Lineage-specific evolution of regulatory landscapes in polyploid plant and its diploid progenitors

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Abstract
Cis-regulatory elements are specific DNA sequences that control gene expression in a spatiotemporal manner, and variation within these elements represents a major source of phenotypic diversity and evolutionary innovation. Nevertheless, how regulatory elements evolve and shape gene expression remains poorly understood, particularly in plants. The well-resolved phylogeny of allopolyploid peanut (*Arachis hypogaea*) and its diploid progenitors, *A. duranensis* and *A. ipaensis*, provides an ideal system to investigate the regulatory evolution at a lineage-specific level. By integrating comparative analyses of sequence similarity, chromatin accessibility, histone modifications, conserved noncoding sequences (CNSs), and gene expression, we reconstructed the evolutionary trajectories of Accessible Chromatin Regions (ACRs), where regulatory elements typically reside, and revealed their distinct contributions to homeolog expression bias, unequal expressions between homeologs, in a polyploid genome. Most ACRs exhibited high sequence similarity, comparable chromatin accessibility, and conserved states for H3K4me3, H3K56ac, and H3K36me3, indicating regulatory stability after hybridization and polyploidization. However, a subset of novel ACRs emerged *de novo* or through sequence mutations in preexisting ACRs, arising at different rates and evolutionary stages. Notably, even highly sequence-conserved ACRs exhibited substantial variation in chromatin accessibility variation, consistent with CNS composition differences and minor sequence variation, although causal relationships remain to be demonstrated. Our analyses further revealed a complex spectrum of CNS dynamics across the diploid-polyploid framework. Overall, our study provides empirical insights into the fine-scale evolution of plant regulatory landscapes and complements previous large-scale comparisons across distant lineages.

Introduction
Cis-regulatory elements, specific DNA sequences bound by transcription factors, regulate gene expression in a spatiotemporal manner (1–4). Genetic variation within these elements represents an important source of phenotypic diversity and evolutionary innovation (5, 6). Nevertheless, the origin and the extent of sequence divergence within these elements and how such divergence shapes regulatory activity and gene expression remain poorly understood, especially in plants.

Unlike the relatively conserved and stable mammalian genomes, plant genomes are highly dynamic, exhibiting extensive variation in size, structure, and compaction (7, 8). Frequent macro- and micro- genomic rearrangements, whole genome duplication, gene duplication and unequal gene retention have all shaped plant genome architecture. These processes likely exert a profound impact on the emergence and diversification of regulatory sequences (9–11). Comparative analyses across distantly related plants have revealed a high rate of turnover in regulatory regions, underscoring the fast-evolving nature of plant *cis*-regulatory landscapes and

the difficulty of tracing their evolutionary trajectories (12–14). Yet, deeply conserved non-coding
regions (CNSs) spanning more than 280 plant species and 300 million years of diversification
have also been identified (10). To reconcile this seemingly contradictory observation between,
rapid turnover versus deep conservation, and to complement broad-scale comparative studies,
analyses among closely related species provides a critical and trackable framework to uncover
how cis-regulatory sequences diversify, acquire new functions, and drive regulatory innovation
over short evolutionary timescales.
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Polyploid plants that have undergone whole-genome duplication provide a natural context to
contrast the evolutionary fates of duplicated regulatory elements. During polyploidization,
regulatory elements are copied alongside genes and may experience relaxed selective
constraints like those acting on duplicated coding sequences (15), thereby enabling divergence
in both sequence and function (5). Recent studies in soybean, cotton, and strawberry have
illustrated that variation in cis-regulatory regions contributes to differential expression among
duplicated genes (16–22). However, most insights come from species with ancient duplication
events or from analyses lacking direct comparisons with their progenitors. Incorporating
progenitor species with polyploid species provides a unique opportunity to reconstruct how
regulatory sequences originate and diversify.
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This framework is particularly well suited for recently formed allopolyploids such as cultivated
peanut (Arachis hypogaea), an allotetraploid crop (AABB, 2n = 4x =40) derived from
hybridization between two diploid progenitors, A. duranensis (progenitor A) and A. ipaensis
(progenitor B), followed by whole-genome duplication (23–29). Two diploid progenitors
diverged ~2 million years ago (Mya), whereas divergence between each A. hypogaea
subgenome and its corresponding progenitor occurred ~0.2 Mya (28). This short evolutionary
timespan has resulted in two subgenomes being clearly distinguishable (25), and likely
insufficient time for extensive cis-regulatory divergence between derived and ancestral states.
Peanut, therefore, represents an exceptional model to examine the early stages of cis-regulatory
evolution and to trace how regulatory elements evolve through genome hybridization and
duplication.
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To investigate sequence variation in cis-regulatory elements and its relationship to regulatory
activity and gene expression divergence, we analyzed three peanut species: A. hypogaea, A.
duranensis and A. ipaensis, using both leaf and root tissues. We integrated Assay for
Transposase Accessible Chromatin Sequencing (ATAC-seq) to map genome-wide chromatin
accessibility, Chromatin Immunoprecipitation Sequencing (ChIP-seq) to profile the histone
modifications, and RNA-seq to quantify transcript abundance. We identified Accessible
Chromatin Regions (ACRs) and classified them into distinct evolutionary categories based on
sequence conservation. Comparative analyses of sequence variation, chromatin accessibility
states, histone modification patterns and CNSs dynamics, revealed distinct evolutionary and
functional signatures among these classes. Together, these results illustrate the dynamic
regulatory landscapes of recently formed polyploids, where ACRs of different evolutionary
origins follow diverse trajectories that collectively shape regulatory balance and expression
diversification within the polyploid genome.

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Results
Mapping the accessible chromatin landscape in peanut
We generated 12 ATAC-seq libraries, comprising two biological replicates of leaf and root tissues
from each of the three peanut species: *A. hypogaea*, *A. duranensis* and *A. ipaensis*. An average
of 84,513,689 raw reads were obtained per library. After adapter trimming and quality filtering,
uniquely mapped reads were used for peak calling, and regions meeting defined thresholds
were designated as ACRs (see Methods). On average 61,539 peaks were identified per replicate
(**Table S1**). The average proportion of reads around Transcription Start Site (TSS) and average
Fraction of Reads in Peak (FRiP) score were 0.35 and 0.48, indicating the good quality of ATAC-
seq data (**Table S1**).
99
Comparative characterization of ACRs across four peanut genomes
We merged ACRs identified from leaf and root tissues for each genome: subgenome A and
subgenome B of *A. hypogaea* and the two diploid progenitors, *A. duranensis* and *A. ipaensis*,
yielding 56,526, 56,847, 113,025, and 95,584 ACRs, respectively (Table S2). These merged ACRs
were used as references for differential chromatin accessibility analysis between leaf and root
tissues. Regions showing no significant difference in chromatin accessibility were classified as
broad ACRs, whereas those with significantly higher chromatin accessibility in either tissue were
designated as leaf-specific or root-specific (**Fig. 1A**).
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Broad ACRs accounted for the majority of ACRs, with tissue-specific ACRs representing a smaller
fraction (**Fig. 1B, Fig. S1** and **Table S2**), consistent with findings from previous studies (13, 14,
30–32). Based on proximity to genes, ACRs were further categorized as genic, proximal (≤2 kb
from genes), or distal (>2 kb). Among tissue-specific ACRs, distal regions were most prevalent,
whereas broad ACRs were enriched in genic and proximal regions (**Fig. 1C, Fig. S1** and **Table S2**).
Although a greater number of ACRs were identified in diploid genomes, both diploids and two
subgenomes exhibited comparable ACR patterns of tissue specificity and genomic distribution
(**Fig. 1B-1C, Fig. S1** and **Table S2**).
117
Identification of ACRs with distinct evolutionary trajectories
The well-resolved phylogenetic relationships between the two subgenomes and their respective
diploid progenitors provide a rigorous framework for tracing the evolutionary trajectories of *cis*-
regulatory elements (**Fig. 2A**). To infer the sequence ancestry of ACRs in polyploid peanut, we
performed BLAST searches of ACRs from each subgenome against the other subgenome and
both diploid progenitor genomes. To preserve the local genomic context of each regulatory
region, we implemented a synteny-based BLAST pipeline adapted from (13). In this framework,
each query ACR was anchored by its flanking homoeologous genes, and BLAST searches were
restricted to the corresponding syntenic intervals in the target genomes (**Fig. S2**). We focused on
ACRs located within syntenic intervals conserved across all four genomes: the two subgenomes
and two diploid genomes. Although restricting analyses to conserved syntenic intervals may
miss non-collinear homologous ACRs and transposable elements (TEs)-mediated relocations (33,
34), it provides a robust baseline for tracking regulatory evolution. Based on sequence
conservation profiles, ACRs were classified into five evolutionary categories reflecting their
possible origins: 1) **subgenome-specific**, present only in one subgenome and absent from the
other subgenome and both diploid genomes, suggesting *de novo* emergence; 2) **polyploid-**

specific, shared between subgenomes but not detected in diploids, indicating formation after
hybridization but prior to genome duplication; 3) **lineage-specific**, conserved only between
subgenome and its corresponding diploid progenitor, representing inheritance from the
ancestral genome; 4) **peanut-conserved**, retained across four genomes, indicating long-term
retention throughout the peanut lineage; 5) **potential peanut-conserved**, detected in three
genomes or shared between a subgenome and the non-corresponding diploid genome (**Fig. 2A**).
140
To evaluate the stability of these classifications, we applied two sequence similarity thresholds
(≥50% and ≥80%). Raising the threshold from 50% to 80% led to a marked reclassification of
ACRs, particularly with many shifting from peanut-conserved to lineage-specific categories (**Fig.**
2B, Fig. S3 and **Table S3**). This shift indicates that sequence divergence has accumulated within
regions previously considered conserved, reflecting their recent evolutionary trajectories.
Because the higher threshold more accurately captures these recent differences, subsequent
analyses were based on ACR classifications defined by the 80% criterion. In both subgenomes,
most ACRs within conserved syntenic intervals were peanut-conserved (68%) (**Fig. 2B, Fig. S3**
and **Table S3**). When including potential peanut-conserved ACRs, this proportion increased to
approximately 80%, underscoring their overall high sequence conservation. Lineage-specific
accounted for ~12%, followed by subgenome-specific (~2-4%) and polyploid-specific ACRs (~2-
4%) (**Fig. 2B, Fig. S3** and **Table S3**). Notably, subgenome-specific ACRs were more abundant in
subgenome A (4%) than in subgenome B (2%) (**Fig. S3** and **Table S3**).
154
155 Tissue specificity patterns were largely consistent across categories, with most ACRs exhibiting
156 broad chromatin accessibility (**Fig. S4** and **Table S4**). Genomic distributions were also
157 comparable, although peanut-conserved ACRs, potential peanut-conserved ACRs and polyploid-
158 specific ACRs had slightly higher proportions of genic ACRs (**Fig. S4** and **Table S4**).
159
TEs disproportionately shape ACRs with different evolutionary trajectories
Repeat sequences are abundant in the peanut genome, with retroelements and DNA
transposons comprising the majority (29). TEs are well known to drive the emergence of novel
cis-regulatory elements by introducing regulatory sequences into new genomic contexts (5, 12,
33–37). In our analysis, TE associations were more frequent among subgenome-specific ACRs
(**Fig. 2C, Fig. S5** and **Table S4**). For instance, 85.8% of subgenome-specific ACRs in subgenome B
overlapped with repeat elements by at least 20 bp, and 20.7 % were entirely embedded within
repeats (**Fig. 2C, Fig. S5** and **Table S4**), underscoring the strong contribution of TEs to regulatory
novelty. Moreover, TE-associated ACRs were consistently more abundant in subgenome B than
in subgenome A (**Fig. S5** and **Table S4**), suggesting that TE insertions may have exerted
asymmetric influences on the evolution of regulatory landscapes between the two subgenomes.
171
Genetic variation patterns among ACRs with distinct evolutionary histories
To evaluate sequence variation among ACRs with different evolutionary trajectories, we
quantified two metrics: the BLASTed ratio, defined as the length of the query ACR divided by the
length of its BLASTed sequence in the corresponding reference region, and the mismatch rate,
defined as the number of mismatches per BLASTed sequence. Subgenome-specific ACRs were
excluded from this analysis because no corresponding sequences exist in the reference regions.
Across the remaining categories, most ACRs exhibited high BLASTed ratios (~1) and low
mismatch rates (~0.01-0.02) (**Fig. 2D, Fig. S6** and **Table S4**). These values contrast sharply with

those reported for conserved ACRs in soybean, where the median BLASTed ratio of recently
duplicated ACRs was ~0.75 and the median mismatch rates was ~0.05 (22). This comparison
underscores the elevated regulatory sequence conservation in this recently formed polyploids
and highlights the influence of divergence time on regulatory sequence decay. Interestingly,
lineage-specific ACRs exhibited the highest BLASTed ratios and lowest mismatch rates, whereas
polyploid-specific ACRs displayed the lowest BLASTed ratios and highest mismatch rates (**Fig.**
2D, Fig. S6 and **Table S4**). This trend is somewhat counterintuitive, as lineage-specific ACRs likely
emerged earlier and might therefore be expected to accumulate more divergence. Their
elevated conservation may instead reflect stronger selective pressure associated with their
single-copy status in the polyploid genome. In contrast, the duplicated nature of peanut-
conserved and polyploid-specific ACRs may lead to relaxed selective constraints, facilitating the
accumulation of mutations.
192
Histone modification dynamics of sequence-conserved ACRs
To characterize the histone modification landscape associated with ACRs, we performed ChIP-
seq for three peanut species in both leaf and root tissues. Chromatin-state conservation was
inferred from three active histone marks: H3K4me3, H3K56ac, and H3K36me3. We did not
profile repressive histone marks or DNA methylation, which may exhibit distinct regulatory
dynamics (38, 39). Genome-wide patterns of histone marks were consistent with previous
reports (12, 40–43), with H3K4me3 and H3K56ac enriched around transcription start sites, and
H3K36me3 enriched within gene bodies (**Fig. S7**). We then intersected ACR coordinates with
peaks of individual histone marks and classified ACRs into three states: transcribed, Kac-
enriched, and unmodified (**Fig. 2E**), based on the combinatorial patterns of the three marks,
following the clustering criteria described in (12).
Comparison of histone states across peanut-conserved ACRs and their corresponding BLASTed
regions revealed that most ACRs (~85%) retained identical histone states between the two
subgenomes, and the majority also shared the same states as both diploid progenitors (~60%)
(**Fig. 2E** and **Table S5**). A small subset (~6%) displayed conserved states within subgenomes and
within progenitors but differed between subgenomes and progenitors (**Fig. 2E** and **Table S5**).
The remaining ~15% exhibited divergent histone states between subgenomes; in most of these
cases, one or both subgenomes matched the histone state of the corresponding progenitor,
highlighting strong parental-legacy effects, whereas only ~2% showed entirely novel histone
states absent from either progenitor (**Fig. 2E** and **Table S5**). Lineage-specific and polyploid-
specific ACRs also displayed high histone state similarity, ranging from 70% to 86%, either
between a subgenome and its corresponding progenitor or between the two subgenomes
(**Table S5**).
216
Varied chromatin accessibility bias among sequence-conserved ACRs
To further investigate the regulatory dynamics of sequence-conserved ACRs, we quantified
chromatin accessibility by counting reads overlapping each conserved ACR and its corresponding
BLASTed sequences. Peanut-conserved ACRs, those with highly conserved sequences across all
four genomes, provide a framework for examining chromatin accessibility dynamics in both
ancestral and derived states. Principal component analysis (PCA) of the normalized chromatin
accessibility counts revealed a strong parental-legacy effect on chromatin accessibility variation.
Principal Component 1 (PC1), explaining ~36.3% of the total variance, separated the two

lineages, subgenome A clustered with *A. duranensis*, and subgenome B with *A. ipaensis*. PC2,
accounting for ~27.5% of the variance, distinguished tissue types (**Fig. 3A** and **Fig. S8**).
Following the conceptual framework of homoeolog expression bias, in which homoeologous
genes exhibit differential expression (44–48), we examined chromatin accessibility bias in
peanut-conserved ACRs to determine whether chromatin accessibility differed between
subgenomes, whether the bias favored subgenome A or subgenome B, and whether such bias
reflected parental legacy or regulatory novelties that emerged following genome hybridization
and duplication. Accordingly, each ACR was classified into three major groups: (i) **parental**
condition, no-bias across all four genomes or a bias consistent with the progenitors; (ii) **no bias**
in progeny, a bias present in progenitors but absent in the polyploid; and (iii) **novel bias in**
progeny, bias distinct from the parental condition (**Fig. 3B**). Within each category, ACRs were
further designated as subgenome A- or subgenome B- dominant (or progenitor A- or progenitor
B- dominant) (**Fig. 3B**), depending on which genome exhibited significantly higher chromatin
accessibility (See Methods).
239
Distributions of ACRs with different chromatin accessibility bias patterns were largely similar
between leaf and root tissues, except that a greater proportion exhibited a parental no bias
pattern in root tissue (**Fig. 3C** and **Table S6**). Across tissues, most ACRs (~70–80%) showed no
bias between the two subgenomes, with the majority (~45–63%) also lacking bias between the
diploid progenitors (parental no bias), and the remainder (~13–23%) showing bias between the
progenitors (no bias progenitor A- or progenitor B- dominant) (**Fig. 3C** and **Table S6**).
Approximately 20% of ACRs exhibited significant accessibility bias between subgenomes,
primarily reflecting parental-legacy effects (parental subgenome A- or subgenome B- dominant
~10%), followed by novel bias (novel bias subgenome A- or subgenome B- dominant ~6%) and
novel reverse bias (novel bias-r subgenome A- or subgenome B- dominant ~2%), in which the
direction of bias was opposite to that observed in the progenitors (**Fig. 3C** and **Table S6**).
251
Subgenome A- or subgenome B- dominant ACRs, regardless of whether they were classified as
parental, novel or novel reverse, were largely restricted to their respective subgenomes and
were typically not detected as ACRs in the other subgenome (**Fig. 3C** and **Table S6**). The
prevalence of these presence/absence differences in chromatin accessibility suggests that this
may represent a major form of regulatory asymmetry among sequence-conserved ACRs
between subgenomes. To investigate this, we overlapped the BLASTed coordinates of ACRs
identified from one subgenome (e.g., subgenome B) in the other subgenome (e.g., subgenome
A) with the coordinates of experimentally identified ACRs in the other subgenome (e.g.,
subgenome A). Subgenome A- or subgenome B- dominant ACRs exhibited markedly lower
overlap, whereas ACRs with other chromatin accessibility bias patterns showed much higher
overlap (**Fig. 3D, Fig. S9,** and **Table S7**). Furthermore, normalized chromatin accessibility counts
for these dominant ACRs were substantially higher in their own subgenome compared to their
BLASTed counterparts in the other subgenome (**Table S8**), reinforcing that differential
chromatin accessibility between subgenomes often manifests as presence-absence variation
rather than as quantitative shifts.
267
268   We next examined sequence variation among ACRs exhibiting different chromatin accessibility
269   bias patterns. Although no strong overall pattern was observed, parental no bias ACRs
270   consistently had the lowest mismatch rates and relatively high BLASTed ratio compared to ACRs

271 with altered chromatin accessibility (**Table S9**), suggesting that sequence divergence contributes
272 to chromatin accessibility variation.
273
274 Comparisons of lineage-specific ACRs between each subgenome and its corresponding diploid
275 progenitor, as well as of polyploid-specific ACRs between subgenomes, also revealed that most
276 did not exhibit significant differences in chromatin accessibility (**Table S10**), consistent with
277 previous observations of histone modification states (**Fig. 2E** and **Table S5**).
278
Age-dependent conservation and loss of CNSs in peanut subgenomes and progenitors
Cis-regulatory elements are often highly conserved, reflecting selection to maintain essential
regulatory function (10, 49–53). To gain further insights into their evolutionary dynamics, we
examined CNSs in the context of polyploid peanut and its corresponding diploid progenitors,
providing complementary perspective to studies that have largely overlooked lineage-specific
regulatory evolution (10, 36, 54, 55).
285
Consistent with previous findings (10), ancient CNSs, such as those conserved at the
Tracheophyte (vascular plants) or Angiosperm level, were less numerous but exhibited higher
copy numbers and greater retention than lineage-specific CNSs, such as family-level CNSs. For
example, 71% of CNSs were conserved at the Fabaceae level, whereas only ~2% were conserved
at the Angiosperm level (**Fig. 4A** and **Table S11**); ~75% of Tracheophyte-level CNSs had at least
two copies, compared with 25% of Fabaceae-level CNSs (**Fig. 4B** and **Table S12**); and 55% of
Tracheophyte-level CNSs were retained across all three peanut species, versus 23.9% of
Fabaceae-level CNSs (**Fig. 4C** and **Table S13**). In contrast to the previously proposed view that
one paralog largely retains ancestral CNSs while the other undergoes rapid turnover (10), we
found that ~42% of Tracheophyte- and Angiosperm-level CNSs were preserved across both
subgenomes and their diploid progenitors (**Fig. 4C** and **Table S13**). This pattern may partly
reflect the relatively recent origin of the allotetraploid, leaving insufficient time for ancient CNSs
to diverge.
299
Beyond these general trends, we observed a more complex spectrum of CNS dynamics across
the diploid-polyploid framework. CNSs could be categorized as: (i) conserved across all four
genomes, (ii) conserved across three genomes, (iii) conserved exclusively in progenitors, (iv)
conserved exclusively in polyploid subgenomes, (v) conserved between a subgenome and its
corresponding progenitor, (vi) conserved between a subgenome and the non-corresponding
progenitor, and (vii) restricted to a single genome (**Fig. 4D-4E**).
306
Notably, the distribution of these patterns varied strongly with CNS age. Ancient CNSs were
dominated by those conserved across all four genomes, followed by those retained only in
progenitors (**Fig. 4D** and **Fig. S10**). In contrast, lineage-specific CNSs were most frequently
conserved exclusively in progenitors, with the second most category being those conserved
across all four genomes (**Fig. 4E** and **Fig. S10**). The next predominant groups within lineage-
specific CNSs included CNSs restricted to progenitor A or progenitor B, as well as CNSs absent in
one subgenome but present in the other three genomes (**Fig. 4E** and **Fig. S10**). These categories
were also among the next common for ancient CNS (**Fig. 4D** and **Fig. S10**); however, progenitor-
restricted CNSs were substantially more frequent than subgenome-missing CNSs among lineage-
specific CNSs (**Fig. 4D** and **Fig. S10**), whereas this disparity was minimal for ancient CNSs (**Fig. 4E**

and **Fig. S10**). These patterns indicate that CNSs of different evolutionary ages follow distinct
trajectories. Ancient CNSs tend to be stably retained, whereas lineage-specific CNSs exhibit
substantial turnover. On the other hand, the large number of lineage-specific CNSs present in
one or both progenitors but absent from the polyploid (**Fig. 4E** and **Fig. S10**), suggests
widespread loss of these elements after hybridization and polyploidization. This may reflect the
inherent dynamism of *cis*-regulatory elements: some may rapidly evolve into new regulatory
sequences and diverge enough to escape detection as a lineage-specific CNSs. Alternatively,
hybridization, whole-genome duplication, and subsequent regulatory reorganization may drive
the extensive loss of lineage-specific CNSs. Further work will be needed to disentangle these
possibilities. We also observed many CNSs conserved exclusively between each subgenome and
its corresponding progenitors, particularly for lineage-specific CNSs (**Fig. 4D-4E** and **Fig. S10**),
further underscoring the strong influence of parental legacy on regulatory architecture. Other
categories occurred at much lower frequencies (**Fig. 4D-4E** and **Fig. S10**).
330
Impacts of the CNS dynamics on chromatin accessibility variation of sequence-conserved ACRs
To explore whether CNS turnover correlates with the chromatin accessibility variation observed
in sequence-conserved ACRs, we examined CNSs dynamics within these regions using a
comparative framework. We applied the same presence/absence–based CNS classification
described above (**Fig. 4D-4E** and **Fig. S10**), but in a more targeted context: each CNS was
categorized based on its association with genes in the same orthogroup across the four
genomes and its location within a sequence-conserved ACR and/or its corresponding BLASTed
regions (see Methods) (**Table S14**). We then summarized the distribution of CNS conservation
states for each ACR using a rank-normalized approach that accounts for both the CNS
conservation state and its abundance (see Methods).
341
The patterns were consistent between leaf and root tissues (**Fig. 5** and **Fig. S11**). ACRs with
parental subgenome-dominant chromatin accessibility were frequently associated with CNSs
conserved exclusively between that subgenome and its diploid progenitor (e.g., P_B-S_B in **Fig. 5**).
ACRs with parental no bias chromatin accessibility tended to contain CNSs conserved across all
four genomes (e.g., P_A-P_B-S_A-S_B in **Fig. 5**). ACRs exhibiting novel bias subgenome-dominant
chromatin accessibility were more likely to harbor CNSs unique to that subgenome (e.g., S_B in
Fig. 5) or CNSs missing in the other genomes (e.g., P_A-P_B-S_B in **Fig. 5**). In contrast, CNSs conserved
across three genomes were often not enriched in any chromatin accessibility category (e.g., P_B-
S_A-S_B, P_A-S_A-S_B, P_A-P_B-S_A in **Fig. 5**). CNSs conserved exclusively in progenitor A (e.g., P_A in **Fig. 5**) or
progenitor B (e.g., P_B in **Fig. 5**) were unexpectedly common across nearly all accessibility groups,
despite they being lost in the subgenomes. CNS exclusively conserved in both progenitors (e.g.,
P_A-P_B in **Fig. 5**) showed a distinctive pattern: they were broadly enriched but specifically
depleted in parental subgenome-dominant chromatin accessibility, suggesting that these ACRs
carry regulatory features that differ markedly between the two progenitor lineages. Together,
these results suggest that CNSs with different conservation states contribute meaningfully to
chromatin accessibility patterns of sequence-conserved ACRs in polyploid genome, shaping their
regulatory activities.
359
We also analyzed lineage- and polyploid-specific ACRs using simplified CNS classifications. For
lineage-specific ACRs, CNSs were categorized as conserved exclusively in the subgenome, in the
corresponding progenitor, or in both. The distributions of the CNS with different conservation

categories did not differ between lineage-specific ACRs with the same versus different
chromatin accessibility states (**Table S15** and **Fig. S12**). For polyploid-specific ACRs, CNSs were
categorized as conserved exclusively in one subgenome or in both. Among polyploid-specific
ACRs with differential accessibility, CNSs conserved in both subgenomes were consistently
depleted (**Table S16** and **Fig. S13**).
368
Homoeolog expression bias in peanut and its underlying potential mechanisms
To characterize homoeolog expression bias in peanut, we performed RNA-seq on leaf and root
tissues from the same three peanut species, each with two biological replicates. To enable cross-
genome comparisons, we focused on homoeologous genes identified across the two
subgenomes and their corresponding diploid progenitors. PCA of normalized gene counts
revealed that PC1 (44.3% of total variance) separated the diploid progenitors from the
subgenomes, PC2 (26%) further distinguished the two subgenomes, and PC3 (19.4%)
differentiated the two tissue types (**Fig. 6A**). Homoeologous genes displayed pronounced
expression divergence both between subgenomes and relative to their diploid counterparts,
whereas the two diploid progenitors showed highly similar expression profiles for these genes
(**Fig. 6A**).
380
Using the same conceptual framework applied to classify chromatin accessibility patterns we
categorized homoeolog expression bias (**Fig. 3B**). The overall distributions of these categories
were highly consistent between leaf and root tissues **(Table S17** and **Fig. 6B**). Most genes
exhibited no expression bias between the two subgenomes, with the majority showing a
parental no bias pattern (**Table S17** and **Fig. 6B**), consistent with previous findings (44). Among
genes exhibiting bias between the two subgenomes, novel bias accounted for the largest
proportion (~20%), followed by parental-legacy effects (~6%), and novel bias reverse (~3%)
(**Table S17** and **Fig. 6B**). These patterns suggest that most expression asymmetries might have
arisen *de novo* following hybridization and polyploidization, rather than being inherited from the
ancestral diploid progenitors.
391
To investigate how ACRs with different evolutionary trajectories relate to homoeolog expression
bias, we performed enrichment analyses linking expression categories with ACR evolutionary
types. Homoeolog expression patterns were simplified into four classes: no-bias, novel-bias,
parental-legacy bias and others. ACRs were first grouped by sequence conservation as peanut-
conserved, potential peanut-conserved, lineage-specific, polyploid-specific and subgenome-
specific and peanut-conserved ACRs were further subdivided by chromatin accessibility bias into
no-bias, novel-bias, parental-legacy bias and others. To ensure robust evolutionary inference,
we restricted the analysis to homoeologous genes associated with ACRs that could be
confidently classified based on BLAST conservation, excluding those had not been BLASTed due
to their absence from conserved syntenic regions. This integrative framework allowed us to
systematically examine whether specific classes of regulatory elements are preferentially
associated with distinct modes of homoeolog expression divergence.
404
To minimize the uncertainty in assigning ACRs to their target genes, we performed an additional
enrichment analysis excluding distal ACRs. Both the full and filtered datasets showed
comparable enrichment patterns, and the overall trends were consistent between leaf and root
tissues (**Table S18** and **Fig. S12**). Although few categories reached statistical significance, the

direction and magnitude of Z-scores offered meaningful trends (**Table S18** and **Fig. S12**). Given
current power and multiple testing restrictions, we interpret these associations as trend-level
and hypothesis-generating, pending expanded sampling and covariate-aware models. Genes
showing no-bias expression were positively associated with peanut-conserved ACRs exhibiting
no-bias in chromatin accessibility, whereas biased expression categories tended to show
negative enrichment, particularly parental-legacy bias expression (**Fig. 6C** and **Table S18**). In
contrast, peanut-conserved ACRs with novel-bias chromatin accessibility were positively
enriched for genes with novel-bias expression but negatively or weakly enriched for no-bias and
parental-legacy bias expression (**Fig.6C** and **Table S18**). Subgenome-specific ACRs exhibited
similar tendencies, displaying negative enrichment with no-bias expression but positive
enrichment with novel-bias and parental-legacy bias expression (**Fig. 6C** and **Table S18**). Lineage-
specific ACRs and peanut-conserved ACRs with parental-legacy bias were generally positively
enriched across all expression categories, with lineage-specific ACRs showing the strongest and
statistically significant association with parental-legacy bias expression (**Fig. 6C** and **Table S18**).
Conversely, polyploid-specific ACRs displayed consistently negative enrichment across all
expression classes, particularly for no-bias expression (**Fig. 6C** and **Table S18**).
425
Together, these results indicate that genes with stable, unbiased expression tend to associate
with conserved regulatory elements maintaining no-bias chromatin accessibility, whereas genes
with biased expression shows enrichment for more recently evolved or divergent ACRs, such as
lineage-specific ACRs, subgenome-specific ACRs or peanut-conserved ACRs with novel-bias
chromatin accessibility. Meanwhile, the correspondence between ACR evolutionary origin and
expression biases, where parental-legacy bias expression associates with lineage-specific ACRs
and novel bias expression aligns with subgenome-specific ACRs, reveals a coordinated
regulatory evolution, highlighting how ACRs with different evolutionary histories contribute to
expression asymmetry in polyploid genome.
435
Discussion
Genetic variation within *cis*-regulatory elements is a major source of phenotypic diversity and
evolutionary innovation. Understanding its evolutionary dynamics is therefore essential for
elucidating the genetic basis of natural and artificial selection, and for facilitating targeted
manipulation of gene regulation (6, 10, 56, 57). Although regulatory sequences exhibit rapid
turnover across distantly related plant lineages (12), deeply conserved noncoding elements have
also persisted across species diverged over 300 million years (10). However, how *cis*-regulatory
elements evolve among closely related species and on lineage-specific time scale remains poorly
understood (10, 12–14). Here we address this gap by systematically examining the evolutionary
trajectories of regulatory elements in allotetraploid peanut and its diploid progenitors (**Fig. 2A**).
446
Beyond associating genetic variation with expression divergence in polyploids (16–19, 22, 58),
we further investigated the evolutionary origins of regulatory variation, offering an additional
dimension to understand *cis*-regulatory dynamics underlying expression changes. We uncovered
a strong correspondence between homoeolog expression bias and ACR variation at both
sequence and accessibility levels, reflecting their distinct evolutionary histories. For example,
parental-legacy bias expression was enriched for lineage-specific ACRs, whereas novel-bias
expression tends to associate with subgenome-specific or peanut-conserved ACRs with novel-
bias chromatin accessibility (**Fig. 6C** and **Table S18**). In contrast, peanut-conserved ACRs with no

bias chromatin accessibility were depleted for biased expression but enriched for unbiased
expression (**Fig. 6C** and **Table S18**). We also detected a pronounced parental-legacy effect on
regulatory evolution (17, 45, 59–61), supported by lineage-specific ACRs (**Fig. 2B**), peanut-
conserved ACRs exhibiting parental-legacy bias chromatin accessibility (**Fig. 3C**) and
conservation of CNSs between a subgenome and its corresponding progenitor. (**Fig. 4D-4E**).
However, how the strength of this parental-legacy effect shifts over time remains an important
open question. Collectively, these findings highlight that ACSs with distinct evolutionary
trajectories exert differential impacts on regulatory landscapes in the polyploid genome.
463
From a broader evolutionary perspective, we characterized the fates of regulatory regions. Most
ACRs were sequence-conserved (**Fig. 2B**), exhibiting exceptionally high sequence similarity
(mean BLASTed ratio ~0.98 and mean mismatch rate ~0.02) (**Fig. 2D**), despite the ~2 Mya
divergence between the diploid progenitors (28). Many of these sequence-conserved ACRs also
maintained comparable histone modification profiles (**Fig. 2E**) and chromatin accessibility (**Fig.
3C**), likely reflecting both the recent formation of the polyploid and strong selective constrains
on regulatory function (10, 49, 50). Meanwhile, we also observed the emergence of new ACRs
that originated either *de novo* or through accumulated mutations in preexisting regulatory
sequences (6, 9–11). Polyploid-specific or subgenome-specific ACRs exemplify *de novo* origins,
arising after hybridization or after polyploidization (**Fig. 2A**). In contrast, some lineage-specific
ACRs likely represent gradual sequence divergence: ACRs that shift from peanut-conserved to
lineage-specific when the BLAST similarity threshold increases from 0.5 to 0.8 (**Fig. 2B**) illustrate
this process: these regions likely existed prior to the divergence of the diploid progenitors and
subsequently accumulated mutations, increasing intra-lineage similarity to ~80% while reducing
inter-lineage similarity to ~50%. Peanut-conserved ACRs with high sequence similarity but
biased chromatin accessibility provide another example of diversification (**Fig. 5**). The generally
lower frequency of subgenome- or polyploid-specific ACRs, compared with the more abundant
lineage-specific ACRs or peanut-conserved ACRs with biased chromatin accessibility, supports
the conclusion that regulatory diversification through gradual mutation accumulation is more
common than *de novo* emergence (10, 11, 22).
484
Importantly, we found that even minor sequence changes can have disproportionately large
effects on regulatory activity. Peanut-conserved ACRs with biased chromatin accessibility
exemplify this, where they exhibited high sequence similarity but substantial chromatin
accessibility differences that often manifest as presence/absence patterns rather than gradual
quantitative changes (**Fig. 3D**). Furthermore, we demonstrate a correlation between changes in
CNS composition and chromatin accessibility bias, a relationship not previously examined in
lineage-specific contexts (10). For instance, the peanut-conserved ACRs with parental
subgenome-dominant bias chromatin accessibility always possessed more CNSs that conserved
between the respective subgenome and its corresponding diploid, whereas peanut-conserved
ACRs with parental no bias chromatin accessibility were enriched for CNSs that conserved across
four genomes (**Fig. 5**). Although these associations are compelling, they remain correlative.
Without allele-specific assays or targeted perturbations, we cannot ascribe causality to CNS
turnover, and trans-acting factors, peak-calling thresholds, or tissue composition may also
contribute (62–64). A limitation is inherent to CNS identification itself, which requires a
predefined degree of sequence conservation. The large number of CNSs conserved exclusively in
one or both progenitors but absent from the subgenomes (**Fig. 4D-4E**) may suggest that these

sequences have undergone functional divergence in the subgenomes, shifting toward new
regulatory roles at the expense of detectable sequence conservation, even if the underlying
sequence changes might be minor. On the other hand, examining transcription factor binding
motifs (65) or footprinting data (56, 66) will also be essential for understanding how conserved
elements transition into novel regulatory functions and how frequently such events occur.
506
507 In summary, our study provides empirical evidence illuminating the evolutionary dynamics of
508 regulatory elements in polyploid plants. By integrating sequence variation, CNSs turnover,
509 chromatin accessibility, histone modification, and expression divergence, we provide a
510 comprehensive view of how *cis*-regulatory evolution drives expression diversification and
511 contributes to evolutionary innovation in plant genomes.
512
Materials and Methods
Plant materials
Three Arachis species were used in this study: the cultivated allotetraploid peanut *A. hypogaea*
(AABB, $2n=4x=40$) and its two diploid progenitors, *A. duranensis* (AA, $2n=2x=20$) and *A.*
ipaensis (BB, $2n=2x=20$). Plants were grown in soil for approximately 10 days at 25 °C under
photoperiodic lighting (16 h light–8 h dark). Mature leaf and root tissues were collected with
two biological replicates per tissue for each species.
520
ATAC-seq library preparation
Nuclei isolation and purification were performed as described previously with modifications
(67). Then, the nuclei quality and density were evaluated with a hemocytometer under a
microscope. The ATAC-seq library amplification was performed following the protocol described
previously (68). About 50k nuclei were incubated with 2 µl Tn5 transposomes in 40 µl
tagmentation buffer at 37 °C for 30 min without rotation. The integration products were purified
using a Qiagen MinElute PCR Purification Kit or NEB Monarch DNA Cleanup Kit and then
amplified using Phusion DNA polymerase for 10–13 cycles. Amplified libraries were purified
using AMPure beads to remove primers.
530
RNA-seq library preparation
The collected tissues were flash-frozen with liquid N2 immediately after collection and stored at
-80C frezeer. The samples were ground to a powder using a pestle and mortar in liquid N2. Total
RNA was extracted and purified using TRIzol Reagent (Thermo Fisher Scientific) following the
manufacturer's instructions. For each sample, 1.3 µg of total RNA was prepared for sequencing
using the Illumina TruSeq mRNA Stranded Library Kit (Illumina) following the manufacturer's
instructions.
538
ChIP-seq library preparation
ChIP was performed following the general protocol of (69). Approximately 0.1 g of tissue were
used for each experiment. The fresh tissue was chopped into 0.5 mm cross sections and
crosslinked in accordance with the referenced protocol. Samples were immediately flash-frozen
in liquid nitrogen after crosslinking. Nuclei were extracted and lysed in 300 µl of lysis buffer and
was sonicated on a Diagenode Bioruptor. Tubes were centrifuged at 12,000g for 5 min and
supernatants transferred to new tubes. Dynabeads Protein A (Thermo Fisher Scientific, cat. no.
10002D) were washed with ChIP dilution buffer and then rotated with antibodies (anti-

H3K4me3, Millipore-Sigma, 07–473; anti-H3K36me3, Abcam, ab9050; anti-H3K56ac, Millipore-
Sigma, 07–667). 100 µl of ChIP dilution buffer for at least 4 h at 4C. Sonicated chromatin was
diluted tenfold in ChIP dilution buffer. For all samples and replicates, 460 µl of diluted chromatin
was incubated with 750 µg of Dynabeads Protein A coated with 1.5 µg of antibody. Samples
were rotated at 4 °C overnight, then washed, reverse-crosslinked, and treated with proteinase K
in accordance with the referenced protocol. DNA was purified by a standard phenol–chloroform
extraction followed by ethanol precipitation. The final libraries were prepared following exactly
with the protocol (69).
555
556   **Sequencing information**
557   All libraries were sequenced using an Illumina NovaSeq 6000 S4 150PE. All the raw reads of
558   ATAC-seq, ChIP-seq and RNA were uploaded to the National Center for Biotechnology
559   Information (NCBI).
560
ATAC-seq data processing and alignment
Raw reads from 12 libraries were cleaned with Trimmomatic version 0.36 (70) using a sliding
window trimming method with default setting except the minimal length of the reads as 30 bp.
The resulting clean reads were mapped to the respective reference genomes, including *A.*
hypogaea, *A. duranensis* and *A. ipaensis* from Legume Information System
(www.legumeinfo.org/collections/arachis/) using BWA-MEM aligner (71). For *A. hypogaea*, the
clean reads were mapped to each subgenome separately and then retained those uniquely
mapped with mapping quality >30 for further analyses. In addition, reads mapped to a blacklist
of regions exhibiting Tn5 integration bias from Tn5-treated genomic DNA and potential
collapsed sequences in the reference (1-kb windows with greater than 4x coverage over the
genome-wide median using ChIP-seq input) were removed using BEDtools (72). The clonal
duplicated were removed using Picard (73). Only ACRs located in the chromosomes were
retained for the following analyses.
574
ACR identification and characterization
MACS2 (74) was used to identify ACRs with the following parameters: --extsize 150 --shift -75 --
slocal 150 --llocal 1500 --max-gap 75 --min-length 175 --qvalue 0.05. To refine the peak calling
and identify small peaks, we used the makewindows function in BEDtools to split the identified
ACRs. Finally, the refined peaks that overlapped or within 50 bp were merged using the
BEDtools merge function. Proportion of reads around transcription start site (TSS) was
calculated as the number of identified ACRs are overlapped with 1 kp upstream of TSS. Fraction
of reads in peak (FRiP) was calculated as the percentage of the sequenced reads that overlapped
with the peaks.
584
ChIP-seq analysis
Reads were cleaned and mapped to reference genome using the same pipeline as for ATAC-seq
analysis. Only uniquely mapped reads with a mapping quality greater than 30 were retained for
downstream analyses. To normalize sequencing depth and compare libraries to input data, we
used the bamCompare function from deepTools2 (75) with the following parameters: "--
normalizeUsing BPM --pseudocount 1 --binSize 10 –extendReads". Signal densities for the
histone modifications from 2kb upstream to 2kb downstream around the genes were calculated
using the "computeMatrix" function in "scale-regions" mode with the setting of "-bs 20 -b 2000

-a 2000 –regionBodyLength 5000". Peak calling was performed using MACS2 (74) with the H3
ChIP-seq samples as controls. For H3K4me3, and H3K56ac ChIP-seq samples, narrow peaks were
identified using the parameters -p 1e-3 -f BAM --keep-dup all. For H3K36me3 ChIP-seq samples,
broad peaks were identified using the parameters --broad --broad-cutoff 0.05. The overlap
between histone modification peaks and the ACRs or their corresponding BLASTed sequences
were performed using bedtoolds. The resulting Matrices were loaded to a custom R script to
generate metaplots and do cluster analyses following (12).
600

Differential chromatin accessibility
To examine quantitative differences in ACRs between the two tissue-types from the same
species, we generated four pseudo replicates by randomly sampling the equal number of clean
reads from the two biological replicates using a customized Python script. Normalized read
counts were then calculated for each peak, and DESeq2 (76) was used to identify regions with
significant differences in accessibility. Peaks with an adjusted P-value (FDR) < 0.05 and an
absolute $\log_2$ fold change > 1 were defined as differential ACRs.
608

The same analytical framework was used to examine accessibility bias across peanut-conserved
ACRs and their corresponding BLASTed regions in the reference genomes (**Fig. 2B**). Parental no
bias was defined as the absence of significant accessibility differences between the two diploids
and between the two subgenomes. Parental subgenome A or B dominant bias was assigned
when a subgenome and its corresponding progenitor both exhibited significantly higher
accessibility relative to the other pair. No bias in progeny referred to cases with no significant
difference between the two subgenomes, further divided into progenitor A- or B-dominant
when one progenitor showed significantly higher chromatin accessibility. Novel bias in progeny
was defined as subgenome A- or B-dominant accessibility in the absence of significant
differences between the progenitors. Finally, reverse bias in progeny (subgenome A- or
subgenome B- dominant) referred to cases where a subgenome showed significantly higher
chromatin accessibility, but its corresponding progenitor had significantly lower chromatin
accessibility.
622

Sequence conservation of ACRs identified in polyploidy peanut genome
To investigate the sequence ancestry of ACRs identified in the two subgenomes of polyploid
peanut, we adopted a BLAST-based approach following the methods described in (13). ACR
sequences were used as queries, and reference regions were defined as the genomic intervals
spanning the two orthologous genes flanking each ACR in the corresponding genomes (Fig. S2).
These reference regions were identified using GENESPACE (77) and classified based on
conservation of the orthologous gene pairs – either conserved across all four genomes (peanut-
conserved), conserved across three genomes or between a subgenome and its non-
corresponding diploid progenitor (potential peanut-conserved), exclusively conserved between
a subgenome and its corresponding diploid progenitor (lineage-specific), exclusively between
the two subgenomes (polyploid-specific), or conserved only in one subgenome (subgenome-
specific). ACRs from each subgenome were BLASTed against the reference regions using BLAST
(78) and the same parameters as described in (13). Sequence intersections were identified using
BEDtools (72) and sequence extractions were performed with seqKit (79). The resulting BLAST
outputs were then combined and filtered using customized scripts.
638

Analyses of CNS dynamics in polyploid-diploid framework

CNS information of three peanut species, *A. hypogaea*, *A. duranensis* and *A. ipaensis*. Was
downloaded from (https://conservatorycns.com/dist/pages/conservatory/index.php). We
summarized the counts, repeats, and conserved information for each CNSs across three peanut
species or four genomes using customed R script. When examining the CNSs located within the
sequence-conserved ACRs, we first used BEDtools to intersect the coordinates of these CNSs
with the coordinates of sequence-conserved ACRs or their corresponding BLASTed sequences,
with the requirement that the CNSs might be completely within the ACRs or BLASTed sequences
regions using the commend of -f 1.0. Then we analyzed all the CNSs within each ACR separately
and further grouped them by the genes they are associated with, requiring that genes in each
group belong to the same orthogroup. Subsequently, we categorized CNSs based on their
conservation states across four genomes – two subgenomes in polyploid peanut and two diploid
progenitors and divided them as: (i) conserved across all four genomes, (ii) conserved across
three genomes, (iii) conserved exclusively in progenitors, (iv) conserved exclusively in polyploid
subgenomes, (v) conserved between a subgenome and its corresponding progenitor, (vi)
conserved between a subgenome and the non-corresponding progenitor, and (vii) restricted to
a single genome.

To quantify the relative enrichment of conservation patterns across CNS categories within each
orthogroup within each ACR, we generated a rank-normalized matrix from the CNS
categorization count matrix. For each CNS, we first identified the categories with non-zero
counts. Categories with zero counts were treated as absent and assigned a rank-normalized
score of zero. For categories with positive values, we computed ranks based on descending
abundance, such that the highest count received rank 1 and the lowest received rank N, where
N is the number of present categories. These ranks were then normalized by dividing by N,
producing values between 1/N and 1. The resulting rank-normalized scores were placed back
into a full vector, with zeros retained for absent categories. This procedure was applied row-
wise to the CNS matrix, producing a final rank-normalized matrix with the same dimensions as
the original input.

RNA-seq data processing and alignment

The adapters and low-quality reads were trimmed using Timmomatic with the following
parameters: SLIDINGWINDOW: 4:15 LEADING:3 TRAILING:3 MINLEN:25. The clean reads were
mapped to reference geome using STAR (80). Gene expression is quantified using GeneCounts
mode and only uniquely mapped reads are counted (--outFilterMultimapNmax 1 --
outFilterMismatchNmax 999).

Differential gene expression analysis

Peptide sequences from the two subgenomes of *A. hypogaea* and the genomes of *A. duranensis*
and *A. ipaensis* were used to identify putative orthologs using OrthoFinder (81). Orthogroups
containing single-copy homologs in a 1:1:1:1 ratio across these four genomes were retained.
Variance-stabilizing transformed gene counts, normalized with DESeq2, were used for PCA. Leaf
and root tissues were analyzed separately, and only orthogroups with an average normalized
gene count greater than 5 in at least one genome were included in subsequent analyses.
Pairwise differential expression analyses were performed in 'DESeq2' package, with genes
showing an adjusted P-value (FDR) < 0.05 and an absolute $\log_2$ fold change > 1 considered

685 differentially expressed. The same criteria used to classify accessibility bias was applied to
686 categorize homoeolog expression bias.
687

Enrichment between expression bias and ACR types

Only Orthogroups exclusively associated with ACRs that had been assigned an evolutionary
Only Orthogroups exclusively associated with ACRs that had been assigned an evolutionary
origin were included in the analysis. Expression bias categories were simplified into four types:
no bias (including "no bias in progeny" and "parental no bias"), novel bias (including "novel bias"
and "novel bias reverse"), parental-legacy bias (including "parental subgenome A- dominant"
and "parental subgenome B- dominant"), and others. Associated ACRs were first classified by
sequence conservation into peanut-conserved, potential peanut-conserved, lineage-specific,
polyploid-specific, and subgenome-specific. Peanut-conserved ACRs were further subdivided
based on their simplified chromatin accessibility bias (analogous to expression bias) into no bias,
novel bias, parental-legacy bias, and others. Permutation-based enrichment analyses were
performed separately for leaf and root tissues using both unfiltered and filtered datasets (with
distal ACRs removed).
700

For each Orthogroup, the proportion of ACRs belonging to each category was calculated by
dividing the number of ACRs of that type by the total number of ACRs within that Orthogroup.
These normalized proportions were then aggregated by expression bias groups. Observed mean
proportions for each ACR category were calculated across Orthogroups within each expression
group. To evaluate the significance of these observed proportions, 10,000 random permutations
were conducted by shuffling expression bias group labels while preserving the ACR composition
within each Orthogroup. For each permutation, mean proportions were recalculated to
generate a null distribution. The null mean and standard deviation were then used to compute
standardized Z-scores, and empirical P-values were derived as the proportion of permuted
means equal to or greater than the observed mean proportion. Multiple testing correction was
applied using the Benjamini–Hochberg (BH) method, and categories with adjusted $P < 0.05$ were
considered significantly enriched.
713

Acknowledgments

714 **Acknowledgments**
715 We sincerely appreciate the members of the Schmitz lab for their valuable analytical
716 suggestions. J.P.M. was supported by an NIH Training Grant (T32GM142623). This research was
717 supported by the National Science Foundation (IOS-1856627) awarded to R.J.S.
718

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 Figure captions
Figure 1. Characterization of ACRs identified in peanut and diploid progenitor.
(A) Screenshot illustrating example of tissue-specific ACRs and broad ACRs. **(B)** Number and
proportion of tissue-specific versus broad ACRs in subgenome B of *A. hypogaea* (top) and
progenitor B (*A. ipaensis*) (bottom). **(C)** Distribution of genic, proximal and distal ACRs within
broad, leaf-specific and root-specific categories in subgenome B (top) and progenitor genome B
(bottom).
904
Figure 2. Identification and characterization of ACRs with distinct evolutionary trajectories.
(A) Schematic illustrating the classification of ACRs based on sequence conservation profiles.
Peanut-conserved ACRs are retained across all four genomes, the two subgenomes of polyploid
peanut and both diploid progenitors, reflecting ancestral conservation predating progenitor
divergence. Potential peanut-conserved ACRs are detected in three genomes, or between a
subgenome and a non-corresponding progenitor. Lineage-specific ACRs are shared between a
subgenome and its corresponding progenitor, representing inheritance from the respective
diploid lineage. Polyploid-specific ACRs likely originated after hybridization but before whole-
genome duplication. Subgenome-specific ACRs are unique to one subgenome, suggesting *de
novo* emergence following polyploidization. **(B)** Distribution of ACRs with distinct evolutionary
trajectories (identified in subgenome A) under two sequence-similarity thresholds, using
BLASTed ratios ≥50% (left) and ≥80% (right) to define sequence conservation. **(C)** Proportion of
each evolutionary ACR category overlapping with TEs (in subgenome B), with the criteria of at
least 20 bp overlap. **(D)** Genetic variation across peanut-conserved, potential peanut-conserved,
lineage-specific, and polyploid-specific ACRs (in subgenome B), measured by mean mismatch
rate and mean BLASTed ratio. **(E)** Distribution of distinct histone modification patterns across
conserved ACRs and their corresponding BLASTed regions in four genomes, the two
subgenomes of polyploid peanut and both diploid progenitors. The top panel summarizes the
three histone modification states (purple, blue and green bars) defined by combinations of the
histone modification marks, H3K4me3, H3K56ac and H3K36me3, with grey bars indicating the
presence of each respective mark. The bottom-left schematic illustrates the classification of
histone modification patterns shown from top to bottom: (i) patterns conserved across all four
genomes; (ii) patterns shared between the two subgenomes but different in both progenitors;
(iii) patterns shared between both subgenomes and both progenitors but differing between
subgenome and progenitor groups; (iv) patterns distinct between the two subgenomes but each
matching its corresponding progenitor; (v) patterns consistent only between subgenome A and
progenitor A; (vi) patterns consistent only between subgenome B and progenitor B; and (vii)
patterns different from both progenitors. The top three categories represent conserved histone
modification patterns between the two subgenomes, whereas the remaining four represent
divergent patterns. The bottom-right heatmap shows the proportion of each histone
modification pattern in leaf and root tissues within subgenome A and subgenome B. P_A:
progenitor A; P_B: progenitor B; S_A: subgenome A and S_B: subgenome B.
937
Figure 3. Characterization of chromatin accessibility bias among peanut-conserved ACRs.
(A) Principal component analyses (PCA) of normalized chromatin accessibility counts for peanut-
conserved ACRs (in subgenome A) across four genome types: two subgenomes and their
corresponding diploid progenitors (left) and across two tissue types: leaf and root (right). **(B)**
Schematic illustrating classification of chromatin accessibility bias based on the relative

chromatin accessibility patterns between subgenomes and between their corresponding
progenitors. Categories include (from left to right): parental no bias (no bias between either
subgenomes or progenitors); parental subgenome A- dominant or subgenome B- dominant (bias
direction in subgenome consistent with that in progenitors); no bias progenitor A- dominant or
progenitor B-dominant (no bias between subgenomes but bias present between progenitors); novel
novel bias subgenome A- or subgenome B- dominant (bias emerging between subgenomes
despite no bias between progenitors); and novel bias-reverse (novel bias -r) subgenome A- or
subgenome B-dominant (bias direction opposite the observed between progenitors). **(C)**
Distribution of these chromatin accessibility bias categories for peanut-conserved ACRs
identified in subgenome A and subgenome B across leaf and root tissues. **(D)** Overlap rate
between BLASTed regions of peanut-conserved ACRs (identified in subgenome B) in subgenome
A and the ACRs identified in subgenome A; red labels denote subgenome B-dominant ACRs. P_A:
progenitor A; P_B: progenitor B; S_A: subgenome A and S_B: subgenome B.
956
Figure 4. Characterization of CNS dynamics in the polyploid-diploid peanut system.
(A) Number and proportion of CNSs identified in the three peanut species, grouped by their
evolutionary age of conservation, from ancient Tracheophyte-level or Angiosperm-level CNSs to
increasingly lineage-restricted CNSs in Dicots, Rosids, and Fabaceae. **(B)** Distribution of CNS copy
numbers across CNSs with different evolutionary age of conservation. **(C)** Distribution of CNSs
with different conservation levels across the three peanut species (left) and across the four
genomes, the two subgenomes of polyploid peanut and the two diploid progenitors (right), for
each evolutionary age of conservation category. Numbers from 1 to 4 indicate CNSs conserved
in one, two, three, or all four species/genomes. **(D-E)** Complex spectrum of CNS dynamics (CNS
conservation categories) across the diploid-polyploid framework for Angiosperm-level CNSs (D)
and Fabaceae-level CNSs (E). CNSs are categorized as: (i) conserved across all four genomes, (ii)
conserved across three genomes, (iii) conserved exclusively in progenitors, (iv) conserved
exclusively in polyploid subgenomes, (v) conserved between a subgenome and its corresponding
progenitor, (vi) conserved between a subgenome and the non-corresponding progenitor, and
(vii) restricted to a single genome. P_A: progenitor A; P_B: progenitor B; S_A: subgenome A and S_B:
subgenome B.
973
Figure 5. Rank-normalized scores of CNS conservation categories across ACRs with different
chromatin accessibility bias.
Heatmap showing the rank-normalized scores of CNS conservation patterns (as defined in Fig.
4D-4E) across peanut-conserved ACRs in subgenome B, annotated according to leaf chromatin
accessibility categories (as in Fig. 3C). For each ACR, CNS conservation categories were
summarized using a rank-normalization approach: all non-zero CNS conservation categories
were ranked in descending order of abundance and normalized by the number of categories
present, producing values between 0 and 1. A value of 1 indicates that an ACR contains CNSs
exclusively from a single conservation category. Rows are grouped by ACR chromatin
accessibility categories (as in Fig. 3C), and the color scale reflects the relative rank-normalized
score within each ACR. P_A: progenitor A; P_B: progenitor B; S_A: subgenome A and S_B: subgenome
B.
986
Figure 6. Homeolog expression bias in peanut and its association with ACRs of distinct
evolutionary trajectories.

(A) Principal component analyses (PCA) of normalized expression counts for homoeologous
genes across four genome types: two subgenomes and their corresponding diploid progenitors
(top) and across two tissue types: leaf and root (bottom). **(B)** Distribution of homeolog
expression bias categories across leaf and root tissues. Biased expression categories were
labelled with red font. **(C)** Heatmap showing enrichment patterns between expression types
(no-bias, novel bias and parental-legacy bias) and ACR evolutionary types including peanut-
conserved (no bias), peanut-conserved (novel bias) peanut-conserved (parental-legacy bias),
lineage-specific, polyploid-specific and subgenome-specific ACRs. Absolute Z-scores greater than
2 are indicated.
998

A

Progenitor A

Subgenome A
Progenitor B

Subgenome B

△ Peanut-conserved △ Potential peanut-conserved △ Lineage-specific △ Polyploidy-specific △ Subgenome-specific

B

Threshold (50%)   Threshold (80%)

C

100%

75%

50%

25%

0%

Lineage-specific Peanut-conserved Potential peanut-conserved Polyploid-specific Subgenome-specific

D

Mean mismatch rate   Mean BLASTed ratio

0.08

0.04

0.00

Lineage-specific Peanut-conserved Potential peanut-conserved Polyploid-specific

1.1

1.0

0.9

0.8

Lineage-specific Peanut-conserved Potential peanut-conserved Polyploid-specific

E

H3K4me3
H3K56ac
H3K36me3

Conserved

Divergent

Ratio
0.3

0.2

0.1

P_A P_B S_A S_B   Leaf Root Leaf Root
SubgenomeA   SubgenomeB

CNS conservation states