO - Visual Integration of Drosophila Enhancer Organization: A tool for integrating and visualizing

2   chromatin accessibility, in vivo transcription factor binding and motif occurrence in tissue-specific

3   differentially expressed genes

7   Running title: A Tool for Visualizing Drosophila Proximal Enhancers

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ABSTRACT

Dissecting gene regulation today relies on many genomic assays – including transcriptional output from

RNA-seq, chromatin accessibility from ATAC-seq, and transcription factor (TF) binding from ChIP-seq.

Whereas numerous tools exist for each modality and some integrate data across modalities, few allow

researchers to interactively explore and visualize how TF binding motifs intersect with transcriptional

activity and chromatin accessibility in a tissue-specific context. Here we introduce VIDEO (Visual

Integration of Drosophila Enhancer Organization), a web-based analysis tool that enables visualization of

conserved TF binding motifs within proximal promoters of genes differentially expressed in specific

tissues. Starting with gene lists derived from in situ hybridization, microarray, and/or scRNA-seq studies

of WT or mutant samples, one can identify the TFs expressed in each tissue and learn if and where the

consensus binding motifs for those TFs are found within the proximal enhancers of a custom gene set.

This pipeline also allows for coincident visualization of active chromatin, as determined from ATAC-seq

data, and for the visualization of DNA binding data from ChIP-seq datasets for specific TFs. To

demonstrate its utility, we apply VIDEO to the well-characterized regulatory system of CrebA and the

secretory pathway in the *Drosophila* salivary gland. We also explore a lesser-known system in the

embryonic hindgut to show how utilization of this tool can serve to generate hypotheses regarding

regulatory interactions.

INTRODUCTION

The application of recent advances in high-throughput sequencing have revealed large sets of differentially expressed (DE) genes from multiple *Drosophila* cell types at distinct developmental stages and under a variety of mutant conditions (KARAISKOS *et al.* 2017; CALDERON *et al.* 2022; SEROKA *et al.* 2022; SAKAGUCHI *et al.* 2023; PENG *et al.* 2024; JACKSON *et al.* 2025; WANG *et al.* 2025). Related technologies have likewise generated extensive *Drosophila* chromatin accessibility and transcription factor (TF) binding datasets with associated binding site consensus motif data (BERGMAN *et al.* 2005; ZHU *et al.* 2011; SHAZMAN *et al.* 2014; NITTA *et al.* 2015; KUDRON *et al.* 2024; RAULUSEVICIUTE *et al.* 2024). These motifs, 6-12 bp long sequences recognized by specific transcription factors, are a critical component of gene regulatory networks, and mapping their patterning across the genome is crucial to formulating hypotheses about TF-target gene interactions. A challenge that remains is connecting these datasets to identify which TFs may regulate a given set of DE genes and where their binding motifs occur in open chromatin.

Existing workflows for motif analysis extract peak regions from ATAC-seq (open chromatin) and ChIP-seq (TF occupancy) data, convert them to FASTA sequences, and run motif discovery tools such as MEME (BAILEY AND ELKAN 1994), HOMER (HEINZ *et al.* 2010) and RSAT (THOMAS-CHOLLIER *et al.* 2012). Single-cell RNA-seq further expands possible motif-analysis inputs by generating lists of genes of interest from differential expression analysis. For all inputs, users must retrieve the peak or promoter/enhancer sequences, wrangle multiple file formats, and merge motif hits with chromatin or binding tracks using command-line tools (e.g. bedtools) and write custom scripts for visualization. This fragmented, multi-step workflow and the lack of user-friendly visualization tools often leaves motif data under-utilized in the context of gene regulatory network studies, and there is still a lack of an intuitive way to visualize and explore TF-motif landscapes in their native genomic context.

Existing visualization tools address parts of the pipeline; MEME-Suite (BAILEY *et al.* 2015) allows for a basic position display for motifs it discovers but does not accommodate externally defined motifs. TFMotifView (LEPORCQ *et al.* 2020) allows browsing of a pre-computed motif library in user-defined genomic regions, but it does not take as input user-generated motifs in standard formats (PFM, PWM, PCM, IUPAC as described in **Supplemental Table S1**) and it does not support the latest *Drosophila* genome assemblies. Additionally, the output of these existing tools fails to utilize the potential of overlaying TF motifs with user-discovered gene lists and sequencing data.

To address these shortcomings, we have built VIDEO, a web-based platform designed specifically for use

with *Drosophila,* with a streamlined, user-friendly graphical interface that requires no programming

knowledge. VIDEO integrates processed ATAC-seq and ChIP-seq peak information into signal tracks with

genomic input, integrates motif discovery while also accepting user-defined motifs in any common

format, and displays motif hits alongside genomic signal overlays. VIDEO allows users to rapidly explore

many transcriptional regulation hypotheses and questions such as: Are the motifs of the TF of interest

enriched in the user-provided list? Do they occur more densely in the list compared to a control group?

Which isoform/transcription start site (TSS) is the richest in the TF motif? Do motif hits occur in regions

of accessible chromatin? Which of these hits are supported by DNA binding data? Currently, there is no

single public resource that can be mined to address all these questions. VIDEO also serves as a data bank

for users who want to extract processed scRNA sequencing data for *Drosophila* embryos (stage 10-16,

(PENG *et al.* 2024) and find TF motifs of interest, enabling both an end-to-end pipeline for initial

hypothesis generation, and a much more detailed and focused pipeline for exploring motif occurrences

in user-generated gene lists with supporting ChIP-seq and ATAC-seq data. VIDEO can be accessed

through this link: VIDEO-motif-tool.

MATERIALS AND METHODS
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Input data and preprocessing

User inputs consist of either a gene list of FlyBase IDs or official gene symbols or a BED file with genomic

peak regions. In the "compare" mode users can input two gene lists with distinct labels and perform the

same downstream analysis. For gene-list inputs, transcript annotations for all isoforms and ±500 bp (or a

user-defined window size up to 2 kb) flanking sequences around each TSS were indexed from a copy of

the *Drosophila melanogaster* reference genome (dm6) FASTA using pyfaidx (v0.8.1.4). For BED-file

inputs, peak coordinates were used directly to extract the corresponding FASTA sequences.

Motif scanning

User-defined motifs (PFM, PWM, PCM, or IUPAC formats) were scanned against the extracted

sequences using FIMO (v4.12.0; (GRANT *et al.* 2011)) with a default significance threshold of p < 1e-4. A

custom Python wrapper launched FIMO as a subprocess, parsed its output, and collated motif hits into a

unified table recording genomic coordinates, strand, p-values, q-values, and matched sequences.

Visualization of motif occurrences

Motif occurrences were plotted with Plotly (v6.2.0) in python. For each gene or peak region, hits were

positioned relative to the TSS (zero), spanning ±500 bp by default. The opacity of each hit reflects its

FIMO match score, with darker hits indicating a stronger match. Hits with negative log-likelihood ratio

scores were excluded, as they are more likely to represent random matches and correspond to higher p-

values. Motif hits are ordered either by the user's input list or by total motif count or mean match score,

as selected in the interface. The plot is interactive and all plotting defaults are configurable in the web

user interface (UI). Match stringency is an adjustable variable and users can further adjust the p-values

for each motif specifically.

Data Integration

To filter motif hits by chromatin context, ATAC-seq and ChIP-seq peaks were integrated via Pybedtools

(v0.12.0) calling bedtools intersect with a minimum overlap of 50% of the motif length.

BigWig Track Overlay

To visualize chromatin accessibility and transcription factor occupancy in the same genomic window as

motif hits, BigWig tracks from ATAC-seq or ChIP-seq experiments were uploaded. For each selected peak

or gene, coverage values are extracted from the corresponding chromosome interval using pyBigWig

(v0.3.22). Coverage is plotted in Plotly alongside motif positions, with the x-axis centered on the peak

midpoint or TSS and consistent y-axis scaling across all uploaded tracks to facilitate comparison. Motif

hits from the motif occurrence plot are overlaid in the top panel and strand direction is accounted for.

VIDEO implementation

VIDEO was implemented using FastAPI (v0.95.2) backend and a React.js (v18.2.0) frontend. Core

dependencies include FIMO (v4.12.0), pyfaidx (v0.5.9), Pybedtools (v0.12.0), and Plotly (v6.2.0). The

application was deployed on an EC2 Instance with Redis as broker. Reference genome FASTA and GTF

annotation (dm6, release 6.45) were downloaded from FlyBase; motif libraries were bulk downloaded

from JASPAR (RAULUSEVICIUTE *et al.* 2024).

Data and code availability

All the code used for developing the tool is available at https://github.com/vidyaajay1/VIDEO-motif-

tool.

Embryo immunostaining

Embryos were fixed and immunostained as described (REUTER *et al.* 1990). Rat αRetn (HABERMAN *et al.*

2003) was used at a 1/20,000 dilution. Biotin-conjugated Donkey αRat secondary was used at a dilution

of 1:2000.

 RESULTS

VIDEO presents an interactive, browser-based overview designed to guide users from raw inputs to

interpretable motif visualizations without the need for multi-tool command-line operations. Here, we

provide a flowchart that can be used as a guide for new users (**Fig 1**). Upon starting the application, the

landing page has a collapsible sidebar navigator that has two functional pages "Motif Viewer" and "TF

Finder" and a tutorial page with instructions (**Fig 2A**). The Motif Viewer has the pipeline for generating

the motif occurrence plot, while TF Finder is a helper page with functions intended to provide users with

a starting gene list and allow discovery of TFs and motifs of interest.

1. Using TF Finder to acquire gene and motif data

Identification of gene sets of interest

There are multiple ways a user can acquire a gene list of interest. These include, but are not limited to,

mining transcriptomic databases such as BDGP (TOMANCAK *et al.* 2007) for genes differentially expressed

(DE) in a specific tissue, extracting tissue-specific DE genes from scRNA-seq data, and identifying TF-

dependent gene sets from comparisons of WT and mutant scRNA-seq, microarray, and/or in situ data.

Users are encouraged to rank their gene lists in order of most important to least if there are available

metrics such as p-values or log fold change expression, because VIDEO can retain the input order while

plotting. To facilitate very straightforward explorations on DE gene sets, TF Finder contains a curated set

of DE genes for two stages of development of the *Drosophila* embryo from Peng et. al 2024 single cell

RNA sequencing data (**Fig 2B**). Users can choose the stage of development and the tissue, and download

a list of the top DE genes in that tissue.

Identify all TFs in a gene list

TF Finder has a helper function that filters any gene list for transcription factors (as reported in FlyBase

(OZTURK-COLAK *et al.* 2024)). Users can identify TFs differentially expressed in their tissue of interest by

choosing the "Use Top 250 DE Genes for Selected Tissue" option or upload their own gene list to extract

the TFs in it (**Fig 2C**). Once the gene list is filtered for TFs, VIDEO also displays the consensus motifs for

each TF (if available) in the JASPAR database with their motif ID, PWM and logo.

Identification of TF motifs or DE gene conserved motifs

If the user does not have a TF motif of interest already, or if the TF of interest does not have any

reported motifs shown in VIDEO, they can make use of the in-built "Motif Discovery for Genes" function

(**Fig 2D**). Again, the user can select the "Top 250 DE Genes for Selected Tissue" option or upload their

own gene list. Once the gene list is uploaded, the user can set a promoter region flanking the TSS and

the minimum and maximum motif width. If nothing is specified, the program will default to its preset

values (6 bp to 10 bp width). On clicking the "Run STREME" button, the motif discovery program will

output a link to view and download the motifs that are enriched in the list of genes. Once the motifs of

interest are decided, it is recommended that the user keep the motifs in PWM format (MEME format). If

the motifs were chosen from VIDEO, the displayed PWM can directly be copied to a clipboard and

pasted as motif input.

2. Using Motif Viewer to create motif hit plots

Setting up genomic input

This is the landing page of VIDEO, and in the regular mode users can upload their gene list or a BED file

with genomic regions of interest (**Fig 3A**). In the compare mode, users are required to upload two gene

lists and give each list a unique name, as elaborated further in the case study later. Once the gene lists

are uploaded, the next step is to choose a promoter window size flanking the TSS of each gene. If a gene

has multiple TSSs, all unique TSSs will be considered and annotated with the transcript ID. If no choice is

made, the program will default to the preset 500 bp upstream and downstream. The last step is to click

on the "Process Genomic Input" button, which validates the gene list and fetches all the transcript

annotations and sequences for downstream processing.

Setting up motif input

The motif input page requires users to drop in their motifs in any of three formats (IUPAC consensus

sequence, PWM, PFM, or PCM), give the motif a name, and choose a color for it for the motif

occurrence plot (**Fig 3B**). If users are dropping in PWM, PFM or PCM, they can either upload it as a .txt

file or copy and paste the matrix and then click "Parse Pasted Matrix". Once the matrix is parsed and all

the motifs are input, clicking the "Process Motifs" button will both validate the motifs and process it for

the plot.

Scanning for motif occurrences

After the gene list and motif data are given to the program, it is ready for scanning the gene list for motif

occurrences. Users can choose the overall p-value threshold for the motif scanning algorithm (default is

0.001) (**Fig 3C**). Once the p-value is set, the user should click on the "Scan FIMO" button to scan for

motif hits.

Integrating ATAC-seq and ChIP-seq data

If the user has supporting chromatin accessibility and/or in vivo TF binding data for their gene lists, they

can upload the BED files with the respective peaks into VIDEO. This is the fourth window in VIDEO and is

completely optional – click on the "Next" button to proceed to the plot if you don't have this data (**Fig**

3D). VIDEO will use the ATAC and ChIP data as an option to filter the motif occurrences for hits that

occur only in "open chromatin" or hits in "ChIP peak" regions and the filters can be turned on and off

during plotting. For this demonstration, we mined scATAC data from (WANG *et al.* 2025) for the hindgut

tissue in stages 10-12. Once the data are uploaded, the user should press the "Filter Motif Hits" button

and then proceed to the next page.

Adjusting the plot settings

This window is an array of configurable plot settings and a motif occurrence plot below it (**Fig 4**). If the

users chose to upload ATAC/ChIP data they can toggle the filters on and off by checking the respective

options under the "Show only:" heading. The motif occurrence plot will default to the order of the input

gene list and will show all the transcripts of each gene, but the user has the option to customize this.

First, go to the "Sort by motif:" heading and select the motif you want to prioritize. This will display

three options, the first two of which are metrics you need to provide for the sorting. Checking "Number

of Hits" will sort the gene list by the number of hits for the selected motif, and checking "Motif Match

Score" will sort the gene list by the motif match score of the best scoring motif hit for each gene. By

default, all the transcripts of a gene will be clustered together with the top scoring transcript, and the

user has the option to choose "Show only best transcripts for selected motif hit", which will show only

the top ranking TSS for each gene based on the metrics the user selected for ranking.

The last tunable setting is the individual p-value threshold for each motif. Some motifs are more

permissive than others and therefore might need further tuning for match stringency, so users have the

option to make the p-values for each motif independently tighter. For this, click on the "Motif p-value

thresholds" dropdown and drag the slider down the desired p-value. Once all the settings are decided,

click on the "Generate Plot" button and the motif plot will be updated accordingly. All the displayed

motif hits can be downloaded from the button "Download merged hits". For the compare mode, users

can also individually download hits for both genes list as a zip file from the "Download individual hits"

button.

Interpreting the motif occurrence plot

The motif occurrence plot contains all the input genes stacked and aligned along the TSS (**Fig 4**). VIDEO

accounts for transcription direction and reverses the strand so that all the genes have upstream to the

left and downstream to the right of the center line (zero). Each row in the plot corresponds to a gene's

promoter region, and is named in the GeneSymbol_TranscriptID convention, where the transcript ID is

unique to a TSS. The plot is interactive, so users are requested to scroll outside the plot region to go

further down the list. Double clicking on the zoomed-in plot can revert it to the original layout, and

there are also helper buttons to download the image, zoom in, and reset axes at the top right corner of

the plot. Hover over any motif hit to see details about the matched sequence, exact coordinates, strand,

p-value and match score. A darker motif hit corresponds to a stronger motif match score, and a more

transparent hit reflects a weaker match score.

Visualizing ATAC-seq and ChIP-seq tracks along with motif hits

The last window of VIDEO is optional and intended for users who wish to visualize ATAC/ChIP-seq signal

tracks in the context of the motif hits. To use this tool, users must first upload their BigWig signal tracks

of interest (up to three tracks for a plot), give it a name, and pick a color for the track and click Process

Tracks. To view the tracks overlaid with the motif hits, select a gene from "Select a peak" dropdown and

select the tracks to visualize, and then click on "Generate Overlay". This will generate a plot with the

motif hits on top and the signal tracks aligned below it, along the TSS with the strand accounted for

(**Supplemental Fig S1**). The y-axis values are all auto-scaled as a group, and users can visualize where the

peaks in the signal are with respect to the motif hits. This plot is also interactive, and hovering over the

tracks will display the signal values and chromosomal coordinates.

3. Exploring TFs of interest in hindgut tissue

We first sought to use VIDEO explore the transcriptional landscape of the hindgut tissue and identify

candidate TFs that make up the regulatory network of hindgut development. Using VIDEO on scRNA-seq

data from stage 10-12 hindgut cells, we identified *orthopedia* (*otp*)*, brachyenteron* (*byn*)*, and retained*

(*retn*) (aka *dead ringer*) as the top three DE TFs, followed by other hindgut-expressed TFs including *fork*

head (*fkh*)*, luna, drum* (*drm*)*,* and *brother of odd with entrails limited* (*bowl*). (**Fig 2C**). We plotted the

occurrences of the binding motifs of the top three TFs in the top 50 DE genes (three of which are the TFs

themselves) in the hindgut to identify candidate early targets of these TFs.

otp encodes a conserved homeodomain TF required for proper hindgut development (HILDEBRANDT *et al.*

2020), and is strongly expressed in the hindgut as well as in subsets of the CNS (**Supplemental Fig S2**). Its

expression in the hindgut is activated by *byn,* where *otp* regulation depends on the number and

arrangement of multiple high and low-affinity *byn* motifs (KUSCH *et al.* 2002). In the *otp* promoter, VIDEO

identified three *byn* motifs downstream of the TSS, along with one *otp* motif (**Fig 4**).

byn (brachyenteron) encodes a T-box TF expressed in cells that internalize to form the hindgut, where it

acts as a broad regulator of hindgut identity (SINGER *et al.* 1996). In the *byn* promoter, VIDEO revealed

two *otp* motifs adjacent to the TSS and one *byn* motif downstream. While this arrangement suggests

potential autoregulation, and Brachyury orthologs in other species do autoregulate, prior work indicates

that *byn* does not autoregulate in *Drosophila* (SINGER *et al.* 1996). The observed *byn* motif may therefore

represent a conserved but non-functional site, and experimental validation, such as deleting this motif

and assaying *byn* expression would help clarify its role.

The third TF recovered in our analysis was *retn*, which encodes an ARID-domain protein that recognizes

AT-rich DNA sequences. *Retn* is expressed in the hindgut epithelium and other tissues such as the

amnioserosa, ring gland, glia, etc.; within the hindgut epithelium, it is expressed strongly in the

boundary cell rows of the large intestine (IWAKI *et al.* 2001; SHANDALA *et al.* 2002) (**Supplemental Fig**

S2B). While its precise role in hindgut development remains unknown, *retn* is required for early

embryonic patterning (SHANDALA *et al.* 1999) and for CNS development (SHANDALA *et al.* 2003), where it

seems to act in combination with other upstream regulators to modulate gene expression.

Not all motif occurrences reflect true binding events, as some may fall outside accessible chromatin

regions or arise by chance. To identify motif hits that are more likely to correspond to regulatory activity

in the hindgut, we filtered the motif hits using published scATAC data (WANG *et al.* 2025) focusing on

hindgut tissue peaks from stage 10-12 embryos. We broadly restricted peaks to those present in at least

50% of hindgut cells (**Supplemental Fig S3**). Under this cutoff, no motifs were detected in the *retn*

promoter, suggesting that the filter was too stringent. Because *retn* expression is limited to a subset of

hindgut cells, we relaxed the cutoff to include peaks present in at least 25% of the hindgut tissue. This

adjustment restored the expected motif hits in the *retn* promoter, confirming that the restricted

expression of *retn* correlates with localized chromatin accessibility. Notably, motifs for *byn, otp,* and *retn*

were observed in the *retn* promoter, suggesting possible regulatory interactions that merit further

testing in mutant or motif-deletion backgrounds. Further reducing the threshold to 10% did not reveal

any additional motifs.

Although all the genes in the list are the top differentially expressed in the hindgut based on scRNA

annotation (further confirmed by their high expression in the hindgut from in situ hybridization data in

BDGP), we found that the promoter regions of 12 genes lacked detectable open chromatin even at the

lowest threshold of 10%. This finding seems counterintuitive: highly expressed genes are generally

expected to display promoter accessibility, since transcription requires open chromatin. The discrepancy

may stem from the technical limitations of scATAC-seq (e.g., the small number of fragments per cell,

high sparsity of signal and locus-level dropout), which, although widely used to assay open chromatin,

can fail to capture regions where the promoter accessibility is weak, hindered, or transient (CHEN *et al.*

2019; LI *et al.* 2021). That said, we were able to identify candidate regulatory events in 75% of the top

DE genes in the hindgut, providing a set of hypotheses that can be further tested experimentally.

4. Case study: CrebA motif occurrences in SPCGs

To assess whether VIDEO can recover known transcription factor-target gene relationships, we used

CrebA as a test case, asking whether its binding motifs are preferentially enriched and stronger in the

promoters of secretory pathway component genes (SPCGs) compared to ribosomal protein genes

(RpGs). The SPCGs represent a well-established transcriptional target set for the bZIP TF CrebA in the

Drosophila salivary gland and other secretory tissues (ABRAMS AND ANDREW 2005; FOX *et al.* 2010; JOHNSON

et al. 2020; JACKSON *et al.* 2025). CrebA directly regulates genes encoding secretory machinery,

enhancing the secretory capacity of embryonic secretory tissues such as the salivary gland and

epidermis. It has also been implicated in dendrite development through upregulation of CopII

components (IYER *et al.* 2013) and in infection tolerance, where its expression in the fat body is induced

by the Toll and Imd pathways to activate secretory pathway genes (TROHA *et al.* 2018).

For our analysis, we compared 90 SPCGs (JOHNSON *et al.* 2020) to 82 cytoplasmic ribosomal protein

genes (RpGs), which serve as a size-matched gene set that, like the SPCGs, is expected to be found in

open chromatin based on the basal levels of these factors being required in all cells. This analysis also

addresses the question of whether CrebA also increases translation by boosting expression of RPGs, a

regulatory role that has been suggested for the mammalian CrebA orthologue Creb3L2 in the

mammalian pituitary gland. The VIDEO workflow for this experiment is described in **Fig 5**. Motif scanning

revealed that CrebA motifs were substantially more abundant in the proximal promoters of SPCGs than

in RpGs (**Fig 6A**). We tuned the p-value thresholds for both motifs so that the core ACGTG motif was

retained in all hits. This adjustment revealed at least one CrebA motif within 500 bp of the TSS in 84%

(76/90) of SPCGs, whereas only 61% (51/82) of RpGs had a motif hit in the corresponding promoter

proximal regions (**Fig 6B**). Further, among the genes that had at least one motif hit, we examined the

density of motif occurrences within 500 bp of the TSS (**Fig 6C**). In SPCGs, nearly three-quarters of motif-

positive genes carried multiple hits, with 22% containing three and 33% harboring four or more motifs.

By contrast, motif-positive RpGs were skewed toward single occurrences, with almost half (47%)

carrying only one motif and a much smaller fraction (18%) with four or more motifs.

These results indicate that CrebA binding sites are not only more prevalent in SPCGs than RpGs, but also

tend to occur in higher local densities near SPCG promoters. This enrichment of clustered motifs

suggests that SPCG regulation may involve stronger or more cooperative CrebA binding compared to

RpGs. To test if the CrebA motif occurrences showed any particular positional preference, we plotted

the positional distribution of the best hit for each motif in each gene, for both gene lists (**Fig 6D**). The

IUPAC consensus sequence (CrebA_1, derived from Jackson et al., 2025) displayed a sharp enrichment

immediately upstream of the TSS in SPCGs, suggesting that high-confidence CrebA motifs preferentially

cluster at promoter-proximal sites in this gene group, whereas in RpGs they lacked a pronounced

positional peak. The more permissive PWM-based motif (CrebA_2, from the JASPAR 2024 database)

showed relatively flat distributions in both SPCGs and RpGs, consistent with its broader degeneracy and

lower information content. Although these trends qualitatively support the idea that SPCGs may be

subject to more direct promoter-proximal regulation by CrebA, quantification using Kolmogorov-

Smirnov tests did not detect significant positional biases for either motif ($p > 0.05$). Thus, while the

observed peak near the TSS in SPCGs is suggestive, we cannot rule out the possibility that it arises from

stochastic variation rather than actual positional preference.

In addition to motif frequency, motif match quality also differed markedly between the two genes sets.

The log-likelihood ratio scores for the motif matches were significantly higher in SPCGs than RpGs,

indicating stronger consensus matches among SPCGs (**Fig 7A**). Likewise, log-transformed p-values

showed the same trend with SPCGs enriched for significant motif hits compared to RpGs (**Fig 7B**).

Interestingly, analysis of motif positions relative to the transcription start sites did not reveal a

significant difference between SPCGs and RpGs (**Fig 7C**). This finding is consistent with the positional

distribution analysis (Fig. 6D), where CrebA_1 showed a visible promoter-proximal enrichment in SPCGs,

but the effect did not reach statistical significance. Together, these data suggest while SPCGs harbor

more frequent and higher-quality CrebA motifs, their relative positioning within the 500 bp promoter

window is not systematically different from RpGs.

Although the precise positional distribution of motifs remains statistically unresolved, likely due to

limited sample size (fewer than 90 genes), the overall enrichment and strength of CrebA motifs in SPCGs

align with their established regulatory role. These results confirm that VIDEO accurately captures known

biology of CrebA regulation, showing abundance, enrichment and stronger motif matches in SPCGs

compared to control. This case study illustrates how VIDEO integrates motif scanning, visualization, and

tailored data for statistical comparison into a streamlined user-friendly workflow that validates known

transcription factor-gene relationships.

DISCUSSION

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Motif analysis remains a cornerstone for connecting TF binding to regulatory function, but practical

barriers continue to limit its accessibility. Existing workflows typically require multiple command-line

tools, file conversions and custom scripts, making motif analysis challenging for researchers without

strong computational backgrounds. Several visualization tools have emerged, but they often impose

restrictions on input formats, motif databases, or integration with genomic datasets.

VIDEO was developed to address these limitations by offering an integrated, web-based platform

tailored to *Drosophila* data. It is the first of its kind to facilitate an intuitive side-by-side comparison of

user-defined gene lists, incorporates and chooses the best annotated TSSs for each motif based on user-

defined metrics such , and allows for ATAC-seq and ChIP-seq data integration (**Table 1**). It also accepts

custom motifs and gene list/BED file inputs and presents all the plots in a graphical, interactive format,

with downloadable, organized motif hit datasets. These features significantly lower the barrier for motif

exploration and visualization compared to existing tools.

Table 1: Comparison of VIDEO features to other known motif search and visualization tools

	VIDEO	**TFMotifView**	**MEME-Suite**	**HOMER**
Online web server	Yes	Yes	Yes	No (command-line)
Input format	Gene lists, BED regions, typed genes	BED regions	FASTA sequences	Gene lists, BED regions
Compare gene lists	Yes	No	No	No
Genome	fly	Human, mouse	Available locally	Available locally
Motif database	Custom motifs, JASPAR	JASPAR	Several	Custom
View motif occurrences in-app	Yes	Yes	Limited	No
ATAC/ChIP data integration	Yes	No	No	No
Includes all TSS	Yes	No	No	No
Scanning algorithm	FIMO	Motif Information Content	FIMO	HOMER

The CrebA case study highlights VIDEO's ability to recapitulate known biology. SPCGs, a well-

characterized set of CrebA targets, displayed both higher motif frequencies and stronger consensus

matches to CrebA than control RpGs. While the motif positioning relative to the TSS did not differ

significantly between the two sets, this could be insightful about CrebA possibly not having a significant

preference for a certain position for binding to its target genes provided the binding motif lies within the

regulatory window.

Beyond this validation, VIDEO provides a flexible framework that is also suited for hypothesis

generation, with helper functions in TF Finder designed to guide users in exploring the transcriptional

landscape of the early *Drosophila* embryo. The application of VIDEO to stage 10-12 hindgut data

illustrates how visualizing motif occurrences with chromatin accessibility in the right context can be

useful for generating hypotheses about gene regulatory networks. The presence of *otp* motifs adjacent

to the *byn* TSS, and *byn* motifs downstream of *byn* itself, suggests possible autoregulatory or cross-

regulatory interactions. While autoregulation by Brachyury homologs has been demonstrated in other

species, it has been excluded for *Drosophila byn* (SINGER *et al.* 1996), raising the possibility that these

sites are conserved but vestigial or that autoregulation may occur in different developmental contexts.

VIDEO thus highlights candidate motifs that can be directly tested by mutation to clarify their

functionality. Motif filtering by ATAC-seq data revealed the importance of thresholding stringency when

studying genes expressed in rarer cell types, and demonstrates how overly restrictive filters can obscure

possibly relevant motifs in genes in specialized subpopulations.

By combining motif scanning, visualization, and data integration into a single user-friendly tool, VIDEO

enables broader adoption of motif analysis approaches and expands the potential for researchers to link

sequence motifs with regulatory function.

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FIGURE LEGENDS

Figure 1. Flowchart of many possible pipelines using VIDEO. TF Finder and Motif Viewer are the two

pages in VIDEO. Each blue and green box represents a module in the page In Motif Viewer (green), these

modules are presented as navigable cards, and in TF Finder (blue) they are shown one below another in

the same page. Gray boxes represent user-provided data (not native to VIDEO). Yellow boxes represent

final outputs from VIDEO. Arrows represent a sequential order of execution of the modules. Arrows start

from a module that gives an output and go towards a module that can use this output as its input.

Dashed lines represent a hierarchical connection between the pages and their modules.

Figure 2. Steps for Using TF finder. (A) The collapsible sidebar navigator in VIDEO with two functional

pages and a tutorial page. **(B)** Getting DE gene lists from TF Finder. Top: before choosing a

developmental stage and tissue type. Bottom: after making the selections and pressing Get DE genes.

(C) Filtering gene lists for TFs. Top: After pressing Filter TFs, Bottom: Motif details after pressing View

PWM **(D)** Discovering motifs from gene lists. Top: Choose parameters and upload gene list before

running STREME. Bottom: After pressing Run STREME, with link to outputs and a button to download

the discovered motifs.

Figure 3. Steps for Using Motif Viewer. **(A)** Set up genomic input. Left: Before uploading gene list, Right:

After uploading data, choosing window size and pressing Process Genomic Input. **(B)** Set up motif input.

Top: Initial window, Middle: Window after selecting the PWM/PFM format. Paste the motif matrix, press

Parse Pasted Matrix and give the motif a name and color. Bottom: After three motifs are added and

Process Motifs is pressed. **(C)** Scan for motif hits. User sets the p-value threshold and presses Scan

Motifs (FIMO). **(D)** Add ATAC/ChIP-seq data if available. Top: Before uploading data, Bottom: After

uploading ATAC data (Wang et al. 2025) and clicking on the Filter and Score Motif Hits button.

Figure 4. The motif occurrence plot. Motif occurrences of *otp, byn* and *retn* the top 50 DE genes in the

hindgut. Plot settings configured to sort the genes by *otp* hits and match score, and showing only the

best scoring transcript per gene. The configured filtered hits can be downloaded using the Download

hits (CSV) button. Each plot corresponds to the plot settings above it. **(A)** Genes ranked by *otp*

occurrences, scored by number of hits and motif match score. **(B)** Plot further filtered for motif hits only

in open chromatin regions. Note that motifs not in open chromatin disappear, and the genes are sorted

by the remaining motif occurrences.

Figure 5. Workflow of CrebA motif occurrence case study (**A**) In the compare mode, input a gene list of

interest (SPCGs) and a control (RpGs) as gene symbols, choose a window size of 500bp, then press

process genomic input. **(B)** Input the motifs of interest two slightly different forms of the CrebA motif –

the IUPAC consensus sequence from Jackson et al. 2025 (CrebA_1, red) and the PWM from JASPAR 2024

(CrebA_2, yellow) and choose colors for each. **(C)** Adjust plot settings and tune p-value thresholds for

individual motifs and then click on Generate Plot.

Figure 6. CrebA motifs are more abundant in SPCGs than RpGs. (A) Plot shows the occurrence of CrebA

motifs in promoter regions of (left) 90 SPCGs and (right) 82 RpGs. Genes are ordered by the number of

hits and motif match scores of CrebA_1 motifs. **(B)** The majority of secretory pathway component genes

(SPCGs) contain at least one CrebA motif (85%), compared to only 62% of ribosomal protein genes

(RpGs). Fisher's exact test confirms a significant enrichment of CrebA motifs in SPCGs (odds ratio = 3.4, *p*

< 0.001). Motif p-value thresholds are CrebA_1: p<5.25e-4 CrebA_2: p<3.24e-4. **(C)** Of the genes that

have at least 1 CrebA motif hit, a larger fraction of the SPCGs had more than 4 hits as compared to RpGs.

(D) The IUPAC sequence version of the CrebA motif (from Jackson et al., 2025, CrebA_1) shows a peak

near at the TSS for SPCGs but not for the RpGs, although on quantification, the preference is not

significant (KS test p = 0.23). The consensus PWM version of CrebA (from JASPAR 2020, CrebA_2) does

not show obvious positional preference for either group.

Fig 7: CrebA motif matches are stronger and more significant in SPCGs compared to RpGs. (A) Motif

match scores (log-likelihood ratio scores) are significantly higher in SPCGs (median = 10.84) than RpGs

(median = 6.25; p = 1.84e-12). **(B)** Motif match log transformed p-values show the same trend, with

SPCGs (median = 3.82) enriched for more significant motif hits (p = 3.49e-9) as compared to RpGs

(median = 3.58). **(C)** Absolute distance to TSS for all motif hits does not vary significantly between SPCGs

(median = 202.0) and RpGs (194.5, p = 0.36). p-values are for all three analyses are from Mann Whitney

U-test.

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586

Figure 1

Figure 2

Figure 3

A

B

Set up motif input ⓘ

| IUPAC ⌄ | IUPAC sequence |
| Motif name | |

⊕ ■ 🗑

| PWM/PFM ⌄ |

+ Add row

Paste motif matrix here (vertical ACGT), no header

Parse Pasted Matrix

Or upload motif file

Choose File   no file selected

Collapse Editor

| Motif name | ⊕ ■ 🗑 |

| PWM/PFM ⌄ | **Motif Parsed: otp** | Edit Matrix |
| otp | ⊕ ■ 🗑 |

| PWM/PFM ⌄ | **Motif Parsed: byn** | Edit Matrix |
| byn | ⊕ ■ 🗑 |

| PWM/PFM ⌄ | **Motif Parsed: retn** | Edit Matrix |
| retn | ⊕ ■ 🗑 |

+ Add Motif

Process Motifs Processed!

C

Scan for motif hits ⓘ

FIMO Threshold: 0.001 ⌄

🔍 Scan Motifs (FIMO)

FIMO: Grant et al. *Bioinformatics* 2011

Back   Next

D

Optional: Add chromatin accessibility/binding data ⓘ

BED file with open chromatin
regions (ATAC-seq peaks):   Choose File   no file selected

BED file with binding peak
regions (ChIP-seq peaks):   Choose File   no file selected

✏️ Filter & Score Motif Hits

Back   Next

Optional: Add chromatin accessibility/binding data ⓘ

BED file with open chromatin
regions (ATAC-seq peaks):   Choose File   hindgut50pct.bed

BED file with binding peak regions
(ChIP-seq peaks):   Choose File   no file selected

Filtered!

Motif hits have been filtered and scored!

Back   Next

Figure 4

Figure 5

Figure 6 (A)

Figure 6 (continued)

B

Proportion of genes with ≥1 CrebA motif hit

Fisher's exact p = 0.00095

C

Number motif hits per gene

D

Positional Distribution of Best Motif Hits

Figure 7

A

Motif match scores (Score_bits) by dataset

B

-log10(p-value) by dataset

C

Absolute distance to TSS by dataset