Nac1 Coordinates a Sub-network of Pluripotency Factors to Regulate Embryonic Stem Cell Differentiation

Graphical Abstract

Highlights

- A sub-network of Nac1, Oct4, Tcf3, and Sox2 promotes ESC differentiation
- Nac1 controls the sub-network to promote ME and repress NE fate selection
- Nac1 and Oct4 favor the ME, and Tcf3 and Sox2 favor the NE, fate choice
- The four TF levels are constrained within quantitative windows in ME and NE cells

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In Brief

Progenitor cells choose a distinct fate between alternative choices during development. Malleshaiah et al. now show that pluripotent mouse embryonic stem cells decide between mesendodermal and neuroectodermal fates through a sub-network of pluripotency transcription factors, Oct4, Sox2, Tcf3, and Nac1, that is coordinated by Nac1 and constrains protein levels within distinct ranges in each fate.
SUMMARY

Pluripotent cells give rise to distinct cell types during development and are regulated by often self-reinforcing molecular networks. How such networks allow cells to differentiate is less well understood. Here, we use integrative methods to show that external signals induce reorganization of the mouse embryonic stem cell pluripotency network and that a sub-network of four factors, Nac1, Oct4, Tcf3, and Sox2, regulates their differentiation into the alternative mesendodermal and neuroectodermal fates. In the mesendodermal fate, Nac1 and Oct4 were constrained within quantitative windows, whereas Sox2 and Tcf3 were repressed. In contrast, in the neuroectodermal fate, Sox2 and Tcf3 were constrained while Nac1 and Oct4 were repressed. In addition, we show that Nac1 coordinates differentiation by activating Oct4 and inhibiting both Sox2 and Tcf3. Reorganization of progenitor cell networks around shared factors might be a common differentiation strategy and our integrative approach provides a general methodology for delineating such networks.

INTRODUCTION

Stem cells give rise to multiple cell types of an organism through progressive differentiation. While successive new fates are being specified, alternative fates are restricted to create distinct cell lineages (Graf and Enver, 2009; Waddington, 1957). Cell fate-specifying information, in the form of spatial cues or intercellular signals, is processed through molecular networks whose causal regulations and dynamics ultimately define the final cellular outcome (Davidson, 2006). Understanding how such a network changes during cell fate choice is thus crucial to understanding development. Embryonic stem cells (ESCs), which are both pluripotent and self-renewing (Evans and Kaufman, 1981; Martin, 1981; Nishikawa et al., 2007), represent a good model system for addressing this problem.

Mouse ESCs are regulated by an ensemble of transcription factors (TFs), including Pou5f1 (Oct4), Nanog, Sox2, Rex1, Nacc1 (Nac1), Klf4, cMyc, and others (Figure S1A), which promote pluripotency by activating their own expression, and that of other pluripotency genes, and by suppressing genes required for differentiation (Cole and Young, 2008; Ng and Surani, 2011; Niwa, 2007; Silva and Smith, 2008). The key stem cell factor Nanog plays a central role in establishing the self-reinforcing pluripotency network through nested positive feedback and feedforward regulations (Cole and Young, 2008; MacArthur et al., 2012). However, how the self-reinforcing regulations of the pluripotency network change as ESCs differentiate into alternative cell fates is not well understood.

Here, we used an integrative and quantitative approach to analyze how these regulations change as mouse ESCs exit pluripotency and choose between the alternative mesendodermal and neuroectodermal fates. In the mesendodermal fate, Nanog plays a central role in establishing the self-reinforcing pluripotency network through nested positive feedback and feedforward regulations (Cole and Young, 2008; MacArthur et al., 2012). However, how the self-reinforcing regulations of the pluripotency network change as ESCs differentiate into alternative cell fates is not well understood.
and Zfp281) (Kim et al., 2008; Wang et al., 2006) and others (Tcf3, Klf5, p53, and Tbx3), which are thought to have various roles in regulating pluripotency (Cole et al., 2008; Ema et al., 2008; Han et al., 2010; Neveu et al., 2010). This set of TFs included the stem cell trinity of Oct4, Sox2, and Nanog (Silva and Smith, 2008); the Yamanaka reprogramming factors Oct4, Sox2, Klf4, and cMyc (Takahashi and Yamanaka, 2006); and the Wnt-responsive Tcf3, which modulates the balance between pluripotency and differentiation (Atlasi et al., 2013; Cole et al., 2008; Wray et al., 2011).

ESCs can be differentiated in vitro into either the ME or NE fate: Chiron (CHIR99021, a Wnt agonist that inhibits glycogen synthase kinase 3\(^\beta\)) plus Activin-A together promote the ME fate, while retinoic acid promotes the NE fate (Gadue et al., 2006; Thomson et al., 2011; Ying et al., 2003). We employed these signals to induce the ME and NE fates from ESCs, and we primarily focused on analyzing the reorganization of the pluripotency transcriptional network during differentiation (Figures 1 and S1; Supplemental Experimental Procedures). To examine the temporal response to the signals, we followed cell populations for time periods of 0, 24, 32, 40, ..., and 120 hr of ME and NE differentiation (Figures 1B and 1C).

We noticed a higher proportion of undifferentiated cells under ME than NE possibly due to Chiron, which is also known to promote pluripotency (Ying et al., 2008).

We then measured the protein levels for all 13 TFs in a 96-hr time course for each differentiation condition using quantitative immunofluorescence in wild-type ESCs (Figures S1F and S1G; Supplemental Experimental Procedures). Different cell populations were used for each immunostaining to measure up to three TFs and nuclear DNA. Oct4 was measured across all immunostainings as a consistency check. At a given time point and condition, protein measurements were fairly consistent across immunostainings (Figure S1H: representative single-cell distributions of Oct4). ME and NE differentiation signals regulated the TF protein levels differently (Figures S1I and S1J: representative single-cell distributions of Oct4 across all time points).

The majority of the median TF protein levels decreased over time (Figures 1D and 1E), with the decrease being more prominent for NE than for ME, particularly for the trinity of Nanog, Oct4, and Sox2. In contrast, Tcf3 decreased more rapidly for ME than for NE, as expected since the ME condition is induced by Chiron, a Wnt agonist, and Wnt signaling inhibits Tcf3 (Atlasi et al., 2013; Wray et al., 2011). Klf5, p53, and Tbx3 showed similar dynamics under both ME and NE conditions, and Zfp281, cMyc, and Dax1 showed the least overall changes. These temporal changes are largely consistent with previous observations involving both directed and undirected differentiation experiments (Lu et al., 2009; MacArthur et al., 2012; Pereira et al., 2006; Thomson et al., 2011). Our results captured the varying
dynamics of the TFs as ESCs exit pluripotency and transit toward the ME or NE fate. These dynamic changes in TFs coincide with the onset of differentiation marked by T or Sox1 expression (Figures 1B and 1C). To understand how these dynamic changes in TF levels influence their causal regulations and the cell fate transition and to identify the key regulators behind these changes, we turned to computational analysis.

Figure 2. Identification of Potential Differentiation Regulatory TFs by Computational Analysis
(A–D) Principal-component analysis (PCA) using the median values of nine pluripotency factors, Oct4, Nanog, Sox2, Klf4, Rex1, Nac1, Zfp281, Dax1, and cMyc (A and B) and these nine plus Tcf3 (G and D) for the ME (red) and NE (blue) conditions is shown. (E and F) PCA with all 13 measured TFs. Principal component (PC) projections are shown as PC1 versus PC2 and PC2 versus PC3 plots.

(G and H) Dynamic Bayesian network (DBN) analysis using median values of the ten TFs (above nine plus Tcf3) for ME (G) and NE (H) differentiation. The nodes are colored using a scale based on dominance scores that have been max-normalized for each network separately as follows: white (≤0.5), red gradient (0.5–1), and red (1). The edges are colored using a scale based on their posterior probability predictions with a threshold of 0.25 as follows: off-white (0.25), gray scale (0.25–0.5), and black (>0.5). The extent of connectivity measured by dominance scores indicated the top four factors in the ME (G) and NE (H) networks. DBNs are drawn using Cytoscape. See also Figure S2 and Experimental Procedures.

Computational Analysis Reveals Potential Pluripotency Factors Involved in Differentiation
We employed principal-component analysis (PCA) to extract the main features of the differentiation dynamics. PCA determines a sequence of orthogonal vectors that best captures the variance in a dataset (Jolliffe, 2002). The contribution of individual factors toward the collective behavior of all factors within a network can be assessed using PCA. We represented the median levels of each TF as a multi-dimensional vector, at all 13 time points (samples every 8 hr from 0 to 96 hr) for both ME and NE conditions.

We first analyzed the extended pluripotency network with the following nine TFs: Oct4, Sox2, Nanog, Klf4, Rex1, Nac1, Zfp281, Dax1, and cMyc (Kim et al., 2008; Wang et al., 2006). The first three principal components (PCs) captured over 95% of the variance in the data (Figures 2A and 2B). PC1, which accounts for ~80% of the variance, captured the change in time; PC2 partially captured the difference between differentiation conditions; and PC3 captured the changes within each condition. Although the two fates had different trajectories, they were indistinguishable along PC2 at several ME and NE time points (Figures 2A and 2B). This was in contrast to the mutually exclusive ME or NE choice observed (Figures 1B and 1C),
suggesting that the nine factors were not sufficient to fully distinguish the two fate choices.

A second PCA analysis, which included each of the additional TFs measured as the tenth TF, revealed a distinct role for Tcf3. While there were no changes with Klf5, p53, or Tbx3, inclusion of Tcf3 surprisingly clearly separated the two differentiation trajectories (Figures 2C, 2D, and S2A–S2C). Tcf3 contributed significantly to PC2 (Figure S2M), suggesting that it played an important role in distinguishing the ME and NE fates. To further verify the role of each TF, we excluded them one at a time and repeated the analysis. While exclusion of Nac1, Oct4, or Sox2 had severe effects on the trajectory separation (Figures S2D–S2F), exclusion of the other TFs on the other hand had less effect (Figures S2G–S2L). In agreement with their role, the four TFs Nac1, Oct4, Sox2, and Tcf3 were the dominant contributors to PC coefficients (Figure S2M).

Furthermore, PCA on all 13 proteins gave no further improvement (Figures 2E and 2F).

As a complementary approach to PCA, we used dynamic Bayesian networks (DBNs). Bayesian networks have been used widely to infer causal relationships and to identify key regulators from complex biological data (Friedman et al., 2000; Pe’er, 2005; Figure S2N). Non-homogeneous DBNs allow different model structures during each segment of a time course, making it particularly applicable to processes like stem cell differentiation in which the underlying causal network is not necessarily static (Dondelinger et al., 2013). We assumed the initial DBN network to be random, and we incorporated in it the previously known ME-promoting function of Oct4 through Sox2 regulation (Thomson et al., 2011) and the general differentiation-promoting function of Tcf3 through Nanog, Oct4, and Rex1 regulations (Cole et al., 2008; Pereira et al., 2006; Wray et al., 2011; Supplemental Experimental Procedures).

We used a Markov chain Monte Carlo algorithm to find optimal and robust non-homogeneous DBNs with the median values of the ten TFs considered above (Oct4, Sox2, Nanog, Klf4, cMyc, Rex1, Nac1, Dax1, Zip281, and Tcf3), at all 13 time points, for ME and NE conditions separately (Figures S2G and S2H). The inferred ME and NE networks were distinct from each other, suggesting that the causal network of pluripotency may undergo major reorganization during differentiation. Change-point analysis further suggested that the initial network changes the most within the first 32 hr of differentiation (Figures S2O and S2P), with no significant changes after 72 hr. These dynamic changes to the pluripotency network precede the emergence of the ME and NE fate markers (Figures 1B and 1C). To identify the dominant regulators in the predicted causal networks, we assigned a score for each protein to estimate the degree of its connectedness (Supplemental Experimental Procedures). These dominance scores indicated the most important proteins that determine the inferred ME and NE networks as follows: Tcf3, Nac1, Oct4, and Klf4 for ME; and Sox2, Tcf3, Nanog, and Klf4 for NE (Figures 2G and 2H).

Given that PCA analysis also had highlighted the potential differentiation roles for Sox2, Oct4, Nac1, and Tcf3, we pursued the hypothesis that these proteins, plus Nanog and Klf4, may play a key role in regulating the ME and NE fate choices. To explore this we turned to quantitative single-cell experiments.

Nac1 and Oct4 Promote the ME Fate and Tcf3 and Sox2 Promote the NE Fate

We first examined the expression ranges and the correlations between protein levels of the above six TFs in pluripotent cells maintained in the presence of leukemia inhibitory factor (LIF) and BMP4 (ES condition). Protein levels were measured in single cells using quantitative immunofluorescence. Max-normalized protein levels indicated that all factors, except Tcf3, occupied all three ranges of expression, low, medium, and high (Figures S3A–S3F). In addition, their expression levels correlated highly with each other (Figure S3G), consistent with the self-reinforcing nature of the pluripotency network (Cole and Young, 2008; Pe’er et al., 2012; Niwa, 2007; Silva and Smith, 2008). Tcf3 levels, on the other hand, remained low in comparison to their maximum under the NE condition and poorly correlated with Nac1, Oct4, and Nanog.

Next, to delineate the quantitative patterns of these six proteins, we measured their levels in single ME and NE cells at 72 hr into differentiation. Although the measurements at earlier time points might be informative, especially to address heterogeneity in response, at 72 hr, differentiation has just reached its maximal level with reduced variability, allowing us to collect data from a statistically significant number of ME- or NE-positive cells (Figures 1 and S1). This time point also was well suited for assessing the impact of the perturbation experiments reported below. The cell fate marker signal was normalized across both ME and NE conditions to the maximum of its single-cell level in the respective condition: ME for T and NE for Sox1. Similarly, each TF level, except that of Tcf3, was normalized across both conditions to its single-cell maximum under the ES condition. Since Tcf3 was highest under NE, it was normalized to its maximum under the NE condition (Supplemental Experimental Procedures). Each normalized protein value was plotted against the normalized ME or NE fate marker (Figures 3 and S3J–S3M). As observed earlier (Figures 1B and 1C), the differentiation was mutually exclusive, with no NE fate visible under ME and no NE fate visible under ME (Figures S3H and S3I).

Normalized TF levels were divided into three ranges as follows: low (0–0.2); medium (0.2–0.7), and high (0.7–1.0); and fate markers were divided into low (0–0.2) and high (0.2–1.0). Two characteristic shapes are observed from the single-cell plots of TF level versus fate marker (Figures 3 and S3J–S3M) as follows: an L shape, indicating mutual exclusivity between ES (undifferentiated cells below the fate marker threshold) and ME (for Sox2, Tcf3, Nanog, and Klf4) or NE (for Nac1, Oct4, and Nanog) cells; and a V shape, in which the protein level can be high when the fate marker is low, but is restricted to be medium if the fate marker is high (for Nac1 and Oct4 in ME and for Sox2 in NE cells). Tcf3 is somewhat similar; it exhibits a truncated L shape under ME, presumably due to its inhibition by Chiron, and a broader cone-like shape under NE, in which its level is medium or high when the fate marker is high.

These results suggested that the central pluripotency factors, such as Nanog and Klf4, are not required for ESC differentiation and were generally repressed (Figures S3J–S3M). The remaining four factors, on the other hand, were differentially regulated.
Nac1 and Oct4 were constrained within a quantitative window, neither too high nor too low, in ME cells, but were repressed in NE cells (above the fate marker threshold in Figures 3A–3D). Similarly, Sox2 was constrained to intermediate levels and Tcf3 to intermediate-to-high levels in NE cells, but they were repressed in ME cells (Figures 3E–3H). Additionally, we tested four other TFs, Esrrb, Sall4, Smad1, and E2f1, for their quantitative levels in ME and NE cells. These TFs are known to promote pluripotency (Chen et al., 2008; Dunn et al., 2014) but were not part of our original dataset. We found that, in a way similar to Klf4 and Nanog, the levels of these factors are equally downregulated in both ME and NE cells (Figure S4). Together, Figure 3. Differential Regulation of Potential TFs in the ME and NE Cells (A–H) Representative qualitative images (left) and quantitative single-cell measurements (right) for Nac1 (A and B), Oct4 (C and D), Sox2 (E and F), and Tcf3 (G and H) in ME and NE differentiation conditions. T and the indicated proteins were quantified by immunofluorescence in Sox1-GFP ESC line. Mean fluorescence intensities of each protein from single cells, normalized as explained in the text, are plotted against the T (ME marker) and Sox1-GFP (NE marker) levels, also normalized, in each panel. Cells were fixed, stained, and measurements were done at 72 hr of differentiation. Dashed lines indicate the division of TF levels into low, medium, and high and fate markers into low and high ranges, as explained in the text. Scale bars represent 35 μm. See also Figures S3 and S4.
these distinct quantitative patterns suggested a requirement of Nac1 and Oct4 for the ME choice and Sox2 and Tcf3 for the NE choice.

To confirm this, we utilized small interfering RNA (siRNA)-mediated perturbation to modulate the protein levels during differentiation. We first tested the siRNA for efficiency in down-regulating the specific TF and effects on ESC survival. While a scrambled siRNA pool was used as a negative control, pools consisting of siRNAs against four sites of each target were used to downregulate the TFs (Supplemental Experimental Procedures). Quantitative immunofluorescence of the target proteins at 72 hr after siRNA transfection showed a significant reduction in their expression, while ES colonies were still largely intact (Figures S5A–S5G). Mock transfections with the scrambled siRNA pool (negative control) did not alter the normal TF levels in ESCs (Figure S5G). However, after 6 days of transfection, a varying degree of effects was observed on ES colony survival (Figure S5H). Consistent with their central role, we found that Nac1, Oct4, Sox2, and Nanog downregulation reduced ES survival by up to 80%. In contrast, Tcf3 and Klf4 had lesser effects, also consistent with their known functions. While Klf4 is redundant for the ES state (Jiang et al., 2008), Tcf3 downregulation is known to stabilize the ES state (Pereira et al., 2006; Wray et al., 2011). Furthermore, mock transfection of ESCs with scrambled siRNA did not induce ME or NE fates at 72 hr after transfection, thus ruling out the differentiation artifacts from transfection reagents alone (Figure S5I). Similarly the transfection reagents did not alter the ESCs’ ability to differentiate into either the ME or NE cell fates (Figures S5G, S5J, and S5K).

We then used the siRNAs to downregulate Nac1 and Oct4 levels during ME differentiation and Sox2 and Tcf3 during NE differentiation. The resulting single-cell protein levels, at 72 hr of differentiation, were normalized as described above but with maximal levels taken from mock-transfected cells (negative control). Forcing Nac1 and Oct4 to relatively low levels compromised the ME choice (Figures 4A and 4B), more prominently with Nac1 perturbation. The NE choice, on the other hand, was equally compromised on reducing either Tcf3 or Sox2 level (Figures 4C and 4D). Ectopic overexpression of these TFs in cells treated with siRNA largely rescued both the target TF levels as well as the respective cell fate marker levels (Figures 4E–4H). Ectopic overexpression of these TFs in cells treated with siRNA largely rescued both the target TF levels as well as the respective cell fate marker levels (Figures 4E–4H). Although acute knockdown of these genes would cause defects in ESC maintenance (Figure S5H), these ME and NE lineage-specific phenotypes at 72 hr suggest that the general ESC differentiation is not compromised under our conditions. In addition to the specificity of siRNA pools used, the above results also confirm the important role of some of the pluripotency TFs in promoting differentiation: ME by Nac1 and Oct4 and NE by Sox2 and Tcf3.
Nac1 Activates Oct4 and Inhibits Tcf3 and Sox2 to Favor the ME Fate

The repression of Nac1 and Oct4 in NE cells and their presence in ME cells indicated that they might oppose the NE choice (Figures 3A–3D). Similarly, the repression of Sox2 and Tcf3 in ME cells and their presence in NE cells indicated that they might oppose the ME choice (Figures 3E–3H). To gain further insights into the possible mechanisms that may explain the distinct quantitative patterns observed, we asked whether the regulations among these TFs were different in ME and NE cells. We hypothesized that in order to promote ME, Nac1 and Oct4 might repress NE by inhibiting Sox2 and Tcf3, and vice versa. To understand these, we systematically tested for the causal relationships among Nac1, Oct4, Sox2, and Tcf3 during both the ME and NE differentiation conditions, using specific siRNA perturbations.

We examined the impact of downregulating Nac1 or Oct4 during ME differentiation and Tcf3 or Sox2 during NE differentiation on changes in the levels of all four identified factors, Nac1, Oct4, Sox2, and Tcf3, as well as Nanog (Figures 5A–5D and S6A–S6Q). Nac1 downregulation during ME differentiation resulted in the decrease of Oct4 level but an increase in both the Sox2 and Tcf3 levels (Figures 5A–5C and S6Q). Similarly, downregulation of Oct4 during ME differentiation resulted in an increased Sox2 level (Figures 5D and S6Q) but had no effect on Nac1 and Tcf3 levels (Figures S6A and S6B). These results indicated that, during ME choice, Nac1 favors Oct4 expression and inhibits both Tcf3 and Sox2 and that Oct4 inhibits Sox2. By inhibiting the NE-promoting TFs, both of these Nac1- and Oct4-mediated actions favor the ME choice, and Nac1 in particular may play a central role. In agreement with this, Nac1 downregulation resulted in a prominent reduction of the ME fate (Figures S6M, S6N, and S6Q).

In analyzing the NE fate, whereas Tcf3 and Sox2 downregulation equally reduced the NE fate (Figures S6O and S6P), we found no evidence that either Tcf3 or Sox2 inhibits either Nac1 or Oct4 (Figures S6E, S6F, S6I, and S6J). It is known that Tcf3 inhibits Oct4 and Nanog under pluripotency condition (Cole et al., 2008; Pereira et al., 2006), but this may not continue in the NE condition. However, both Nac1 and Oct4 are downregulated in NE (Figures 3B and 3D), suggesting their indirect regulation, either through retinoic acid signaling itself or other factors, rather than one of the specific factors considered here. Similarly, a role...
Figure 6. Nac1 Regulates the Extent of NE Choice and Suppresses It in Naive ESCs through Tcf3 Inhibition

(A and B) The model predictions for changes in Nac1, Tcf3, Sox2, and Oct4 levels on partial knockdown (KD) of Nac1 (A), and changes in the ME (T) and NE (Sox1) fate choices (B) upon partial knockdown of indicated protein, during the ME (red) and NE (blue) conditions, are shown.

(C–G) Nac1 downregulation-mediated changes in Tcf3, and Sox1-GFP levels during NE differentiation (C and D), under ES condition (E and F), and T and Sox1-GFP levels under ES condition (G). Images show the overlay of Nac1 and Sox1-GFP images at the indicated condition.

Figure 6 legend continued on next page.
for the ME differentiation signals to inhibit Tcf3 was observed (Figure 3G).

Downregulation of Nac1 and Oct4 under ME differentiation and Sox2 during NE differentiation led to an enhanced reduction in undifferentiated ESCs (as measured by Nanog levels) (Figures S6C, S6D, and S6H). Similarly, while Oct4 and Nac1 were limited from their high range (0.7–1) of expression under ME (Figures 5A and 5A), they were restricted to the low range (0–0.3) in cells with downregulated Sox2 under NE condition (Figures S6E and S6F). On the other hand, Tcf3 downregulation had no effects on Oct4, Nac1, Sox2, and Nanog levels (Figures S6I–S6L). These results are consistent with the Nac1-, Oct4-, and Sox2-mediated activation of Nanog to maintain pluripotency (Figure S1A; Kim et al., 2008), and they further suggest that these positive regulations are lost in ME- and NE-positive cells (Figures S3J and S3K).

Nac1 Modulates the Extent of NE Fate Selection through Tcf3 Inhibition

Together, the above data suggested that the self-reinforcing regulations among a subset of TFs—Nac1, Oct4, Tcf3, and Sox2—in ESCs are modified so as to favor the ME choice over the NE choice. To reveal further functional outcomes that might be concealed in our data, we constructed a mathematical model using parsimonious assumptions to describe how the levels of Nac1, Oct4, Sox2, and Tcf3 adjust to each other during either the ME or NE differentiation (Figure 5E; Supplemental Experimental Procedures). The model was formulated to regulate ME choice through repression of Tcf3 by Nac1 and of Sox2 by both Nac1 and Oct4, or NE choice through repression of Nac1 and Oct4, mediated by either the NE differentiation signal itself or by some as-yet-unknown factors. The model was set up to exhibit a single steady state for each differentiation regime (Figure S6R), thereby reflecting the mutually exclusive choice between ME and NE fates (Figures 1B and 1C). The model’s parameter values were chosen manually to be consistent with our experimental findings, so that the eventual steady-state levels of Nac1, Oct4, Sox2, and Tcf3 fell within the ranges for cells expressing high levels of the respective ME or NE fate marker. We also performed parameter sensitivity analysis by varying each of the model parameters by ±25% and found that the model is robust to changes in parameter values (data not shown). Of the various TF concentrations and their combinations tested, only high levels of Nac1 and Oct4 favored the ME fate, and high levels of Tcf3 and Sox2 favored the NE fate (Figure 5F).

Next, we utilized this simple model to predict the effects of selective perturbation of Nac1, Oct4, Sox2, and Tcf3 on both the TF levels and the fate selection. For the most part, simulated partial knockdown of these factors recapitulated experimental results with siRNA perturbations (Figures 6A, S6S, 5A–5D, and S6A–S6P). However, the model also showed an increased level of Tcf3 (Figure 6A) and an enhanced NE fate selection when Nac1 was knocked down in NE condition (Figure 6B). To confirm these unexpected predictions, we first downregulated Nac1 during NE differentiation and consistently found twice the level of Tcf3 (Figure 6C). To test the fate selection predictions, we downregulated each of the four TFs during both ME and NE differentiations, and we measured the changes to both cell fates (Figures S7A–S7L). The results confirmed the ME-promoting role of Nac1 and Oct4 and the NE-promoting role of Sox2 and Tcf3. In addition, enhanced Sox1-GFP signals upon Nac1 downregulation during NE differentiation confirmed the model prediction (Figures 6B, S7E, and 6D). To examine the role of Nac1 further, we conducted similar experiments in ESCs in the absence of any differentiation signals. Surprisingly, downregulating Nac1 resulted in up to four times higher Tcf3 levels (Figure 6E). Nac1 downregulation alone was sufficient to trigger higher Sox1-GFP expression and induce NE fate as the default choice in up to 48% of naive ESCs (Figures 6F–6H). Furthermore, ESCs with ectopic overexpression of Nac1 were blocked from NE fate selection during the NE differentiation process (Figures 6I–6K). These results suggested that Nac1 inhibits the NE-promoting Tcf3 in ESCs as well as in ME and NE cells.

Both the model and the experimental results indicated that downregulation of Tcf3 or Sox2 did not have any effects on the ME choice (Figures S7C and S7D). Similarly, Oct4 downregulation had no effect on the NE choice (Figure S7F). Taken together, these results suggest an important central role for Nac1 in regulating ESC differentiation into both the ME and NE cell fates. While Nac1 in combination with Oct4 was required for the ME choice, it strongly opposed NE and its downregulation was required for the NE choice.

Nac1 Differentially Binds to the DNA Regulatory Regions in ES, ME, and NE Cells

How does Nac1 coordinate other key pluripotency TFs to regulate ESC differentiation? Fate-specific modulation of gene expression often results from direct binding of a TF to the regulatory regions of its target DNA (Davidson, 2006). In ESCs, pluripotency factors bind extensively to gene regions of each other to regulate their expression (Cole et al., 2008; Kim et al., 2008). Similarly, Oct4 is known to bind the regulatory region of Sox2 during ME differentiation (Thomson et al., 2011). To know if Nac1 implemented similar mechanisms, we assessed its binding to genomic DNA in ES, ME, and NE cells. We performed chromatin immunoprecipitation (ChIP) to isolate Nac1-bound genomic DNA followed by next-generation sequencing (ChIP-seq) to identify the specific gene regions and their enrichment.
Figure 7. Nac1 Binds Differentially to Regulatory Gene Regions in ES, ME, and NE Cells

(A) Percentage of Nac1-bound regions associated to annotated genes of the mouse genome (build mm9) in ESCs is shown.

(B) De novo motifs identified from the Nac1-bound target region sequences in ESCs. Three motifs covering the highest percentage of targets are shown.

(C) -log10 (binomial p value) for various biological processes and molecular functions.

(D) Nac1 & Oct4-centric module

(E) Nac1 & Myc-centric module

(F) Heatmaps showing expression levels of Oct4, Tcf3, Sox2, Nac1, and Klf4 in ES, ME, and NE cells.

(G) Protein level changes for Nac1, Oct4, Nanog, Tcf3, and Sox2 in ES, ME, and NE cells.

Please cite this article in press as: Malleshaiah et al., Nac1 Coordinates a Sub-network of Pluripotency Factors to Regulate Embryonic Stem Cell Differentiation, Cell Reports (2016), http://dx.doi.org/10.1016/j.celrep.2015.12.101
Because little is known about which target genes Nac1 regulates, we first assessed such regulation in ESCs. The majority of Nac1-bound regions were proximal to the transcription start site (TSS) of associated genes (within 5 kb up or downstream of the TSS), suggesting that it played a role in regulating the transcription of its target genes (Figure 7A). The de novo Nac1-binding motifs identified were dominated by cytosine followed by guanine content and were found in a majority of its target sequences (Figure 7B). Nac1 motifs shared a similarity with TF motifs involved in the regulation of ESCs, development, and transcription (Kim/5/7/1, Sp1, Smad3, etc.; Bouwman and Philipsen, 2002; Jiang et al., 2008; Mullen et al., 2011). In addition, we observed significant enrichment of Nac1 binding at gene regions associated with the regulation of gene transcription, translation, ESC state, development, cell cycle, and signaling (Figure 7C; Table S1).

In ESCs, ChIP-seq profiling of several pluripotency TFs has led to the identification of Oct4- and Myc-centric modules (Chen et al., 2008; Ng and Surani, 2011). While the Oct4-centric module includes the core factors Oct4, Sox2, and Nanog, among others, the Myc-centric module includes c-Myc, E2f1, and Zfx. To know how Nac1 targets compare with these TFs, we analyzed its target genes with respect to the Oct4 and Myc module TF targets (Figures 7D and 7E). Nac1 shared a few hundred targets with each individual factor and all three factors of the Oct4 module (433) (Figure 7D). Surprisingly, in comparison to the Oct4 module, Nac1 shared approximately twice the number of targets with all three factors of the Myc module (846) (Figure 7E). This cross-analysis suggests that Nac1 is an important member of the Myc module rather than the Oct4 module. Moreover, the majority of Nac1-binding regions are adjacent to the TSS (Figure 7A), which is a characteristic feature of the Myc-centric module (Chen et al., 2008; Ng and Surani, 2011). Together these cross-comparisons reveal important new insights into how Nac1 regulates its target genes.

We then analyzed differential binding patterns of Nac1 in ES, ME, and NE cells by ChIP-seq. ESCs were grown in culture with LIF and BMP, and sorted T-GFP- and Sox1-GFP-positive cells at 72 hr of differentiation were used as ME and NE cells, respectively. Differential enrichment peaks were observed for all four differentiation-promoting pluripotency factors (Figure 7F): Nac1 bound to regions of Oct4, Sox2, Tcf3, itself, and Klf4 in ES and ME cells. In NE cells, on the other hand, there were no peaks for Oct4 and the peaks at other TF regions were weakly enriched. To verify these observations further and to rule out the artifacts of sequencing, we performed qPCR for the sites detected in ChIP-seq by using ChIP samples from ES, ME, and NE cells. We validated ChIP-qPCR by testing known Nac1-binding regions for Oct4, Sox2, and Nanog in ESCs (Kim et al., 2008) and multiple non-target regions from ChIP-seq analysis. Although there was no enrichment of non-target regions, known and new Nac1 binding regions were significantly enriched (Figure 7F) in ESCs. Nac1 in ESCs bound to its own regulatory regions and that of Tcf3 on a par with its binding to Sox2 and Nanog regions (Figures 7F and S7I). Nac1 bindings to these TFs were very similar in ME cells, except for its slightly reduced binding to Oct4 and an increased binding to Tcf3 (Figure S7I). A striking difference was observed, however, in NE cells where Nac1 bound to none of the factors significantly (above 3-fold), except for Tcf3 and Sox2.

In assessing Nac1 ChIP-seq data, we also observed differential enrichment of gene regions associated with various cell-signaling pathways, mesoderm, neural tube patterning, and mouse phenotypes (Figures S7J and S7K). Particularly the ME-related signaling pathways, such as Wnt and Nodal, were enriched in ME cells more than in NE cells (Gadue et al., 2006). Verification of selected gene regions related to mesodermal development (Smad2/3/4) and Wnt/Activin/Nodal signaling genes by qPCR confirmed their high enrichment in ME cells, followed by ESCs and lesser enrichment in NE cells (Figure S7F; Fei et al., 2010; Gadue et al., 2006). Assessment of NE-promoting retinoic acid-signaling targets (Rhinn and Dollé, 2012) gave similar results: higher enrichment in ES and ME cells than in NE cells. These results are consistent with the different Nac1 protein levels; higher levels in ES and ME cells allowed maximal binding to Nac1 targets, and its minimal level in NE cells resulted in the loss of binding.

**DISCUSSION**

The data presented here reveal differentiation-promoting functions of Nac1; confirm known roles of Oct4, Sox2, and Tcf3 (Atiasi et al., 2013; Kishi et al., 2000; Thomson et al., 2011); and extend the repertoire of pluripotency TFs also utilized to regulate early stages of ESC differentiation. Through integrative and quantitative approaches, we have shown that these four pluripotency factors regulate the ME versus NE choice of ESCs in distinct ways. Nac1 in particular appears to play important roles in fine-tuning the extent of ME and NE fate selection, as well as the ESC state. Although we cannot rule out direct or indirect participation of other TFs in regulating the ME and NE fates, the TFs we have identified play significant roles in fate selection.
Although Nac1 had been shown to be an important TF for the pluripotency (Kim et al., 2008; Wang et al., 2006), its functions in ESCs as well as in other cell types remain surprisingly poorly understood. Nac1 is known to promote tumor proliferation and gain of resistance to chemotherapy, including in cell types that originate from ME (Jinawath et al., 2009; Nakayama et al., 2006). It is conceivable that Nac1 may play a role in the putative link between stem cells and cancer, with diverse functions in regulating both development and disease.

Nac1-knockout mice exhibit a lower survival rate for embryos or newborns, with surviving mice showing gross skeletal abnormalities (Yap et al., 2013). We observed significant enrichment of Nac1 binding at regions associated with mouse phenotypes such as abnormal rib development (Figure S7K), but the effect of Nac1 knockout on early embryogenesis, especially in mice that do not survive, has not been studied and would complement the results obtained here.

Similar to our results, certain TFs are reused during cell fate specification from progenitors during hematopoietic development (Mercer et al., 2011; Rothenberg et al., 2010). For instance, during early hematopoiesis, PU.1 regulates myeloid, B cell, and T cell fate choices from their common multipotent progenitor cells (Mercer et al., 2011). PU.1 along with CEBPα also regulates the macrophage versus neutrophil fate choice of granulocyte-macrophage progenitor cells (Lasio et al., 2006). In these cases, it is still not clear how the causal networks are reorganized from the progenitor cells to differentiated cell types. However, taken together with our findings in ESCs, these results suggest a common strategy of reorganizing the progenitor cell networks around a subset of shared factors. This may be an efficient way to specify successive and alternative cell types during development.

The data presented here suggest that the ES state may exist as a composite balance among several of the differentiation-promoting TFs (Figure 7G). Regulation of the quantitative pattern of Nac1, Oct4, Sox2, Tcf3, and the central factor Nanog may dictate the balance of the pluripotent ES state versus the ME and NE states. In ESCs, Nac1, Oct4, Sox2, and Nanog spanned all three possible ranges (low, medium, and high) of protein levels, and they were expressed in strong correlation with each other (Figures S3A–S3G). This suggests that, irrespective of large cell-to-cell variations, the self-reinforcing regulations among these TFs may promote a mutually balanced stable ES state that resists differentiation. A strict stoichiometry of protein levels among these TFs may reinforce the ES state, and Nac1 might play a key role in it. The differentiation signals, however, effectively compromised the ES state by eliminating Nanog-dependent positive regulations (Cole and Young, 2008; MacArthur et al., 2012; Niwa, 2007; Silva and Smith, 2008). They promoted a distinct fate choice by restraining the levels of specific TFs within quantitative windows and repressing those required for the alternative fate. However, the roles of extrinsic signals for ME and NE lineage commitment are not completely resolved in our data. While we show that the signals reconfigure the TFs into forming new lineage-specific networks, it is not clear how the reconfigured TFs integrate with signaling effectors, such as Smad 2/3 and RARs. The connections between the transcriptional and signaling networks are currently not well understood.

In this study, we have investigated the differentiation process through which pluripotent ESCs decide between the ME and NE fates. We used computational analysis of population-averaged data (Figure 2) to suggest which pluripotency TFs might play a key role in these cell fate transitions, and then we exploited quantitative single-cell analysis and siRNA perturbation to identify the causal interactions that underlie differentiation (Figures 3, 4, and 5). This allowed us to make predictions at the molecular level, such as the role of Nac1, which emerged as a central regulator of differentiation. Our integrative methodology fully exploits protein-level data, in contrast to previous work on pluripotency network reconstruction in ESCs, which relies on a Boolean (on/off) approximation (Dunn et al., 2014; Xu et al., 2014). We believe our approach can be extended to analyze complex molecular networks underlying cell fate choice in other developmental and disease contexts, especially when augmented with new methods for measuring multiple proteins in single cells (Bendall et al., 2011).

**EXPERIMENTAL PROCEDURES**

**ESC Culture and Differentiation**

Mouse ESCs, R1 (E14-Tg2A), T-GFP, and Sox1-GFP, were cultured by standard methods in knockout DMEM supplemented with non-essential amino acids, sodium pyruvate, L-Glutamine, β-mercaptoethanol, 15% fetal calf serum (HyClone), and LIF (1,000 μg/ml, Millipore ESG01106). CHIR99021 (Stemgent 04-0004-02) and Activin-A (R&D Systems 338-AC-010) and retinoic acid (Sigma-Aldrich) were used at the indicated concentrations for ME differentiation and NE differentiation, respectively.

**Quantitative Immunofluorescence**

Immunofluorescence was performed as described previously (Muñoz-Descalzo et al., 2012). The primary antibodies used are listed in the Supplemental Experimental Procedures. A Nikon Ti inverted confocal microscope fitted with perfect focus, 20× Plan-Apochromatic objective (numerical aperture [NA] 0.75) and Hamamatsu ORCA-AG cooled charge-coupled device (CCD) camera were used for imaging.

**Image Analysis and Normalization**

Quantifications of fluorescence intensities were performed by semi-automated image analysis using Cell Profiler (Broad Institute; Carpenter et al., 2006). Cells were segmented using the DAPI signals. Protein normalizations were performed across multiple conditions using the maximum signal of a protein as described.

**Perturbations**

Target-specific ON-TARGET plus SMART siRNA pools, consisting of four constructs, from Dharmacon (Thermo Scientific) were used to downregulate TFs. Cells were transfected with 25 nM of either scrambled siRNA (negative control) or target-specific siRNA pools using DharmaFECT reagent 12 hr prior to the addition of differentiation signals. Rescue of TFs during differentiation was achieved through their ectopic overexpression from pmCherry-N1 vector under CMV promoter.

**ChIP**

ChIP was performed using the Magna ChIP reagents (Upstate 17-409) according to the manufacturer’s guidelines. ESCs grown in LIF and BMP4 medium and sorted T-GFP- (ME) and Sox1-GFP- (NE) positive cells were used to perform ChIP with 3 μg of Nac1 antibody (Abcam ab29047).

**ChIP-Seq Analysis**

Nac1-bound DNA was sequenced by next-generation sequencing using Illumina HiSeq2500. Sequences were aligned to mm9 mouse genome build using Bowtie, peaks were called by MACS, and genes were assigned to peaks using
GREAT and Chip-Enrich. While DREME and TOMTOM were used for motif discovery, gene ontology and pathway enrichment were performed using GREAT and Chip-Enrich.

**ChiP-qPCR**

The qPCR on ChiP DNA from ES, ME, and NE cells was performed in triplicates on CFX96 RT-System (Bio-Rad) using the RT2 SYBR Green Fluor (QIAGEN 330510). Ct values were first normalized to the input and then to the normal mouse IgG (negative control) to calculate fold enrichment ($2^{\Delta \Delta Ct}$). Relative fold enrichment was calculated over Igx1a. The primers used for qPCR analysis are listed in the Supplemental Experimental Procedures.

**Mathematical Modeling**

Simulation and analysis of mathematical model were performed using open source tools for Python (SciPy). Description of the model, its assumptions, and parameters are provided in the Supplemental Experimental Procedures. Full descriptions of the experimental and computational methods are provided in the Supplemental Experimental Procedures.

**PCA and DBN**

PCA was performed with custom code written in R by using singular value decomposition of the centered matrix of median values. DBN on ME and NE data was performed by implementing a custom modified version of the EDISON package (Dondelinger et al., 2013) in R. Modifications introduced to the EDISON package are described in the Supplemental Experimental Procedures. The DBN code is available for download at https://github.com/meghapadi/StemCellDBN.

**SUPPLEMENTAL INFORMATION**

Supplemental Information includes Supplemental Experimental Procedures, seven figures, and one table and can be found with this article online at http://dx.doi.org/10.1016/j.celrep.2015.12.101.

**ACKNOWLEDGMENTS**

We thank Anna-Katerina Hadjantonakis and Gordon Keller for the ESC lines used in this study; Rebecca Ward, Marc Kirschner, Stephen Michnick, Tathagata Dasgupta, Satabhisa Mukhopadhyay, Victor Li, and members of the J.G. and A.-M.-A. labs for comments and discussions; Panos Xenopoulos and Silvia Munoz-Descazo for help with ESC culture and immunofluorescence; Tathagata Dasgupta for help with the Bayesian analysis; the Nikon Imaging Center at Harvard Medical School for help with microscopy; and the Flow Cytometry facility at the Department of Systems Biology. This work was supported by the following funding sources: the Human Frontier Science Program grant RGP0029/2010 (M.M., A.-M.-A., and J.G.); the Canadian Institutes of Health Research award 201411MFE-339075-254265 (M.M.); the NIH award R01GM105375 (M.M. and J.G.); the European Research Council Advanced Investigator award (P.R. and A.-M.-A.); the National Human Genome Research Institute grant K25HG008631 (M.P.); and the NIH grant 1R01 HL111759 (J.G.).

Received: June 19, 2015

Revised: October 19, 2015

Accepted: December 23, 2015

Published: January 28, 2016

REFERENCES


