Transcriptomic and epigenomic characterization of the developing bat wing

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Bats are the only mammals capable of powered flight, but little is known about the genetic determinants that shape their wings. Here we generated a genome for Miniopterus natalensis and performed RNA-seq and ChIP-seq (H3K27ac and H3K27me3) analyses on its developing forelimb and hindlimb autopods at sequential embryonic stages to decipher the molecular events that underlie bat wing development. Over 7,000 genes and several long noncoding RNAs, including Tbx5-as1 and Hot tip, were differentially expressed between forelimb and hindlimb, and across different stages. ChIP-seq analysis identified thousands of regions that are differentially modified in forelimb and hindlimb. Comparative genomics found 2,796 bat-accelerated regions within H3K27ac peaks, several of which cluster near limb-associated genes. Pathway analyses highlighted multiple ribosomal proteins and known limb patterning signaling pathways as differentially regulated and implicated increased forelimb mesenchymal condensation in differential growth. In combination, our work outlines multiple genetic components that likely contribute to bat wing formation, providing insights into this morphological innovation.

The order Chiroptera, commonly known as bats, is the only group of mammals to have evolved the capability of flight. They are estimated to have diverged from their arboreal ancestors ~51 million years ago. Their adaptations for flight include substantial specialization of the forelimb, characterized by the notable extension of digits II–V, a decrease in wing bone mineralization along the proximal-distal axis, and the retention and expansion of interdigit webbing, which is controlled by several genes, including Tbx3, Brinp3, Meis2, the 5′ HoxD genes and components of the Shh-Fgf signaling loop, providing an informative contrast that can be used to highlight the genetic processes involved in bat wing formation. Previous studies that examined gene expression in developing bat forelimbs and hindlimbs reported differential expression of several genes, including Tbx3, Brinp3, Meis2, the 5′ HoxD genes and components of the Shh-Fgf signaling loop, suggesting that multiple genes and processes are involved in generating these morphological innovations4–8. Gene regulatory elements are thought to be important drivers of these changes: for example, replacement of the mouse PrxI limb enhancer with the equivalent bat sequence resulted in elongated forelimbs9. However, an integrated understanding of how changes in regulatory elements, various genes and signaling pathways combine to collectively shape the bat wing remains largely elusive.

To characterize the genetic differences that underlie divergence in bat forelimb and hindlimb development, we used a comprehensive, genome-wide strategy. We generated a de novo whole-genome assembly for the vesper bat, M. natalensis, for which a well-characterized stage-by-stage morphological comparison between developing bat and mouse limbs is available10. In this species, the developing forelimb noticeably diverges from the hindlimb from developmental stages CS15 and CS16, with clear morphological differences seen at a subsequent stage, CS17 (ref. 10). This developmental window is equivalent to embryonic day (E) 12.0 to E13.5 in mouse9,10. M. natalensis embryos were obtained and transcriptomic (RNA-seq) data and ChIP-seq data for both an active (acetylation of histone H3 at lysine 27, H3K27ac; refs. 11,12) and a repressive (trimethylation of histone H3 at lysine 9, H3K9me3; refs. 11,13) mark were generated for these three developmental stages (Fig. 1).

RESULTS

The M. natalensis genome

High-coverage genomes for three bat species (Pteropus alecto14, Myotis davidii14 and Myotis brandtii15) and low-coverage genomes for two bat species (Myotis lucifugus and Pteropus vampyrus16) have

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been published. However, the evolutionary distance of these species from *M. natalensis* (43 million years since the last common ancestor) precludes the use of their genomes in RNA-seq and ChIP-seq data analyses. We thus generated a draft genome from an adult *M. natalensis* male at 77x coverage, named Mnat.v1. The quality of the Mnat.v1 genome is comparable to that of the high-coverage bat genomes (Supplementary Table 1). It has an estimated heterozygosity level of 0.13%, with repetitive regions making up 33% of the genome. We annotated 24,239 genes (including protein-coding genes and long noncoding RNAs (lncRNAs)) in Mnat.v1. Of the highly conserved genes used by the Core Eukaryotic Genes Mapping Approach (CEGMA)\(^1\), 92.7% were found in their entirety, with an additional 3.3% partially detected, further confirming Mnat.v1 to be a reliable substrate for subsequent genomic analyses.

Differentially expressed limb transcripts

To identify gene expression differences that could be involved in the morphological divergence in bat limb development, we examined the transcriptomes of whole-autopod tissue from forelimbs and hindlimbs at three sequential developmental stages (CS15, CS16 and CS17). Principal-component analysis (PCA) showed an expected segregation pattern, with principal component 1 (PC1) reflecting the developmental stage and principal component 2 (PC2) reflecting the tissue type (forelimb or hindlimb; Fig. 2a). We found 2,952 genes differentially expressed in forelimb and hindlimb and 5,164 genes differentially expressed in comparisons of any two sequential stages (adjusted \(P\) value \(\leq 0.01\); Online Methods). Pairwise tests for differential expression directly comparing forelimb and hindlimb at each stage (for example, CS15 forelimb versus CS15 hindlimb) contributed an additional 1,596 genes. Combined, these analyses identified 7,172 differentially expressed genes (adjusted \(P\) value \(\leq 0.01\); Fig. 2b and Supplementary Table 2).

Differentially expressed genes were grouped by their expression profiles across the samples into 38 manually defined clusters using hierarchical clustering (Supplementary Fig. 1). These clusters were functionally annotated, with several Gene Ontology (GO) terms correlated with differential expression (Fig. 2b and Supplementary Table 3). Grouping the genes displaying the most significant differential expression on the basis of biological functions of interest (for example, DNA binding and transcriptionsal regulation, limb morphogenesis, bone morphogenesis, apoptotic process and others) identified both genes with known roles in limb development and genes with potentially new functions in bat wing development (Supplementary Fig. 2). For example, the genes differentially expressed in forelimb and hindlimb involved in DNA binding and transcriptional regulation included *Hoxd10, Hoxd11, Meis2, Pitx1, Tbx4* and *Tbx5*, comprising all genes that were previously shown to be differentially expressed in bats\(^5\)-\(^7\) along with several genes showing higher expression in forelimb that have not yet been characterized (Fig. 2c). We also observed hindlimb-specific increased expression for several genes, notably *Msx1* and *Msx2*, both of which are key genes involved in apoptotic activity during interdigit tissue regression\(^18\).

We characterized a limited number of differentially expressed genes of interest in both mouse and bat embryos using whole-mount *in situ* hybridization (Supplementary Fig. 3). Among these, *Mllt3* was chosen for its strong forelimb expression at CS15 and CS16 (Fig. 2c) and was found to be uniquely expressed in bat forelimb in a region restricted to the distal edge where digits III–V are slated to develop (Fig. 2d). *Mllt3* is thought to be a Hox gene regulator, with *Mllt3*-null mouse mutants exhibiting axial defects\(^19\); however, no gross skeletal limb abnormalities were observed in homozygous knockout mice (Online Methods and Supplementary Fig. 4). *Lhx8*, a known regulator of neuronal development\(^20\), had higher expression in CS16 and CS17 forelimbs (Fig. 2c). Whole-mount *in situ* hybridization analysis showed localized *Lhx8* expression in the posterior portion of the wrist region, specifically in the junction between the base of digit V and the plagiopatagium, whereas no expression was detected in mouse limbs (Fig. 2d). Together, these experiments support our RNA-seq analyses and highlight genes that have previously uncharacterized roles in limb development.

Bat-specific lncRNAs

lncRNAs have been shown to be important developmental regulators in several tissues, including the limb\(^21\). To find potential lncRNAs associated with bat limb development, we annotated transcripts that did not show similarity to known protein-coding genes, identifying 227 potential lncRNAs (Supplementary Table 4). Among these, 188 exhibited some sequence conservation across mammals, 12 of which were similar to characterized lncRNAs in IncRNAdb v2.0 (ref. 22). Five putative lncRNAs were identified as being conserved only in bats, and 34 were only present in *M. natalensis*. Within this data set, eight known IncRNAs showed differential expression between forelimb and hindlimb, including Hottip and an uncharacterized lncRNA, Tbx5-as1 (Fig. 3). Hottip is thought to be required for activation of 5′ HoxA genes, which are important regulators of autopod patterning during limb development\(^21\). Both Hottip and *Hoxa13* showed elevated lncRNA expression in all three stages examined (Fig. 3a). A comparison of their expression patterns showed both to be more strongly expressed in interdigit tissue of the bat hindlimb. Although Hottip expression was concentrated in distal interdigit tissue, *Hoxa13* expression was more apparent in digit tips (Fig. 3b,c). The bat Tbx5-as1 transcript maps close to Tbx5 (Fig. 4a), in the antisense orientation and is similar to human Tbx5 antisense RNA1-as1 transcript (GenBank, NR_038440). Tbx5-as1 was the most differentially expressed lncRNA, with elevated expression in the forelimb relative to the hindlimb across all stages (Fig. 3a). Although its role is unknown, its associated gene, *Tbx5*, is required for forelimb bud initiation, with
inactivation in mice abolishing forelimb skeletal formation\textsuperscript{13,24}. In support of coupled activity for these transcripts, the expression patterns of Tbx5 and Tbx5-as1 were similar: both were restricted to the base of digits I–V during late CS16 (CS16L) and CS17, with clear expression in proximal interdigit tissue (Fig. 3d,e).

**ChIP-seq highlights forelimb regulatory regions**

Changes in gene regulatory elements have been shown to be important drivers of morphological adaptations\textsuperscript{25}, including the bat wing\textsuperscript{9}. To identify regulatory elements that could be involved in controlling gene expression in developing bat limbs, we performed low-cell-number ChIP-seq using antibodies for both H3K27ac (active regions\textsuperscript{14,15}) and H3K27me3 (pressed regions\textsuperscript{13}) on autopods from CS15, CS16 and CS17 forelimbs and hindlimbs, identifying numerous putative regulatory regions (Supplementary Table 5). Using the Genomic Regions Enrichment of Annotations Tool (GREAT)\textsuperscript{26}, after converting peaks to mouse genomic coordinates, we found significant enrichment for the H3K27ac peaks in several limb development–associated categories among GO morphological processes, mouse phenotypes and Mouse Genome Informatics (MGI) expression data (Supplementary Fig. 5).
To further validate our ChIP-seq results, we examined genes known to be specifically expressed in the forelimb (Tbx5) or hindlimb (Pitx1) and observed correspondence with H3K27ac and H3K27me3 peak presence and RNA expression (Fig. 4a,b).

We next set out to analyze differences between forelimb and hindlimb active and repressed ChIP-seq peaks. Differential enrichment analysis was carried out on H3K27ac and H3K27me3 peaks separately, identifying 14,553 and 19,352 differentially enriched regions, respectively (pairwise false discovery rate (FDR) ≤ 0.05; Online Methods and Supplementary Table 6). Of these regions, 2,475 were differentially enriched between forelimb and hindlimb for both H3K27ac and H3K27me3 signal. These regions were analyzed using hierarchical clustering based on H3K27ac and H3K27me3 enrichment (Fig. 4c), from which 17 manually defined clusters were identified with distinct H3K27ac and H3K27me3 enrichment patterns (Supplementary Fig. 6). GO term analysis of the nearest gene for each region showed enrichment for terms associated with limb development. For example, cluster 9 showed higher H3K27ac levels in forelimb and H3K27me3 levels in hindlimb, whereas cluster 11 showed the opposite pattern. The regulatory marks for both clusters had a general correspondence with RNA-seq expression levels for the neighboring genes and included fitting GO biological term enrichment for developmental processes (Fig. 4c and Supplementary Table 7).

Bat accelerated regions

To identify genomic changes that might be associated with the innovation of the bat wing, we used a comparative genomics approach27 that leveraged the growing number of bat genomes4–10. Whole-genome alignments were generated using the repeat-masked genomes of 18 other species, including 6 bats, 9 non-bat mammals and 3 non-mammal vertebrates. We next used phyloP28 to identify accelerated sequences, sequences that are conserved in vertebrates but changed significantly in the common ancestor of the bat lineage and were marked by H3K27ac in all ChIP-seq experiments. This analysis identified 2,796 bat accelerated regions (BARs; FDR ≤ 0.05) with an average size of 240 bp (Supplementary Table 8). Genomic regions over-represented for BARs were identified by comparison to vertebrate conserved regions overlapping H3K27ac peaks. Genes contained in these regions were subjected to functional annotation clustering, showing enrichment for categories relating to transcription factors, chromatin conformation and DNA binding (FDR ≤ 0.05; Supplementary Table 8). The region most highly enriched for BARs included the genes Lrrn1 (leucine-rich repeat neuronal 1) and Crbn (cereblon) (Supplementary Fig. 7a). Lrrn1 is expressed at significantly higher levels in bat hindlimb than forelimb (adjusted P value = 6.73 × 10^{-11}; Supplementary Table 2) and was also shown to be expressed in developing mouse limbs29. It was also shown to be important for midbrain-hindbrain boundary formation regulated by Fgf8 (ref. 30). Crbn is a known thalidomide target, thought to be important in limb outgrowth through its regulation of Fgfs31, but did not show significant expression differences between forelimb and hindlimb (Supplementary Table 2). Another BAR-dense region was around Fgf2 and Spry1 (Supplementary Fig. 7b). Fgf2 is known to have regenerative capabilities in the limb32 and was both the most highly expressed and had the most significant fold change in expression between forelimb and hindlimb across all stages among the Fgf genes in our study (Supplementary Fig. 7c). Spry1 was shown to be involved in limb muscle and tendon development33 but did not have significant expression differences between forelimb and hindlimb (Supplementary Fig. 7c). Combined, our ChIP-seq and BAR analyses highlight specific candidate sequences and genomic regions that might have had a role in development of the bat wing.

Wing developmental pathways

We next used Ingenuity Pathway Analysis (IPA) to identify signaling pathways that were differentially activated across the data set and could contribute to the differences in patterning between bat forelimb and hindlimb autopods. Interestingly, the top pathway in our analysis, showing strong hindlimb activation, was the elongation initiation factor 2 (EIF2) signaling pathway (Fig. 5a and Supplementary Table 9), which has an important role in regulating the initiation of protein synthesis. A closer inspection showed that 41 ribosomal protein genes, which are coordinately downregulated in bat forelimb at CS15 and CS16 (Fig. 5b), were largely responsible for this score.
Ribosomal protein expression has been shown to be highly heterogeneous across tissues during embryonic development, including in the limb. Mutations in the loci encoding the ribosomal proteins RPL11, RPL35A, RPS7, RPS10 and RPS19 are known to lead to limb malformations in individuals with Diamond-Blackfan anemia. The Rpl38 gene, whose encoded protein facilitates the translation of several HoxA genes by an IRES-dependent mechanism and that is mutated in tail short mice, which have skeletal patterning defects, was downregulated in CS15 and CS16 bat forelimbs. Rictor, a negative upstream regulator of these ribosomal proteins, showed higher expression in forelimb (Supplementary Table 2). Rictor is a subunit of the mTORC2 complex, having a role in actin cytoskeleton organization, with conditional deletions in mice resulting in narrower and shorter limb bones. Combined, our pathway analyses suggest that ribosomal proteins and their regulators could have an important role in bat wing development through the translational control of specific subsets of mRNA transcripts.

Several pathways known to have an important role in limb and bone development, including Fgf, Wnt and Bmp signaling, were among the top ten IPA canonical pathways coordinately activated or repressed in bat forelimb as compared to hindlimb (Fig. 5a). Fgf proteins are known to mediate limb patterning by signaling the initial outgrowth of the limb bud from the apical ectodermal ridge. This pathway showed consistent activation in CS15–CS17 forelimbs (Fig. 5c), with expression of Fgf2, Fgf7, Fgf19 and Hgf in forelimbs at CS16 and CS17 (Fig. 5c). We also observed higher hindlimb expression for several
Figure 5 Gene signaling pathways predicted by IPA to be differentially activated in forelimb and hindlimb autopods across three bat developmental stages (CS15, CS16 and CS17). (a) The top ten canonical IPA-annotated pathways ranked by absolute IPA activation z scores for at least one developmental stage (* * * P < 0.001). The IPA z-score scale predicts the activation (or inhibition) state of the respective signaling pathways in bat forelimbs relative to hindlimbs. (b) Heat map showing the RNA-seq expression patterns of gene members of the EIF2 signaling pathway. The z-score scale represents a sample of the mean-subtracted average of the regularized log-transformed read counts in each sample. Genes that, when mutated, cause Diamond-Blackfan anemia are highlighted in bold, and others are indicated in dark blue genes known to be important regulators or markers of different stages of bone development, including mesenchymal condensation, chondrocyte differentiation, proliferation, maturation and hypertrophy. Genes that are established markers of a particular cell type are indicated in dark blue (

Fgf antagonists, including Spry2, Spry4 and FgfrI. Wnt ligands are secreted from the limb bud ectoderm and block cartilage formation in the periphery of the limb bud via the β-catenin pathway. We observed overall suppression of the canonical Wnt/β-catenin pathway in forelimb versus hindlimb, with higher levels of several canonical Wnt pathway antagonists in the forelimb and canonical Wnt receptors in the hindlimb (Fig. 5c), including Lef1, which showed strong CS15 hindlimb expression in whole-mount in situ hybridization (Supplementary Fig. 3). The Wnt–planar cell polarity (PCP) pathway has an important role in elongation of the limb along the proximal-distal axis and was activated in bat forelimbs at all stages (Fig. 5c). This upregulation included the PCP pathway ligand Wnt11, which has been shown to antagonize the Wnt/β-catenin pathway.

β-catenin signaling is known to suppress condensation of mesenchymal cells in endochondral bone development. To test our prediction that β-catenin signaling is diminished and leads to larger fields of condensing mesenchymal cells (Fig. 5d), we stained sagittal sections of bat forelimb and hindlimb autopods using peanut agglutinin (PNA), a galactose-specific lectin that binds to cell surface markers on condensing precartilage mesenchymal cells. Hematoxylin
and eosin staining of matched sections demarcated the progression from condensation of mesenchymal cells (CS15) to differentiation of chondrocytes (CS16) and progression to mature chondrocytes (CS17) in both forelimb and hindlimb autopods (Supplementary Fig. 8). At CS15, PNA staining was more intense, centered on the emerging digit IV, in sections of forelimb as compared to hindlimb autopods (Fig. 6). By CS16, all five digits were clearly visible in both bat forelimb and hindlimb sections (Fig. 6 and Supplementary Fig. 8). Whereas PNA staining was diminished as chondrocytes differentiated and matured, forelimb digits showed more intense, continued recruitment of condensing mesenchymal cells in the distal domain of digits II–V at both CS16 and CS17 (Fig. 6 and Supplementary Fig. 8). These data suggest that the timing and size of the initial digit condensations and subsequent recruitment of mesenchymal condensations are different in bat forelimb and hindlimb autopods from CS15 onward and that the foundation for the rapid elongation of forelimb digits could be established earlier than CS20, as previously proposed45.

In the limb, Bmp signaling regulates both bone formation and interdigit tissue regression46. We observed two distinct phases of Bmp signaling in our data sets (Fig. 5c). During digit initiation and specification (CS15), we observed high levels of the Bmp inhibitors Gremlin and Bmp3 in the hindlimb, which indicates a slight developmental lag at this stage, whereas the transcripts for Bmp receptors (Bmpr1a, Bmpr1b, Bmpr2 and Acrv1) and ligands (Bmp5 and Gdf5) were more abundant in the forelimb. Mutations in Bmp5 were shown to decrease mouse limb width47, and overexpression of GDF5 in chicken increases skeletal length46. The pattern of Bmp signaling started to switch at CS16, with CS17 forelimb showing higher levels of Bmp3 and Gremlin. Expression of these Bmp antagonists in the forelimb is consistent with the observed decrease in Msx1 and Msx2 expression. A similar suppression of the Bmp signaling pathway has been shown to have an important role in retention of interdigit webbing in duck47. Ranking genes from our differentially expressed signaling pathways for consistency across the RNA-seq and ChIP-seq data sets (Supplementary Table 9) found Mox2 (Bmp signaling) and Fzd10 (Wnt/β-catenin pathway) to be positively correlated for RNA-seq, H13K27ac and H3K27me3 signal. These genomic regions contain 8 and 12 BARs, respectively (within 500 kb of the transcription start site), suggesting that they could be important determinants of bat wing development.

DISCUSSION
To identify genetic components that might contribute to bat wing development, we carried out whole-genome sequencing combined with RNA-seq and ChIP-seq for H3K27ac and H3K27me3 on developing bat forelimbs and hindlimbs at three key developmental time points. Overall, we found that multiple genetic components are likely to contribute to development of the bat wing. These include numerous gene expression changes, both in known limb developmental regulators and newly characterized ones, such as Mllt3 and Lhx8. IncRNAs could also have a strong influence on wing development, with observed forelimb-hindlimb expression differences for Hottip and Tbx5-as1, an uncharacterized IncRNA. Combined pathway analysis highlighted numerous signaling pathways that seem to be differentially activated. These include ribosomal proteins, whose alteration has been shown to result in limb malformations35. Suppression of the Wnt/β-catenin pathway in the forelimb is consistent with condensation of larger fields of digit mesenchymal cells in the developing bat wing. In contrast, Wnt-PCP signaling, which maintains the polarity of proliferating chondrocytes in the growth plate, was more active in forelimb (Fig. 6d) and could potentially set the foundation for extended digit growth. Interestingly, the Bmp signaling pathway showed two distinct phases, with the inhibitors Gremlin and Bmp3 expressed at high levels early in the hindlimb and at later stages in the forelimb, with different tissue and temporal identity for Bmp activators, fitting with the diverse roles of this pathway in chondrogenesis, osteogenesis and apoptosis. Combined, the differential activation of these pathways is consistent with changes in expression of key genes in long bone development, including enhanced expression of chondrogenic markers (for example, Sox6, Aggrecan and Mmp9) across CS15–CS17 (Fig. 6d). These expression changes could be driven by gene regulatory elements with potential candidate sequences residing in our ChIP-seq data sets.

Our study obtained unique genomic data from wild-caught non-model organisms. Although restricted sample sizes, biological variation and gross tissue sampling may have reduced the scope of the experiments and the power of some of the analyses, we were able to generate robust genomic, transcriptomic and epigenomic data sets, which identified potential regulators of the processes involved in bat limb development. As bats are not currently amenable to transgenic experimentation, future functional characterization of the genes, IncRNAs and regulatory elements identified here could be performed in the mouse, with the potential to further understanding of their functional importance in the limb. Combined, our results uncover, on a genomic level, the molecular components and pathways that may have a key role in formation of the bat wing and provide a foundation for future studies examining such unique morphological innovations.
Accession codes. The whole-genome shotgun assembly is available under BioProject PRJNA283550. RNA-seq data are available from the Sequence Read Archive (SRA) under accession PRJNA270639 (SRP051253). ChIP-seq data are available from the SRA under accession PRJNA270665 (SRP051267).

Note: Any Supplementary Information and Source Data files are available in the online version of the paper.

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AUTHOR CONTRIBUTIONS

W.L.E., S.A.S., M.K.M., J.E.V., J.D.W., N.I. and N.A. conceived key aspects of the project and planned experiments. S.A.S., M.K.M., Z.G., A.V.P., C.M.-N. and N.I. collected bat embryos. S.A.S. and C.M.-N. extracted genomic DNA and RNA from bats. E.T., K.A.N. and L.C. generated genome sequencing libraries. W.L.E., S.A.S., M.K.M., Z.G., A.V.P., R.M.B., S.N., N.M., J.E.V. and N.I. performed experiments. T.F. and K.S.P. developed statistical methods. W.L.E., S.A.S., M.K.M., Z.G., T.F., L.C., K.S.P., J.D.W., N.I. and N.A. analyzed data. W.L.E., S.A.S., M.K.M., L.C. and N.I. wrote the manuscript. All authors commented on and revised the manuscript.

COMPETING FINANCIAL INTERESTS

The authors declare no competing financial interests.

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METHODS

Methods and any associated references are available in the online version of the paper.
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ONLINE METHODS
Genome assembly. DNA was extracted from the leg muscle tissue of a single male *M. natalensis* using phenol-chloroform. The 4-pg protocol of the Nextera Mate-Pair Sample Preparation kit (Illumina) was used to generate libraries with insert sizes of 2 kb, 5–6 kb and 8–10 kb. For the libraries with insert sizes of 5–6 kb and 8–10 kb, multiple reactions were pooled (four and seven, respectively) before size selection. The smaller-insert libraries were generated with the TruSeq DNA LT Sample Preparation kit (Illumina), following the manufacturer’s instructions. All libraries were sequenced on the Illumina HiSeq 2500 platform. The 175-bp and 300-bp paired reads were trimmed on either side to a minimum quality of 17 using Trimmomatic48. Trimmed reads were then used to calculate the 27-bp k-mer frequency using KmerFreq_HA in SOAPdenovo49. The 175-bp paired reads were then merged using their theoretical 25-bp overlaps and FLASH50. The remaining read pairs were trimmed to a minimum quality of 17 on the 3’ end before all reads were error corrected using Corrector_HA in SOAPdenovo. k-mers in a read with a frequency of 3 or lower were corrected to a more common k-mer. These changes were limited to two instances for the non-overlapping paired-end reads and four instances for the 175-bp reads. After these corrections, further erroneous k-mers were removed, to obtain a minimum read length of 60 bp. Duplicate reads were removed using PastUniq51. In combination, the reads totaled over 77x coverage, of which 17.5x was composed of long-insert mate pairs. Processed reads were assembled using SOAPdenovo49 and a k-mer size of 49. Pairs with one read mapping to a contig and one read mapping to a gap in a scaffold were used to fill in gaps with GapCloser (SOAPdenovo; submitted as whole-genome sequencing data under BioProject PRJNA283550). Heterozygosity was estimated using the Burrows-Wheeler Aligner (BWA)52 and SAMTools53. The coherency of the genomic sequence was tested with CEGMA54, using mammalian optimization.

**Genome annotation.** The *M. natalensis* genome was annotated using the Maker2 pipeline54. Repetitive regions comprised 33% of the genome and were soft-masked using RepeatMasker. Several transcriptome assemblies were used to annotate genes. These included draft assemblies of the *M. natalensis* forelimb and hindlimb RNA-seq data for each of the three time points (three assemblies) and a pooled assembly of all the RNA-seq data. Combined, these yielded 6.1 million transcripts that were aligned to the genome using BLAST55. In addition, 960,000 *M. brandtii* RNA-seq transcripts, from the liver, kidney and brain55, were aligned using relaxed BLASTN settings (75% coverage, 80% identity and an e-value cutoff of $5 \times 10^{-9}$), and 51,778 mouse proteins from the RefSeq protein database were aligned using BLASTP. After alignment, Exonerate56 was used to clear up intron-exon boundaries. All *initio* gene prediction with one read mapping to a contig and one read mapping to the earlier annotation, and AUGUSTUS58, which was run using human optimization. Once annotation was complete, gene predictions with poor evidence (annotation edit distance (AED) >0.75) were ignored. Finally, PASA59 was used to identify transcripts that did not have a match in the mammalian UniProt database were compared to sequences in InRNAdb v2.0 (ref. 22) and the GENCODE v7 InRNA gene annotation database60. These noncoding transcripts were also compared using BLAST25 to mouse, human, dog, horse, cat and other bat genomes to identify new InRNA transcripts that were conserved either in bats or in a subset of mammals. The Coding Potential Calculator was used to score whether a transcript was likely to be coding or noncoding61. For differential expression analysis, raw sequencing reads were mapped to the *M. natalensis* draft genome using TopHat64. Read counts for each gene were calculated for each replicate using HTSeq65, and tests of differential expression were carried out using DESeq2 (ref. 66). After differential expression testing, genes with P values adjusted for multiple testing (FDR) <0.01 in any of the five differential expression tests were clustered for similar expression using the R package hclust and displayed in a heat map. Additionally, genes found to be differentially expressed in forelimb and hindlimb across all stages were grouped on the basis of specific GO categories and subjected to analysis by clustered heat maps.

**In situ hybridization.** Mouse embryos (C57BL/6 strain UCT3) were supplied by the Animal Research Facility, University of Cape Town. *M. natalensis* embryos were collected from a maternity roost at the De Hoop Nature Reserve, South Africa (Cape Nature Conservation permit number AA007-00133-0056) as previously described62. Ethical approval by the University of Cape Town for sampling of bats was granted by the Science Faculty Animal Ethics committee (2012/V41/NI and 2012/V39/NI) and for the use of mouse embryos was granted by the Health Science Animal Ethics Committee (FHS AEC 014/07). Bat and mouse embryos from equivalent stages were matched as described by Hockman et al.10. Fixation and storage of embryos and whole-mount *in situ* hybridization probe synthesis and conditions were as previously described63. The primers used to generate the probes for whole-mount *in situ* hybridization are summarized in Supplementary Table 10.

**Mllt3 mouse skeletal preparations.** Skeletons of newborn Mllt3 homozygous knockouts64 and wild-type littermates were stained for cartilage with Alcian blue and for bone with Alizarin red as previously described65. Briefly, newborn mice were euthanized, skinned, eviscerated, fixed in 95% ethanol for several days and then incubated at 37 °C for 2 d in Alcian blue stain (15 mg of Alcian blue in 80 ml of 95% ethanol and 20 ml of glacial acetic acid). Samples were rinsed twice in 95% ethanol for 2 h each. Specimens were cleared in 1% potassium hydroxide for 4–5 h and counterstained overnight with Alizarin red stain (50 mg of Alizarin red in 1 l 2% potassium hydroxide). Finally, samples were cleared in 20% glycerol and 1% potassium hydroxide followed by 50% glycerol and 1% potassium hydroxide for several days for each step and then stored in 80% glycerol.

**ChiP-seq analysis.** Developing bat forelimbs and hindlimbs (dissected from CS15, CS16 and CS17 embryos) were cross-linked with 1% formaldehyde for 10 min. Reactions were quenched with glycerine, and samples were flash-frozen in the field. Cross-linked limbs were then combined in the laboratory into pools of 4–7 pairs per stage for chromatin sheering using a Covaris S2 sonicator. Sheared chromatin was used for immunoprecipitation with antibodies against active (H3K27ac; Abcam, ab4729) and repressive (H3K27me3; Millipore, 07-449) chromatin marks using the Diagenode LowCell# ChiP kit, following the manufacturer’s protocol. Libraries were prepared using the Rubicon ThruPLEX-FD Prep kit following the manufacturer’s protocol and sequenced on an Illumina HiSeq 2500 instrument using single-end 50-bp reads to a sequencing depth of at least 25 million reads (submitted to the SRA under accession SRP012679). Uniquely mapping raw reads were aligned using Bowtie69 with default settings. Peak regions for each histone mark were called using SICER70, informed by the estimated average fragment size of the chromatin after sheering, as measured by Agilent 2100 Bioanalyzer. Peaks from all samples were merged using BEDTools71 and partitioned using BEDOPS72 (Supplementary Fig. 10). Differentially enriched regions in forelimb and hindlimb for each stage and histone mark were obtained following a methodology similar to MAnorm, which uses a linear model that assumes that peaks shared by samples can serve to normalize ChiP-seq data sets for differing signal-to-noise ratios. In this methodology, genomic regions not appearing as a peak in any sample were obtained from the partitioned regions using
BEDTools and were used to normalize for the background noise present in each sample. Furthermore, a set of common regions for each histone mark was obtained, and these regions were used to normalize the ChiP-seq signal between all samples by creating a scaling factor based on the average signal in the shared peaks minus the average noise in non-peak regions. After removing duplicate reads with Picard MarkDuplicates, read counts were obtained with the BEDTools coverage command. The average noise from each region based on its genomic size was subtracted from each region’s read counts. Noise-subtracted read counts were then normalized by multiplying each by the signal scaling factor and a read depth scaling factor, to create an enrichment score for each portioned region. Pairwise differential enrichment tests were carried out using a Bayesian model \(^7\), which is also used by MANorm, followed by adjustment for multiple testing using the \(R\) package \texttt{p.adjust}.

Comparative genomics. Whole-genome alignments were carried out using LASTZ \(^7\) with soft-masked genome assemblies from 18 species (\textit{E. fuscus, M. brandti, M. davidi, M. lucifugus, P. alpha, V. vampyrus}, \textit{Bos taurus, Canis familiaris, Equus caballus, Felis catus, Homo sapiens, Loxodonta africana, Monodelphis domestica, Mus musculus, Sus scrofa, Danio rerio, Arolits carolinensis and Gallus gallus}) using the repeat-masked \textit{M. natalensis} genome as a reference. If no publically available repeat-masked genome was available, RepeatMasker was run using the mammal repeat database and default conditions. Alignment files were then chained, netted and converted to MAF files using UCSC Genome Browser utilities. Individual MAF files from each pairwise species alignment were combined into a multiple MAF file using the roast command, which is part of the Multiz-TBA package \(^7\). Tree models for both conserved and non-conserved sequences were created for the species used in the multiple MAF file by phyloFit \(^7\). These tree models were used in phastCons \(^7\) to alignment were combined into a multiple MAF file using the roast command, files were then chained, netted and converted to MAF files using UCSC Genome Browser utilities. Individual MAF files from each pairwise species alignment were combined into a multiple MAF file using the roast command, which is part of the Multiz-TBA package \(^7\). Tree models for both conserved and non-conserved sequences were created for the species used in the multiple MAF file by phyloFit \(^7\). These tree models were used in phastCons \(^7\) to identify vertebrate conserved sequences in the \textit{M. natalensis} genome and to generate base-by-base conservation scores to be displayed in genome browsers. BARs were identified by using phylOp \(^27,76\) to test for acceleration in the common ancestor of the bat lineage over regions identified as vertebrate conserved sequences after filtering for quality alignments. Genomic regions enriched for BARs were identified by scanning the genome using a 100-kb sliding window with a step size of 50 kb while counting BARs and phyloP-tested regions within them. On average, phyloP found acceleration in 0.812\% of the sequences tested. The expected number of BARs in each region was then set to be the number of sequences tested by phyloP in that region multiplied by 0.00812. Regions enriched for BARs were identified by comparing the average expected number of BARs to the observed number of BARs using a Poisson test. After correction for multiple testing, genes contained in or overlapping the genomic regions with significant over-representation of BARs were analyzed for functional annotation clustering using DAVID \(^7\,7\) with the background set to the genes contained in regions with valid multiple-sequence alignments and H3K27ac peaks.

Ingenuity Pathway Analysis. Pairwise differential expression testing for forelimb and hindlimb at each stage identified a total of 3,140 bat genes (FDR < 0.05). This list was filtered for genes that had an average FPKM value >2 and that had been mapped to a human Entrez Gene ID, resulting in 2,751 genes. IPA (Qiagen) was used to analyze this set of 2,751 genes to determine whether specific canonical signaling pathways and their upstream regulators were coordinately regulated across three developmental stages (CS15, CS16 and CS17) using fold change values. A Fisher’s exact right-tailed test identified significantly enriched pathways, and a \(z\) score was computed to determine whether the pathway was activated or inhibited at each stage. IPA was also used to predict upstream regulators that would explain the patterns of differential gene expression observed across the data set. Coherency (marked by 1) was tested by comparing significant differences between forelimb and hindlimb ChiP-seq and RNA-seq signals for genes that were differentially expressed in the top ten canonical IPA pathways (Fig. 5a). Significantly different acetylation marks were required to be antagonistic to their equivalent methylacylation marks, with at least a single mark being significantly different in the forelimb and hindlimb. RNA-seq levels in the forelimb and hindlimb were also required to be positively correlated with any significantly different acetylation marks.

Histochemistry. Bat embryos were fixed in 4% paraformaldehyde for 3 h at room temperature, washed in PBS and stored in 30% sucrose in PBS at 4 °C for 5–6 d. Whole limbs were dissected from these embryos and embedded in tissue freezing medium (Leica Biosystems). These were sectioned to be 8 \(\mu\)m thick, using a Leica CM1850 cryotome at –17 °C, collected on Superfrost Plus (Thermo Scientific) slides and stored at –70 °C. Serial sections were stained with either hematoxylin and eosin or PNA. Slides containing sections were fixed in phosphate-buffered formalin for 5 min, washed in distilled water for 1 min and then stained with hematoxylin for 30 s. Slides were rinsed in 0.3% acetic acid (70% ethanol) and running water and incubated in Scott’s water for 1 min. After rinses in running water and 80% alcohol, slides were stained with acid-based eosin for 2.5 min. Slides were dehydrated through an ethanol series and dipped in xylene, and coverslips were secured with DPX mountant (Sigma).

For PNA staining, bat autopod sections were fixed for 10 min in acetone. Slides were washed three times in PBS and blocked in 3% BSA in PBS for 1 h at room temperature. Sections were incubated with 100 \(\mu\)g/ml FITC-conjugated PNA (Sigma, L7381) in 3% BSA in PBS at 4 °C overnight. Control slides were incubated in 3% BSA in PBS only. All slides were washed in PBS and stained for 10 min in 1 \(\mu\)g/ml Hoechst nuclear stain, before another three PBS washes. ProLong Gold anti-fade reagent (Life Technologies) was used to mount coverslips. Sections were photographed on a Nikon Ti-E inverted fluorescent microscope using the same standardized camera setting for all sections.

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