Single-cell mRNA profiling reveals heterogeneous combinatorial expression of *Hoxd* genes during limb development

**Short title:** Single-cell Hox combinations in developing limbs

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**HIGHLIGHTS**

- Collinear expression of *Hox* genes is only weaved at the tissue scale
- Enhancer-sharing to specific target genes is reduced at the single-cell level
- *Hoxd* gene combinatorial expression is linked to distinct transcriptional signatures
- In presumptive digits, *Hoxd* combinations follow a pseudotime trajectory
ABSTRACT
A global analysis of gene expression during development reveals specific transcription patterns associated with the emergence of various cell types, tissues and organs. These heterogeneous patterns are instrumental to ensure the proper formation of the different parts of our body, as shown by the phenotypic effects generated by functional genetic approaches. However, variations at the cellular level can be observed within each structure or organ. In the developing mammalian limbs, expression of Hoxd genes is differentially controlled in space and time in cells that will pattern the digits and the arms. Here we analyze single-cell transcriptomes of limb bud cells and show that Hox genes are expressed in specific combinations that match particular cell types. In the presumptive digits, we find that the expression of Hoxd gene is unbalanced, despite their common genomic proximity to known global enhancers, often expressing only a subset of the five genes transcribed in these cells. We also report that combinatorial expression follows a pseudo-time sequence, suggesting that a progression in combinatorial expression may be associated with cellular diversity in developing digits.

INTRODUCTION
Limb morphogenesis is controlled by several key transcription factors, amongst them members of the Hox gene family, in particular genes from the HoxA and HoxD clusters. During early limb development, the posterior Hoxd genes are expressed in precise, partly overlapping domains (Dolle, 1989a #13; Dolle, 1993 #14), which will pre-figure the various parts of the future appendices, i.e. the hands and feet (autopods) and the more proximally located arm (stylopod) and forearm (zeugopod) segments. Recently, it was shown that expression of Hoxd9 to Hoxd13 in presumptive digits is under the control of the same set of enhancer elements, located in the gene desert centromeric to the cluster itself (Montavon, 2011 #4; Andrey, 2013 #2; Fabre, 2017 #9). However, their global expression patterns display some difference, with a broader expression of Hoxd13 within future digit 1 (the thumb), whereas Hoxd9 to Hoxd12 transcripts were found only in future digits 2 to 5. This difference is likely due to the existence of a quantitative collinearity (Dolle, 1991 #9; Spitz, 2003 #16; Montavon, 2008 #4), whereby a gradual increase in the amount of steady-state mRNA levels is observed from Hoxd9, expressed at the weakest level, to the robust transcription of Hoxd13.
By using DNA FISH, the extent of chromatin interactions between Hox genes and their enhancers in single cells showed variability (Fabre, 2017 #9; Rodriguez-Carballo, 2017 #8), and super-resolution microscopy confirmed that the HoxD gene cluster can display a variety of structural conformations in various future autopod cells (Fabre, 2015 #2). This heterogeneity is hard to reconcile with chromosome conformation datasets produced at this locus (e.g. Montavon, 2011 #5; Fabre, 2015 #32), since the latter approach reflects the averaged behaviors of a cellular population. Consequently, a higher variability can be expected in cell-specific Hox gene transcriptions, when compared to previously reported expression profiles (Duboule, 1989 #33; Dolle, 1989 #15; Montavon, 2008 #4).

While genetic approaches have revealed the critical function of these genes during limb outgrowth and patterning, the homogeneous or heterogeneous impact of mutations at the cellular level is more difficult to evaluate. The ablation of Hoxd13 alone leads to a morphological effect in digits weaker than when a simultaneous deletion of Hoxd11, Hoxd12 and Hoxd13 is achieved (Dolle, 1989a #13; Zakany, 1997a #12; Delpretti, 2012 #17), suggesting that Hoxd11, Hoxd12 and Hoxd13 functionally cooperate during digit development. However, how this cooperation occurs at the cellular level is unknown. This question is made even more complex by the regulatory strategies that evolved at the HoxD locus, where several neighbor genes can be regulated by several enhancers in the same large domains.

It was indeed recently reported that this cluster lies between two large topologically associating domain (TADs) (Dixon, 2012 #34; Andrey, 2013 #1; Beccari, 2016 #35; Fabre, 2015 #32), each of them containing a range of enhancer elements acting in the same domains. The TAD located centromeric to HoxD (C-DOM) contains several enhancers specific for autopod (digit) cells, whereas T-DOM, the TAD located telomeric to HoxD, hosts a series of enhancers specific for future arm and forearm cells (Andrey, 2013 #1). In addition, genes with a central position in the cluster such as Hoxd9, Hoxd10 or Hoxd11 are targeted by enhancers belonging to the two different TADs, initially in zeugopod cells, then in autopod cells, suggesting an even greater heterogeneity in transcripts distribution. In order to try and evaluate Hoxd transcript heterogeneity during limb development, we produced single-limb cell transcriptomes of different origins, to see whether the apparently homogenous behavior in Hox gene transcriptional program as observed upon large-scale analyses was confirmed at the cellular level. We report here that Hoxd genes transcripts are present in various combinations in different limb cells and discuss the importance of these results in our understanding of how Hoxd genes are regulated and how their global functions are achieved in these structures.
RESULTS

Heterogeneity of posterior Hoxd gene transcripts in single-cells

In order to document the expression pattern of Hoxd13 at the single-cell level, embryonic day (E) 12.5 limb sections were use in RNA-FISH experiments (Fig. 1A). As expected, we observed a high expression specificity in presumptive digits cells in the distal part of the forelimb, with the highest transcript levels in cells located at the boundary between the digital and the interdigital compartments, while lower levels were scored in interdigital mesenchyme. Signal was neither detected within the digital compartment, nor in more proximal parts of the limb ([Montavon, 2008 #4]) (Fig. 1A). However, a high heterogeneity in gene expression was recorded, with stippled signal pattern contrasting with the broader expression domain previously described. As a consequence, we asked whether all cells expressing Hoxd13 would also contain Hoxd11 transcripts, given that both genes are under the same regulatory control in these distal cells ([Montavon, 2011 #5]{Andrey, 2013 #1}). We micro-dissected autopod tissue to obtain a single-cell suspension and performed double fluorescent RNA labelling. The single-cell preparation was then analyzed by FACS and revealed that only a minority of cells were in fact expressing detectable levels of Hoxd11 and/or Hoxd13 (Fig. 1B). Amongst positive cells, the largest fraction was Hoxd13 positive and negative for Hoxd11 (d13+d11−; 53%), whereas double positive cells (d13+d11+) represented 38% only and 9% of the cells contained Hoxd11 mRNAs alone (d11+) (Fig. 1B).

Because a substantial number of cells did not express any Hoxd genes, we enriched for the positive fraction using a mouse line containing a GFP reporter sequence knocked in Hoxd11. In these mice, GFP was produced in those cells where Hoxd11 had been transcribed (Fig. S1). We monitored the fluorescence at E12.5 and observed a pattern recapitulating Hoxd11 endogenous expression ([Dolle, 1989a #15; Dolle, 1989b #13]) (Fig. 1C). E12.5 limb cells from these animals were FACS sorted using the GFP (Fig. 1D) and, under these conditions, the double labelling of GFP positive cells increased to more than a third of the cells (Fig. 1E). However, amongst the positive cells, the ratio between the three Hoxd positive populations (Hoxd13 only, Hoxd11 only and double-positive) was roughly the same as before (37%, 7% and 55%, respectively). To confirm the presence of these different populations, we performed Hoxd13 RNA-FISH on sections from Hoxd11::GFP E12.5 forelimbs (Fig. 1F) and observed a high variability in GFP levels (Fig. 1G). We found that high levels of Hoxd13 were observed in cells with either little or no Hoxd11 activity (Fig. 1H), yet the majority of cells displayed high signals...
for both *Hoxd11* and *Hoxd13*, suggesting that in these cells the two genes were regulated in a similar manner.

To quantify a potential correlation between *Hoxd11* and *Hoxd13* expression levels in these GFP-positive cells, we binned *Hoxd11* positive cells in three categories: negative cells (*d11*\textsubscript{neg}, orange), cells expressing at low levels (*d11*\textsubscript{low}, red) and cells expressing at high levels (*d11*\textsubscript{hi}, grey; **Fig. 1I**, left panel). Flow cytometry analysis revealed that higher *Hoxd13* levels were clearly observed in the *d11*\textsubscript{hi} population, indicating that in single cells, whenever both genes are expressed, they tend to respond to enhancers with the same efficiency (**Fig. 1I**). To relate these latter results with the level of GFP observed by microscopy (**Fig. 1H**), we monitored the levels of GFP in single-cells and found a correlation between abundant *Hoxd11* mRNAs and higher levels of the GFP protein (**Fig. 1I**, right panel). Altogether, these results suggested that some cellular heterogeneity exists with respect to *Hoxd* gene transcription in presumptive digit cells, with the possibility for sub-populations of cells to selectively express either one or the two genes. Overall, these observations contrast with the view that all limb cells transcribe all posterior *Hoxd* genes, a view conveyed by the global analysis of expression patterns by whole-mount in situ hybridization (WISH).

**Single limb cells transcriptomics**

To obtain a wider view of this cellular heterogeneity as well as to see whether it depends on the position and fate of various limb cells, we performed single-cell RNA-seq to expand the analysis to all *Hox* genes. Because of its potential to detect as little as single-digit input spike-in molecules, we used the Fluidigm microfluidics C1 captures to obtain the maximal intensity of transcript detection ([Svensson, 2017 #36]). To enrich for cells expressing at least one *Hoxd* gene, we used only the GFP positive cells sorted by flow cytometry from the *Hoxd11::GFP* mouse E12.5 forelimbs (see **Fig. 1C-E**). After capture, the cells were sequenced at very high depth to reach the finest sensitivity of gene detection, with an average of about 8.7 M reads per cell (**Fig. S2**).

We first showed that autopod and zeugopod cells portray distinct transcriptional signatures that can be observed in a machine learning algorithm that reduces dimensionality (t-SNE). In this plot representation, we saw only little intermingling between autopod and zeugopod cells (**Fig. 2A-B**). To ensure that the single-cells signatures were specific to the two populations, we performed a differential expression analysis between the distal and proximal limbs. As shown in the MA-plot we found that genes specific to one or the other populations
were indeed known markers of the two tissues (Fig. 2C). In fact, most of the autopod-specific genes are part of a tight interactive network established through weighted aggregation of known interactions (Figs. 2D and S3), thus demonstrating the high level of gene detection in our single-cells (Figs. 2D and S2).

To visualize the relative mRNA contributions of all Hoxd genes, we plotted their cumulative expressions with color coded single-cell (Figs. 2E and S4). While the distribution of absolute levels mirrored quite well the pattern previously established using other approaches ([Fabre, 2017 #9; Montavon, 2008 #4; Montavon, 2011 #5]), we observed again a selectivity of expression, which also applied to Hoxd12, Hoxd11 and Hoxd10. Of note, amongst autopod cells positive either for Hoxd13 and/or for Hoxd11, we identified similar proportions as before, with the majority of cells expressing both Hoxd11 and Hoxd13, 40 percent containing Hoxd13 mRNAs only and 10 percent with only Hoxd11 mRNAs. To assess the potential covariances between the five posterior Hoxd genes (from Hoxd9 to Hoxd13), we classified by Spearman’s rank correlation the genes that covaried with at least one of the Hoxd genes. A hierarchical clustering from these 76 genes showed a clear segregation between Hoxd11/Hoxd13 on the one hand, and Hoxd9, Hoxd10 and Hoxd12, on the other hand (Fig. 2F). While Hoxd9, Hoxd10 and Hoxd12 were closely associated in the presence or absence of their mRNAs, Hoxd11 and Hoxd13 were part of two different sub-clusters associated with different set of genes, suggesting that the cell-specific expression of combinations of Hoxd genes may have biological relevance.

**Combinatorial Hoxd genes expression observed in single-cells**

Therefore, despite their shared tissue-specific regulatory landscapes, all Hoxd genes are not systematically expressed by the same cells. A discretization of the expression levels allowed us to score the various mRNA combinations observed either in autopod (Fig. 3A), or in zeugopod (Fig. S5) single-cells. In the autopod, the largest population was composed of cells expressing Hoxd13 only, followed by a population expressing both Hoxd11 and Hoxd13 and then by an unexpected pool of cells with only Hoxd10 and Hoxd13 mRNAs. Cells containing three or more distinct Hoxd mRNAs were a minority and only 11 percent of cells expressed four genes, from Hoxd10 to Hoxd13. We asked whether these unambiguous associations were random or coupled with specific gene signatures by performing a tSNE on all autopod and zeugopod cells and we observed that groups of cells containing different combinations of Hoxd mRNAs tend to segregate, suggesting that their differences in gene expression is not restricted to Hoxd genes only (Fig. 3C).
We performed separate tSNE for autopod and zeugopod cells by clustering cells according to their Hoxd combinatorial patterns (Fig. 3D) and observed that some combinations tend to cluster together. This effect was particularly clear in autopod cells whenever a sufficient number of cells (>5) was plotted and we noticed that the transcriptional diversity increased along the second dimension of the tSNE when a higher diversity of Hoxd mRNAs was scored in the same cells. In zeugopod cells, groups of cells also segregated, though not as distinctly, suggesting a more homogeneous distribution of Hoxd mRNAs. These results suggested that subpopulation of autopod cells transcribe various combinations of Hoxd genes.

**Analysis of Hoxd cellular clusters**

To more precisely assess this apparent cellular selectivity in Hoxd gene expression, we first determined whether particular cell clusters were at a specific phase of the cell cycle. While most cells with G2 scores were observed with either Hoxd13 mRNAs only or with four posterior Hoxd genes active, we did not detect any significant difference associated with a specific combination of mRNAs (Fig. S5). We next performed a differential gene expression analysis to assess the degree of relationship between the six main cellular groups (Fig. 4A-C). Most of the differentially expressed genes (343 genes, Table S1, and Fig. S6) were scored between cells expressing only Hoxd13 and cells expressing either three (Hoxd11-Hoxd13), or four genes (Hoxd10 to Hoxd13) (Fig. 4A). Amongst these differentially expressed genes, many displayed strong autopod expression, including Jag1, which is downregulated in the absence of the HOX13 proteins (\{Sheth, 2016 #18\}). Out of 31 genes differentially expressed between cells containing either Hoxd13 and Hoxd11 mRNAs, or Hoxd10, Hoxd11, Hoxd12 and Hoxd13 mRNAs, only eight were specific to these two combinations (Smarcc1, Mrps17, Snrpd2, Supt6, Tax1bp1, Rab5c, Ncbp2, Map3k7).

Noteworthy, clustering of expressed transcripts showed a hierarchical organization with a progression from those cells expressing Hoxd13 only to two, then three and finally four Hoxd genes (Fig. 4C). As some of these genes were previously identified either as HOX proteins targets (e.g. Ppp2ca {Salsi, 2008 #37}), or being part of a Hox functional pathways (e.g. Uty, Hoxa11os), we assessed whether specific targets genes could be associated with particular combinations of Hoxd mRNAs. We generated a supervised clustering showing the covariance of known targets genes in a spearman correlation matrix (Fig. 4D, E). When the 199 cells originating from both the autopod and the zeugopod were considered, we found a clear partition of target gene mRNAs into two groups, corresponding to the nature of Hoxd mRNAs present during the cell cycle.
The presence of *Hoxd9* and *Hoxd10* mRNAs aggregated with targets genes such as *Hand2* and *Sfrp1*, whereas *Hoxd11*, *Hoxd12* and *Hoxd13* were co-expressed with different target genes such as *Ppp2ca* and *Bmp2/4*. Finally, the highest clustering across all cells was observed between *Hoxd12*, *Hoxd13*, *Dach1* and *Lhx9*, thus revealing a robust link between these genes.

When only autopod cells were considered, we observed two groups, with *Hoxd9* and *Hoxd10* transcripts in one cluster, while the more centromeric genes *Hoxd11*, *Hoxd12* and *Hoxd13* were transcribed in the other (Fig. 4E). As the former group did not express any of those genes typically up-regulated in distal cells (*Hoxd12* and *Hoxd13*), we wondered whether such differences in transcript distribution may reflect various stages in the progression of distal limb cells towards their final fates. Therefore, we implemented a measure of cellular pseudo-age, a strategy that evaluates a temporal hierarchy amongst single-cells based on their respective transcriptomes ([Haghverdi, 2016 #19]; [Trapnell, 2014 #20]; [Haghverdi, 2015 #49]). This approach allows to plot cells along a linearized axis to infer whether the combination alignments observed in the tSNE may correlate with a modulation of the time component.

We performed a pseudo-time analysis on the single-cells isolated from both the autopod and zeugopod and found that cells indeed spread along the pseudo-temporal axis that was linearized through a diffusion map (Fig. 5A-B). In these maps, while zeugopod cells did not distribute well along a temporal frame (Fig. 5B), the autopod cells were much better aligned (Fig. 5A). As demonstrated with gene expression clustering (Fig. 4. A-C), specific combinations are distributed along the temporal axis in a way related to the various combinations of *Hoxd* mRNAs, with the *Hoxd13*-only cells at one extremity of the axis and the *Hoxd10 to Hoxd13* combination at the other extremity (Fig. 5C-D). Altogether, this clustering analysis showed that different combinations of *Hoxd* gene mRNAs may affect distinct groups of target genes. Noteworthy, it also revealed a preference for mRNA combinations involving neighbor genes, thus emphasizing the importance of genes’ position for their co-regulation.

**DISCUSSION**

Limb bud cells require the expression of *Hox* genes originating from two separate clusters, *HoxA* and *HoxD*. We describe here the single-cell combinatorial expression of *Hoxd* genes found in cells sorted out by using a *Hoxd11::GFP* mouse strain. Albeit some cells tend to show higher level of *Hoxa* genes when the *Hoxd* genes were low, this was not the general rule. The fact that we did not score many *Hoxa*-mRNA positive cells after the enrichment for *Hoxd* genes expression (Fig. S4) may however reflect a compensatory mechanism whereby a strong
global expression of one cluster would result in the weak transcription of the other. Distinct cellular content for either Hoxd or Hoxa mRNAs could account for the different phenotypic effects of inactivating these genes upon limb morphology, as exemplified by Hoxd13 and Hoxa13 ({Fromental-Ramain, 1996 #28}{Zakany, 1997 #30}{Kmita, 2005 #27};{Beccari, 2016 #35}). A more comprehensive single-limb cell sequencing strategy will fix this issue.

Our data show that Hoxd quantitative collinearity (Dollé et al., 1991) is to be considered at the global level since it results in fact from a sum of combinatorial expression of various genes in different cells. We emphasized that in autopod cells, the most frequently expressed gene is Hoxd13, as was expected from previous studies where it was described that this gene is expressed the strongest {Montavon, 2008 #4}. Apart from Hoxd13, other Hoxd genes were more sparsely activated, indicating either a stochastic process or a functional requirement for specific mRNA combinations in different cell types. This heterogeneous cellular situation raises two separate questions regarding first the underlying regulatory mechanism and, secondly, the potential functional significance.

**Different regulatory conformations?**

We had previously shown that the regulation of posterior Hoxd genes in the distal limb bud was not entirely the same for all genes. In particular, Hoxd13 is the only gene to be expressed in presumptive thumb cells, the other Hoxd mRNAs being excluding from this very digit ({Montavon, 2008 #4}). Also, the deletion of the Hoxd13 locus lead to the upregulation of Hoxd12 in thumb cells, yet not of the other remaining genes, suggesting that this thumb-specific expression was associated with the final and most 5’ position of the gene on the cluster ({Kmita, 2002 #21}). The recent identification, in the posterior part of the HoxD cluster, of an unusually high density of bound CTCF molecules ({Soshnikova, 2010 #22}) may cause this transcriptional selectivity through the use of various sites, taking advantage of their distinct orientations ({Fabre, 2017 #9}{Rodriguez-Carballo, 2017 #8}).

In this view, the peculiar orientations of CTCF binding sites may allow the transient stabilization of various loop conformations, for example after extrusion driven by the cohesin complex ({Fudenberg, 2016 #44} {Haarhuis, 2017 #50}). Accordingly, distinct combinations of posterior Hoxd mRNAs could reflect the formation of specific loop extrusion patterns, in any single cell, as a choice between a fixed number of possibilities as determined by the presence of bound CTCF, with some conformations being favored over others. Of note we found that the cohesin-loading factor Nipbl is strongly downregulated in cells from the Hoxd13 group when
compared to cells expressing the full combination (\( \text{Hoxd10 to Hoxd13} \)). Mutations in this gene has been found in patients with Cornelia de Lange syndrome that have notably a clinodactyly of the 5th finger. Recently, a report showed that mice heterozygotes for \( \text{Nipbl} \) display polydactyly, and that lower dose of \( \text{Hoxd11 to Hoxd13} \) in these mice further enhance this phenotype {Lopez-Burks, 2016 #45}. Other chromatin regulators were found enriched in this list of genes including \( \text{Jag1, Brd7, Jmjd6, Phf8, Ddb1, Hdac1, Swi5, Smarcc1, Smarce1, Hmgb3, Dnm3os, Cbx1 and Lmnbl} \) (Table S1). The product of the latter gene has been associated with the architecture of a large domain of inactive chromatin (LADs, Kind, 2013 #25), where the \( \text{HoxD} \) cluster is not located (Vieux-Rocha et al PNAS 2015). Since reduced levels of \( \text{Lmnbl} \) gene product have been shown to be associated with reduced expression of polycomb targets including posterior \( \text{Hox} \) genes {Sadaie, 2013 #46}, its increased expression in the cells containing mRNAs from \( \text{Hoxd10 to Hoxd13} \) may reflect a global change in chromatin configuration ({Reddy, 2008 #23};{Peric-Hupkes, 2010 #24};{Kind, 2013 #25};{Gonzalez-Sandoval, 2016 #26}). How would this change relate to a more permissive expression of \( \text{Hoxd} \) genes, as a cause or as a consequence, remains to be established.

The analyses of single cell transcriptomes revealed an unexpected hierarchical progression of \( \text{Hoxd} \) genes expression, from cells expressing a single posterior gene (\( \text{Hoxd13} \)) to the full combination (\( \text{Hoxd10 to Hoxd13} \)). This global transcriptional sequence was inferred from a pseudo-time approach, a method whereby a temporal progression of cells is deduced based on their transcript patterns ({Haghverdi, 2016 #19};{Trapnell, 2014 #20}). We tested this hypothesis using diffusion pseudotime and found that autopod cells are much more subject to align along a developmental trajectory. This specificity may be link to the particular way \( \text{Hoxd} \) genes are regulated in distal limb buds, with a rapid and strong activation of \( \text{Hoxd13} \) due to its leading position in the cluster, towards the various enhancers ({Montavon, 2008 #4}). It is possible that the recruitment of additional \( \text{Hoxd} \) genes located nearby may be progressive, along with local epigenetic modifications, which could be inherited from one cell to its daughter cells. In this view, the number of \( \text{Hoxd} \) genes expressed would increase along with mitotic divisions leading to the hierarchical progression observed.

**Additive cellular or emerging functions?**

The second question relates to the potential different functions that limb bud cells may display by carrying distinct combinations of \( \text{Hoxd} \) mRNAs. The question here is to discriminate between two views of the genotype-phenotype relationship during limb bud development,
between a situation where each cell would express a determined combination of *Hoxd* mRNAs, for example in response to its topological position within the growing limb or to its own "regulatory history", i.e. the regulations at work in its ancestor cells or, alternatively, a stochastic distribution of chromatin architectures leading to a globally balanced distribution of cells expressing various *Hoxd* mRNAs. In the former context, the resulting limb phenotype would derive from the additive effect of every single cells providing one out of the possible sets of information delivered by the various transcriptomes associated. In the second framework, the phenotype would derive from the random mixture of multiple cells expressing distinct transcriptomes with a given balance fixed by the choice of possible chromatin conformations.

Genetic approaches cannot easily discriminate between these alternatives. In previous studies where the functions of *Hoxd* genes during limb development were aimed to be assessed separately, various combinations of multiple inactivations were used. In most cases however, this consistently led to limited phenotypes due to a fair level of redundancy, particularly amongst *Hoxd* and *Hoxa* genes, preventing precise functions to be attributed to specific (groups of) *Hox* genes ([Zakany, 1997 #30]). However, the use of multiple gene inactivations has revealed that the transcription of *Hoxd11* and *Hoxd12* contributed functionally and thus added to the mere presence of *Hoxd13* transcripts, even though autopods double mutant for *Hoxd13* and *Hoxa13* would no longer grow ([Kmita, 2002 #21]; [Delpretti, 2012 #17]; [Fromental-Ramain, 1996 #28]). This is coherent with our data suggesting that the specific presence of *Hoxd11* or *Hoxd12* mRNAs is associated with distinct transcriptomes containing additional key regulators of cell fate and chromatin remodeling genes.

Therefore, part of the limb phenotypes observed in *Hoxd* multiple mutants may result from the different response of a sub group of cells, which would be differentially impacted by the loss of a given gene. For example, cells that express only *Hoxd13* or a combination of *Hoxd13* and *Hoxd10* mRNAs may be sensitive to the absence of *Hoxd11* transcripts in the corresponding mutant stock. Our results thus stress the necessity to keep in mind the cellular heterogeneity of transcriptional programs even in instances where WISH patterns seem to reveal homogenous distributions of transcripts. In this context, transcript patterns at the single-cell level can help solve the interpretation of genetically-deficient phenotypes, even though the co-regulation of *Hoxd* genes and the functional redundancy of their products make this statement difficult to apply to the present work.

Collectively, our observations revealed the existence of distinct combinations of *Hoxd* genes at the single-cell level during limb development. In addition, we document that the
increasing combinatorial expression of Hoxd genes in this tissue is associated with specific transcriptional signatures and that these signatures illustrate a time progression in the differentiation of these cells. Further analysis at different developmental stages may enable the reconstruction of the cell fate trajectories and the state transitions that causes the cellular heterogeneity of the early limb bud tissue.

**METHODS**

**Animal experimentation**

All experiments were performed in agreement with the Swiss law on animal protection (LPA), under license No GE 81/14 (to DD). Forelimbs tissue samples were isolated from Hoxd11::GFP heterozygous animals at embryonic day 12.5 (E12.5) with day E0.5 being noon on the day of the vaginal plug. The cloning steps for the generation of the Hoxd11 transgenic mice is described in Fig. S1A-C. Briefly, the knock-in was done by introducing a bi-cistronic cassette along with an IRES sequence. Hoxd11 was inactivated by the insertion of a TauGFP sequence in frame into the coding sequence (Fig. S1C). The BamH1 site was used for insertion of the IRES cassette (Fig. S1A). Fig. S1D shows how the cassette was introduced as a single-copy knock-in. The GFP signal observed in this mouse stock reflects the endogenous distribution of Hoxd11 transcription.

**RNA-FISH**

E12.5 forelimbs were micro-dissected and fixed with 4% paraformaldehyde for 3 hours. Then the limbs were treated with sucrose at 5, 10 and 15% and then frozen in OCT. 25 μm cryostat sections were dried for 30 minutes, post-fixed in 4% paraformaldehyde for 10 minutes and quenched with 0.6% H2O2 in methanol for 20 minutes. Slides were then processed using the Ventana Discovery xT with the RiboMap kit. The pretreatment was performed with mild heating in CC2 for 12 minutes, followed by protease3 (Ventana, Roche) for 20 minutes at room temperature. Finally, the sections were hybridized using automated system (Ventana) with a Hoxd13 probe diluted 1:1000 in ribo hyde at 64°C for 6 hours. Three washes of 8 minutes in 2X SSC followed at hybridization temperature (64°C). Slides were incubated with anti-DIG POD (Roche Diagnostics) for 1 hour at 37°C in BSA 1% followed by a 10 minutes revelation with TSA substrate (Perkin Elmer) and 10 minutes DAPI. Slides were mounted in ProLong fluorogold. Images were acquired using a B/W CCD ORCA ER B7W Hamamatsu camera.
associated with an inverted Olympus IX81 microscope. The image stacks with a 2 μm step were saved as TIFF stacks. Image reconstruction and deconvolution were performed using FIJI (NIH, ImageJ v1.47q) and Huygens Remote Manager (Scientific Volume Imaging, version 3.0.3).

RNA flow cytometry

Double in situ hybridization in single cells for RNA flow cytometry was performed using PrimeFlow RNA (Affymetrix, Santa Clara, CA) reagents following the manufacturer’s protocols. Cell viability was assessed by live/dead fixable dead cell; violet (ThermoFischer; L34955). Hsp90ab RNA probe (a gene expressed ubiquitously) served as a positive control. Hoxd11 and Hoxd13 RNA probes were used for the actual analysis. Cell staining was analyzed on a FACS Astrios located at the EPFL flow cytometry platform. Data analysis was performed by using FlowJoX (Treestar, Ashland, OR). The labelling and flow cytometry were performed on dissociated cells from eight forelimbs obtained from four different animals pooled together.

Single-cell dissociation and fluorescence-activated cell sorting

Pools of embryonic forelimbs obtained from eight embryos were dissociated into a single-cell suspension using collagenase from Sigma (collagenase type XI) at 37°C for 15min with 10 sec trituration. Cells were then resuspended in FACS solution (10% FCS in PBS with 2mM EDTA). Fluorescence activated cell sorting was performed using the MoFlow ASTRIOS EQ cell sorter with a 100-μm nozzle. Through flow cytometry analysis performed using FlowJo (FlowJo LLC ©) we detected 1,602,844 cells positive for GFP in the autopod tissue and 235,000 simply negative. In the zeugopod tissue, 1,527,167 cells were positive, whereas 1,296,068 were negative giving thus a total of 87% GFP positive autopod cells and 54% positive zeugopod cells.

Single-cell RNA sequencing, library preparation and mapping

Dissociated single-cells were obtained from eight Hoxd11::GFP forelimbs micro-dissected at E12.5. Cells with the highest level of GFP fluorescence (top 20%) were sorted using an Astrios cell sorter with a 100-μm nozzle. 75bp large reads were uniquely mapped to the latest Mus Musculus reference genome (mm10) and the ERCC sequences using bowtie2 (REF:doi:10.1038/nmeth.1923) in local mode. Raw counts for the annotated ENSEMBL mouse genes (GRCm38) and the ERCC were obtained using the RNA-seq module of the HTSstation portal (iDavid, 2014 #274)). Table S2 summarizes the raw counts. All single-cell RNA-seq
data can be found in the Gene Expression Omnibus (GEO) repository under accession number GSEXXX.

**Filtering low quality cells and genes expressed at low levels**

Those counts were used to filter out some low-quality cells based on the following criteria: total number of reads mapped > 250, number of genes "expressed" > 2000 ("expressed"=with count > 0), and percent of reads mapped to Spike-Ins sequences < 25%. A total of 199 cells was retained (123 zeugopods and 76 autopods cells). Genes expressed at low levels were also removed from the rest of the analysis and only genes present (raw count > 0) in at least 10% of either the 76 autopods or the 123 zeugopods cells were retained. *Hox* genes were manually added if they did not satisfy these criteria. A total of 10'948 genes remained. ERCC with null counts through the remaining cells were also excluded from the rest of the analysis. Table S4 summarizes those criteria and Fig. S2.

**Normalization**

Raw counts were normalized with spike-in counts using the R package scran (methods used `computeSpikeFactors` and `normalize` version 1.0.4) (doi:10.18129/B9.bioc.scran). Prior to normalization, size factors were mean-centered to their batch of origin. An additional normalization step was also applied in order to correct for a potential gene length bias. Table S5 compiles all the normalized values.

**Grouping of Hoxd gene combinations for differential gene expression analyses**

*HoxD* groups were defined per cell and were composed by *Hoxd* genes with a minimum normalized expression of 5 when count represented at least 5% of the most expressed *Hoxd* genes in the cell. The differential gene expression analysis was performed with the R package `limma` (version 3.28.21) ([Ritchie, 2015 #42]). Genes with a minimum absolute log fold change of 2 and a BH adjusted p-value less than 0.01 (false discovery rate (FDR) of 1%) were considered differentially expressed.

**tSNE**

The tSNE were computed using the package Rtsne (version 0.13) with the following parameters: 2 dimensions and a perplexity of 30, a maximum of iterations of 3000 and a seed set at 42. The top highly variable genes (HVG) that were used to plot the tSNE in Fig. 3C-D
were selected using the trendVar & decomposeVar methods of the R package scran (version 1.0.4) (10.18129/B9.bioc.scran).

**Pseudo-time**

Diffusion maps (Coifman, 2005 #275) are key tools to analyze single-cell differentiation data. It implements a distance metric relevant to how differentiation data is generated biologically, as cells follow noisy diffusion-like dynamics in the course of taking several differentiation lineage paths (Haghverdi, 2016 #19). The distances between cells reflect the transition probability based on several paths of random walks between the cells (Angerer, 2016 #7). The analysis was performed using the R package destiny (http://bioconductor.org/packages/destiny).

**Network visualization and Venn Diagram**

Network shown in **Fig. 2** was built using weighted interaction networks from various sources of data and is able to process user data into such networks using a system that distinguishes between three different types of user-defined data in its import procedures: real- and binary-valued interaction networks, e.g. physical interaction networks; real-valued gene profile datasets, e.g. multi-sample microarray expression datasets; and binary-valued gene profile datasets (Mostafavi, 2008 #38). Network shown in **Fig. 4** is a summary network of differentially expressed genes that was made with the R package Igraph (version 1.1.2; Csardi, G. & Nepusz, T. 2006). The Igraph Software Package for Complex Network Research. InterJournal 2006, Complex Systems, 1695. http://igraph.org

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COMPETING FINANCIAL INTERESTS
The authors declare no competing financial interests.

FIGURES LEGENDS

Figure 1 – Heterogeneity of posterior Hoxd mRNAs in single-cells
A. Fluorescent in situ hybridization of Hoxd13 mRNA on section reveals discrete expression pattern in the autopod of E12.5 mouse forelimb. B. Single-cell double-labelling of Hoxd13 and Hoxd11 mRNA from E12.5 autopod cells (up, schematic) by fluorescent hybridization followed by flow cytometry detection. The density plot (below) shows a high proportion of double negative cells. C. Hoxd11 expression pattern revealed using a Hoxd11::GFP knock-in mouse strain, with high expression in digits and low expression in interdigital cells of autopod forelimbs, together with strong signals in zeugopod cells. D-E. Scatterplot profile from FACS to enrich for Hoxd11-positive cells using cells with high levels of GFP fluorescence (D, red) subsequently double-labelled for both Hoxd13 and Hoxd11 mRNAs. E. Flow cytometry analysis from the GFP-positive cells from D are shown in a density contour plots where colors highlight the four population of cells expressing various levels of Hoxd11 and Hoxd13. F-H. Double-labeling of GFP (green, marker of Hoxd11-positive cells, F-G) and Hoxd13 (H, red, FISH) with DAPI (magenta) suggests four different combinations of Hoxd positive cells: Double positive for Hoxd13 and Hoxd11, single Hoxd13-positive; single Hoxd11-positive and double negative for Hoxd13 and Hoxd11. I. Histograms showing the correlation of expression where high levels of Hoxd13 are associated with high level of Hoxd11 (two panels on the left). Higher levels of GFP are also observed in cells expressing at least one of the Hoxd genes (right panel). J. Schematic of the four different combinations observed in E and G-H: double negative for Hoxd13 and Hoxd11 (purple); single Hoxd11-positive (green), single Hoxd13-positive (red) and double positive Hoxd13 and Hoxd11 (yellow).

Figure 2 – Single-cell transcriptomics from forelimbs autopod and zeugopod
A. Schematics of forelimb territories harboring different combinations of Hoxd genes activity. Single-cell RNA-seq was performed on micro-dissected 12.5 forelimb autopods (AP, blue) and zeugopods (ZP, yellow) tissues derived from Hoxd11::GFP limbs and positive for GFP. B. T-distributed stochastic neighbor embedding (tSNE) plot of gene expression relationships amongst the 199 single-cells from AP (blue) and ZP (yellow) shows a segregation along the tSNE2 axis.
C. MA-plot produced from cross-analysis between AP and ZP cells. Known genes with differential expression between tissues are indicated on the graph. D. Gene nodes from the differentially expressed genes (DEG) up-regulated in autopod tissue, as computed using co-expression and interaction meta-analysis. E. Cumulative combinatorial expression of Hoxd genes in the autopod cells indicating variability of gene expression across Hoxd genes from cell to cell. F. Spearman’s rank correlation heatmap and hierarchical clustering of genes that covaried with at least one posterior Hoxd gene in the autopod cells. Below are shown the names of Hoxd genes (bold) and their known target genes.

Figure 3 – Combinatorial Hoxd genes expression in single-cells
A. Supervised cluster analysis reveals 16 combinations (clusters) of posterior Hoxd genes in autopod single-cells. B. Boxplots showing normalized expression for all Hoxd genes for the six main clusters from autopod cells, with a representative single cell shown in the top-right corner. C. tSNE with color-coded combinations of posterior Hoxd genes in autopods and zeugopods. D. tSNE showing groups of cells sharing the same combination of expression in autopod (left) and zeugopod (right).

Figure 4 – Analysis of Hoxd mRNAs combinations clusters
A. Network diagram of differentially expressed genes between the six main combinations of posterior Hoxd genes. B. Venn diagram showing the number of overlapping genes differentially expressed between the same combinations. C. Heatmap and unsupervised clustering of the 343 differentially expressed genes in the six groups of cells. D-E. Spearman’s rank correlation heatmaps and clusterings of posterior Hoxd genes and their targets in the full set of 199 cells from autopod and zeugopod (D) and only in autopod cells (E).

Figure 5 – Diffusion pseudo-time across single-cells
A. Diffusion maps of the 76 autopod cells colored per HoxD combinations. This dot plot shows a progression from the simplest combinations to more complex ones. The most significant dimensions, i.e. the first two eigenvectors (DC1 and DC2) are displayed. B. Diffusion maps of the 120 zeugopod cells colored per Hoxd mRNA combinations. C. Hoxd groups centroid. The centroid of a Hoxd group, represented in the diffusion map of the 76 autopod cells. D. Distribution of autopod cells as shown in A are color coded to show their progression along diffusion pseudo-time.
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Figure 1
