Phosphoproteomic profiling reveals a defined genetic program for osteoblastic lineage commitment of human bone marrow–derived stromal stem cells

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Bone marrow–derived mesenchymal stem cells (MSCs) differentiate into osteoblasts upon stimulation by signals present in their niche. Because the global signaling cascades involved in the early phases of MSCs osteoblast (OB) differentiation are not well-defined, we used quantitative mass spectrometry to delineate changes in human MSCs proteome and phosphoproteome during the first 24 h of their OB lineage commitment. The temporal profiles of 6252 proteins and 15,059 phosphorylation sites suggested at least two distinct signaling waves: one peaking within 30 to 60 min after stimulation and a second upsurge after 24 h. In addition to providing a comprehensive view of the proteome and phosphoproteome dynamics during early MSCs differentiation, our analyses identified a key role of serine/threonine protein kinase DI (PRKDI) in OB commitment. At the onset of OB differentiation, PRKDI initiates activation of the pro-osteogenic transcription factor RUNX2 by triggering phosphorylation and nuclear exclusion of the histone deacetylase HDAC7.

[Supplemental material is available for this article.]

The skeleton is a dynamic tissue constantly undergoing bone matrix resorption and formation, two coupled processes mediated by osteoclastic and osteoblastic cells, respectively (Parfitt et al. 2011). Decreased osteoblastic bone formation relative to osteoclastic bone resorption is the cellular basis of age- and osteoporosis-related bone loss (Feng and McDonald 2011). Osteoblasts differentiate from skeletal stem cells, also known as stromal or mesenchymal stem cells (MSCs), which are present within the bone marrow stroma. MSCs are among the most suitable cell types for cell therapy applications due to their ease of isolation, multilineage differentiation potential, their immunomodulatory and regenerative properties, as well as their excellent safety record (Jafari et al. 2014; Zaher et al. 2014). Therefore, MSCs have been used in an increasing number of clinical trials for enhancing tissue repair in different contexts such as regeneration of skeletal defects (Andrzejewska et al. 2019). Determining the signaling mechanisms that regulate MSCs commitment and differentiation into the bone-forming osteoblastic cells can pave way for the development of novel approaches for enhancing bone formation and treating osteoporotic bone loss (Abdallah et al. 2015).

The current model of osteoblast (OB) differentiation from stem cells is based on a large number of in vitro studies and consists of a progressive differentiation sequence that includes phases of cell proliferation followed by cell maturation, extracellular matrix production and, lastly, matrix mineralization (Lian and Stein 1995). Although this model is useful as an overview, there is a need for delineating the molecular and signaling events underlying these phases. Several previous studies have used a reductionist approach to determine factors and signaling pathways that are important for OB differentiation and identified a number of key signaling cascades including Wnt, Hedgehog, Notch, and bone morphogenetic proteins (BMPs) (Abdallah et al. 2015; Hojo et al. 2015). However, such approaches fail to provide information regarding how signaling events are initiated and integrated in a coherent, well-orchestrated biological process.

Over the last decade, mass spectrometry (MS)-based proteomics has evolved as a powerful platform for studying complex biological processes at a systemic level. We and others have previously used MS-based quantitative proteomics using stable isotope labeling by amino acids in cell culture (SILAC) to provide a global overview of human embryonic stem cell proteomes and phosphoproteomes during differentiation (Prokhorova et al. 2009; Van Hoof et al. 2009; Rigbolt et al. 2011a; Rigbolt and Blagoev 2012), to monitor receptor tyrosine kinase signaling in MSCs differentiation to osteoblasts (Kratchmarova et al. 2005), and to determine secreted factors during OB differentiation (Kristensen et al. 2012), as well as proteome and secretome variations during muscle cell differentiation (Henningsen et al. 2010; Le Bihan et al. 2015). Here, we combined SILAC-based quantitative proteomic and phosphoproteomic analyses to elucidate the signaling events occurring during the first 24 h of OB differentiation of...
human MSCs, identifying more than 6000 proteins and 15,000 phosphorylation sites.

Results
Temporal profiling of the proteome and phosphoproteome during hMSC commitment to osteoblastic lineage

SILAC-labeled hMSCs were either left unstimulated or induced to undergo OB differentiation for 30 min, 1, 6, or 24 h (Fig. 1A). Cell lysates were separated into nuclear and cytoplasmic fractions and each combined into two different SILAC pools allowing comparable quantitation of proteins and phosphorylation events at the five time points. Cell lysates were either subjected to 1D gel electrophoresis followed by in-gel digestion for proteome determination or to enrichment of phosphopeptides before LC-MS/MS measurements for delineation of phosphorylation events at a site-specific level (Fig. 1A).

We identified 6252 proteins (Supplemental Table S1), of which 565 showed changes in at least one time point ($P < 0.05$ with at least 1.5-fold increase or decrease). The number of regulated proteins increased from an average of 250 proteins during the first 6 h to nearly 350 at 24 h (Fig. 1B). The phosphoproteomic analysis identified 15,059 unique phosphorylation sites with an overall distribution of 88% on serine, 10% on threonine, and 2% on tyrosine residues (Supplemental Table S2). Quantitative information was obtained for 10,930 sites, of which 797 displayed significant changes in abundance in at least one time point ($P < 0.05$ with at least twofold increase or decrease compared to unstimulated cells). In contrast to the observed changes in protein abundances, the highest number of altered phosphorylation sites was observed shortly after induction of differentiation, followed by a drop at 6 h, before a secondary boost at 24 h, hinting to the presence of two signaling waves during early OB differentiation (Fig. 1B).

The proteomic landscape changed within the 24 h after induction of OB differentiation

To examine the dynamics of protein expression at a global level, we used unsupervised fuzzy c-means clustering (Supplemental Fig. S1A). Because the majority of the protein alterations occurred at later time points, we focused on the data from the 6- and 24-h time points. Proteins grouped into six clusters with a general down-regulation (Clusters 1 and 3), up-regulation (Clusters 2 and 6), or transient changes (Clusters 4 and 5) (Supplemental Fig. S1A). Analyses using Gene Ontology (GO) (Ashburner et al. 2000) revealed enrichment for several biological processes and molecular functions in each individual cluster (Supplemental Fig. S1B,C). We found enrichment in different processes related to “extracellular matrix” in Clusters 1 and 2 that represent early responsive proteins. Proteins related to “collagen metabolism” and “extracellular matrix disassembly” were found in Cluster 1 (down-regulated at 6 h and 24 h after induction), whereas proteins associated with “extracellular matrix organization” were found in Cluster 2 (up-regulated 6 h and 24 h after the induction). Proteins involved in “DNA replication and regulation of cell proliferation” were also enriched in Cluster 2, suggesting that the initial phase of MSC osteoblast differentiation is associated with enhanced cell division, which is in accordance with previous findings (Lian and Stein 1995; Marumoto et al. 2017).

Some of the regulated proteins showed significant changes only 24 h after the onset of differentiation (Cluster 6). Several of the proteins in this cluster were associated with the transforming growth factor beta (TGFβ) and Hippo pathways indicating increased contribution of these signaling networks in the differentiation process at this later time point. A significant fraction of the proteins underwent only a temporal change in their expression (Clusters 4 and 5). Proteins linked with Toll-like receptor signaling were overrepresented in Cluster 4, characterized by a transient decrease, whereas proteins involved in “transcriptional regulation” were enriched in the group with temporary elevated levels (Cluster 5).

Because transcription factors play pivotal roles in cellular differentiation responses, we extracted and quantified transcription factors present within the identified protein clusters. We found 140 transcription factors with DNA-binding motifs according to their GO terms (Fig. 2), of which 32 showed temporal changes. Among these, 22 were regulated at either protein or phosphosite level and were linked with terms related to cell differentiation, bone mineral density, and ossification (Supplemental Fig. S2). For example, CCAAT enhancer binding protein beta (CEBPβ) and delta (CEBPΔ), which are members of the basic leucine zipper (bZIP) family, were up-regulated at the 6- and 24-h time points. The two transcription factors are known to interact and positively regulate
RUNX family transcription factor 2 (RUNX2)-mediated gene expression (Gutierrez et al. 2002). Fork-head transcription factors were in general down-regulated with the exception of FOXO3, which was slightly up-regulated at 6 and 24 h in agreement with a role of this factor in protecting osteoblasts from oxidative stress and apoptosis (Ambrogini et al. 2010). Furthermore, the zinc-finger proteins ZFP36L1 and ZFP36L2 were also found increased at the protein level (Fig. 2). These two factors are known to be up-regulated in response to several extracellular signals such as TGFB, TNFA, EGF, and glucocorticoids. Both have been reported to play important roles in cell differentiation (Stumpo et al. 2009; Chen et al. 2015; Tseng et al. 2017) and, for ZFP36L1, to favor OB differentiation over adipogenesis (Tseng et al. 2017).

Changes in the phosphoproteome revealed two intracellular kinase-signaling waves associated with OB lineage commitment of hMSC

Various members from most of the kinase families were represented in our data set, covering 187 quantified kinases: 30 were altered at the protein level, and 45 showed changes in their phosphorylation status (Fig. 3). To assess the importance of the identified fluctuations, we selected 10 kinases hitherto not associated with osteoblast differentiation and examined the effect of their individual knockdown on the conversion of hMSCs into osteoblasts. Of these kinases, eight were regulated at their level of expression and two at their phosphorylation status. Knockdown of seven of the 10 candidates had a significant impact on alkaline phosphatase activity at day 6 of differentiation (Fig. 4). Moreover, silencing of the kinases identified as possible negative regulators by quantitative proteomics, that is, down-regulated in the MS data set, led to stimulation of osteoblast conversion, whereas knockdown of the potential positive regulators resulted in reduced differentiation.

We next extracted overrepresented linear phosphorylation motifs from the identified phosphorylation sites and possible kinases capable of phosphorylating these motifs (Fig. 5A, left). The predictions were then correlated with the quantified dynamics of the corresponding phosphorylation sites (Fig. 5A, right). This depiction allows identification of kinases affiliated with highly regulated motifs and their dynamic changes during OB differentiation. The prediction identified the mitogen-activated protein kinase 14 (MAPK14, also known as p38) as a putative kinase for many of the regulated sites (Fig. 5A) and indeed MAPK14 is an important known regulator of OB differentiation (Rodríguez-Carballo et al. 2016). However, other MAPK members could not be excluded as possible candidates because the phosphorylation motif is shared among them. We identified activating phosphorylation sites on two of these kinases, MAPK3 and MAPK1 (also known as ERK1 and ERK2, respectively), both peaking at 30 min (Supplemental Table S2). Phosphorylation of MAPK14 itself was not detected in

Figure 2. The levels of several transcription factors are altered during early osteoblast commitment. Identified and quantified transcription factors were ordered by the phylogenetic distance of their DNA-binding domains and aligned using iTOL (Letunic and Bork 2019). Histogram in the outer ring represents the regulation at protein level during induction of osteoblastic differentiation. Color codes for the different families are shown on the right.
the MS analyses, but western blot analysis revealed that, in contrast to ERKs, MAPK14 maintained high levels of activation even at 1 h (Supplemental Fig. S3A).

To obtain further insights into the involvement of MAPKs, we extracted the 1287 predicted phosphorylation sites for the MAPK family present in our data set, clustered them according to their dynamic changes, and assessed for overrepresented biological processes in the corresponding clusters (Supplemental Fig. S3B,C). We observed that sites correlating with the dynamics of MAPK14 activation, that is, with increased phosphorylation at the first two time points after stimulation and peaking at 1 h (Cluster 3), were enriched in categories associated with OB lineage commitment, for example, "osteoblast development" and "positive regulation of osteoblast differentiation." This was not the case for proteins belonging to clusters that resemble the activation profile of the ERKs with its maximum effect after 30 min. This argues that despite several of the MAPK family members were activated, MAPK14 was crucial in mediating the early steps in OB conversion of hMSCs. In this Cluster 3, we also identified other terms associated with MAPK14 activity such as "transcription from RNA polymerase II promoter" and "positive regulation of transcription from RNA polymerase II promoter," including sites Ser320 in FOSL2, Ser666 in HCFC1, Ser90 and Ser100 in JUND, Ser250 in NFIX, Ser727 in STAT1, and Ser371 in TP53BP1. The recruitment of RNA polymerase II complex to regulate transcription requires MAPK14 activity (Ferreiro et al. 2010); furthermore, this process has also been related with the chromatin remodeling necessary for OB differentiation (Villagrá et al. 2006).

Besides MAPK14, we identified other kinases possibly involved in the regulation of commitment to osteoblastic lineage. Several putative sites for AKT protein kinases and their downstream targets, RPS6KB1 (ribosomal protein S6 kinase B1) and RPS6KB2, were observed to be regulated immediately following OB induction. Conversely, the promiscuous casein kinase 2 alpha catalytic subunits (CSNK2A1–3) were found to be a possible kinase for many of the overrepresented but not regulated sites (Fig. 5A).
As a complementary approach, we grouped regulated phosphorylation sites into eight clusters using fuzzy c-means (Fig. 5B). Although some phosphorylations were only affected at 24 h (Cluster 2), most sites underwent changes during earlier time points and returned close to baseline levels (Clusters 3–8), indicating temporal and stage-specific effects. We then separately partitioned KEGG pathways and kinase substrate enrichment analyses to explore for discrete signaling pathways and individual kinases, respectively, with significant overrepresentation within any of the clusters. We observed enrichment of several receptor tyrosine kinase (RTK) signaling pathways, for example, EGF, VEGF, and insulin in Clusters 6–8 (Supplemental Table S3). In agreement with the preceding findings arguing for an involvement of the MAPks and the placement of these kinases as nodal points in RTK networks, MAPK signaling was also tightly associated with the same clusters (Supplemental Table S3). Besides MAPK, other signaling cascades known to be involved in OB differentiation, like MTOR and FOXO pathways, were enriched in these clusters as well (Teixeira et al. 2010; Pramojanane et al. 2014; Dai et al. 2017). An additional set of sites that showed increased phosphorylation at the onset of differentiation (Cluster 8) were affiliated with another nodal point kinase family, AKT (Fig. 5B), reflecting the role of the AKT signaling in the initial phase of OB differentiation. Collectively, Clusters 6–8 could therefore be assigned to a first wave of phosphorylation events and appear to be a compilation of several signaling cascades.

Regarding the pathways associated with the second activation wave, sites targeted by group II of the PAK (p21 [RAC1]-activated kinase) family of kinases (PAK4, PAK5, and PAK6) were overrepresented in Cluster 2 containing sites with up-regulation at 24 h with little or no fluctuations at earlier time points (Fig. 5B). Although this group of kinases has been implicated in several biological processes and tumor formation, its role in OB biology remains unexplored (Wells and Jones 2010). Of note, proteins involved in Hippo and TGFβ signaling were up-regulated at their protein levels at this latest time point (Supplemental Fig. S1), arguing for an involvement of these signaling pathways in the second kinase activation wave as well. A pattern therefore emerges, in which several intertwined kinase-signaling pathways commonly initiated by RTKs are activated, and their downstream effectors MAPK and AKT act as the radial points during the first wave. As these routes clung off, a more defined program is initiated, involving PAK, TGFβ, and Hippo signaling networks.

**PRKDI regulates osteoblast differentiation via HDAC7 phosphorylation**

From the kinase prediction analyses, serine/threonine protein kinase D1 (PRKDI) showed the highest enrichment with substrates predominantly found in Clusters 1 and 7 (Fig. 5B). In accordance, Western blotting with phospho-specific antibody revealed that the activating Ser910 phosphorylation on PRKDI was indeed augmented 30 min after induction and declined to baseline levels at the 6- and 24-h time points (Fig. 6A; Supplemental Fig. S4A). We therefore assessed the involvement of PRKDI in the commitment of hMSCs to osteoblastic cells using a PRKDI-specific chemical inhibitor, CRT0066101 (Harikumar et al. 2010). The inhibition of PRKDI had only a mild effect on the early phosphorylation status of MAPK14, a canonical target of PRKDI (Supplemental Fig. S3A). We also examined the effect of stimulating hMSCs with phorbol myristate acetate (PMA), which is a potent activator of several members of the PKC family of kinases, including PRKDI. Despite the very strong activation of PRKDI (Fig. 6B; Supplemental Fig. S4B), the PMA treatment of the cells did not augment the phosphorylation of MAPK14 (Supplemental Fig. S3A), further indicating that MAPK14 is not the main target of PRKDI in the initial stages of OB lineage commitment.

It has previously been shown that PRKDI regulates the cytoplasmic-nuclear shuttling of members of the class Ila histone deacetylases and transcriptional repressors (HDAC4, −5 and −7) by direct phosphorylation (Dequiedt et al. 2005; Parra et al. 2005; Sinnett-Smith et al. 2014). We therefore hypothesized that PRKDI may regulate OB differentiation through phosphorylation and relocalization of one or more of the HDACs. Temporal profile of the phosphorylation status of the residue in the three HDACs (HDAC4-Ser632/HDAC5-Ser661/HDAC7-Ser486) controlling for their nuclear-cytoplasmic shuttling and targeted by PRKDI, was virtually identical to the activation pattern of the kinase (Fig. 6A; Supplemental Fig. S4A). In support of PRKDI being responsible for this phosphorylation, CRT0066101 treatment resulted in decreased HDAC phosphorylation, whereas activation of PRKDI by PMA increased the phosphorylation of HDACs (Fig. 6B; Supplemental Fig. S4B). Although the phospho-specific antibody recognizes all three HDACs when phosphorylated at these positions, the phosphorylation dynamics of HDAC4 Ser632 and HDAC5 Ser661 were quantified in the MS analyses and neither showed significant changes (Supplemental Table S2), indicating that the PRKDI effect is mediated mainly through phosphorylation of HDAC7 Ser486. Because this phosphorylation site is known to regulate HDAC cellular localization (Parra et al. 2005), we assessed the effect of the PRKDI inhibitor on the distribution of HDAC7 in hMSCs. Upon induction of OB differentiation, HDAC7 was translocated from the nucleus to the cytoplasm in vehicle treated cells. This relocalization was completely prevented by the PRKDI inhibitor (Fig. 6C; Supplemental Fig. S4C), suggesting that PRKDI regulates OB commitment of hMSCs through phosphorylation and nuclear exclusion of the RUNX2-repressor HDAC7. In support, the mRNA levels of four RUNX2 target genes and early osteoblast gene markers were all significantly decreased by the PRKDI inhibitor (Fig. 6D). In addition, siRNA-mediated...
Figure 5. Motif analyses revealed kinases with likely involvement in osteoblastogenesis. (A) Significantly overrepresented linear phosphorylation motifs were identified with Motif X and matched to kinases predicted with NetworKIN (Linding et al. 2008), in which an increase in purple intensity equals an increase in the percentage of the given site that is predicted to be phosphorylatable by the given kinase (left). The identified motifs were hierarchically clustered according to the average phosphorylation dynamics of the sites matching each motif. The average change in phosphorylation of sites matching the motif in each time point is color-coded blue for decreased or red for increased compared to undifferentiated cells. Only average phosphorylation changes with a mean significantly different from zero on confidence level 0.01 were included. (B) Class I sites were clustered according to their temporal dynamics (left). Fisher’s exact test with Benjamini–Hochberg adjustment was used to identify kinases with significant overrepresentation of substrates within any of the clusters (right).
HDAC and PRKD1 phosphorylation levels during osteoblast differentiation, repress other bZIPs and especially MAF (Kataoka et al. 1996), which restricts OB differentiation because it has the ability to bind and regulate at all time points (Fig. 2). It is possible that MAFK devoid of a transactivating domain, namely MAFK, was downregulated by siRNA targeting PRKD1 (siPRKD1 #1, siPRKD1 #2). For D and E: (*) P < 0.05, Student’s t-test; n = 3.

knockdown of PRKD1 in hMSCs resulted in reduced alkaline phosphatase activity on day 6 of differentiation (Fig. 6E), further indicating the functional importance of PRKD1 for the OB differentiation.

Discussion

Lineage commitment and differentiation of hMSCs into osteoblastic cells is a complex cellular process. Our analyses provide a first global view of the complex proteomic and phosphoproteomic changes during the initial 24 h of hMSCs transition to the OB lineage. We identified 6252 proteins and 15,059 unique phosphorylation sites during the early steps of OB conversion. The data exposed that osteoblast commitment involves complex makeover already at the first 24 h, including changes in phosphoproteins associated with cell proliferation, extracellular matrix, and vascular remodeling, all known to be important for efficient OB differentiation (Lian and Stein 1995).

By modulating gene expression, transcription factors are responsible for a large part of the dynamics in the proteome. We found that the expression of several members of the bZIP family was increased during OB differentiation, arguing for a strong involvement of these transcription factors in regulating OB commitment. Both CEBPB and CEBPD were up-regulated, and both have previously been shown to be involved in osteoblastic lineage commitment through cooperative actions with RUNX2 (Gutierrez et al. 2002; Tominaga et al. 2008). In contrast, another bZIP factor devoid of a transactivating domain, namely MAFK, was downregulated at all time points (Fig. 2). It is possible that MAFK restricts OB differentiation because it has the ability to bind and repress other bZIPS and especially MAF (Kataoka et al. 1996), which plays an essential role in OB development through interplay with RUNX2 (Nishikawa et al. 2010). Moreover, we observed a strong temporal inhibition of the NF-kB transcription factors and Rho activity (Supplemental Fig. S1). These observations are in agreement with previous data showing a strong inhibitory effect on OB differentiation by the two signaling pathways (Novack 2011; Strzelecka-Kilisek et al. 2017).

The dynamics of regulated phosphorylation sites and protein expression changes indicated the existence of two consecutive signaling waves, the first peaking immediately after induction of differentiation followed by a second commencing 24 h after initiation (Supplemental Fig. S5). During the first wave, a myriad of signaling cascades are activated, including a number of RTK pathways. The exact contribution of each signaling component awaits further analyses, but it is likely that their mutual crosstalk is of pivotal importance for a balanced and timely initiation of OB differentiation. This is exemplified by the fact that several of the phosphorylated proteins are embedded in more than one pathway. RAF1 and TSC2, when phosphorylated on Ser29 and Ser1319, respectively, are both key nodes in interaction between MTOR and MAPK pathways (Mendoza et al. 2011). Other sites such as Ser2 in RRAGC or Ser226 in MAP2K2 are known to have the potential to activate each of the pathways independently. Both pathways can converge on the same transcription factors such as FOXO (Mendoza et al. 2011), whose localization is also affected by STAT3 (Oh et al. 2012), which we find phosphorylated at Ser727 as well (Supplemental Table S2). Collectively, numerous pathways with several shared nodal points are activated at this first wave during differentiation.

The second wave of signaling appears to rely on a completely different set of signaling cascades (Supplemental Fig. S5). Of the lesser studied but highly enriched pathways in our data, the up-regulation of the Hippo pathway suggests strong involvement of this cascade 24 h after induction of differentiation (Lee et al. 2015). This concurs with an up-regulation of the expression of transforming growth factor beta receptor 2 (TGFB2), arguing for crosstalk between these two pathways in the OB commitment as they also cooperate in other biological processes (Luo 2017). KEGG-based enrichment analysis of phosphorylation sites augmented at this later stage suggested activation of the DNA repair machinery as well (Supplemental Table S5). The DNA damage response was previously associated with differentiation processes (Sherman et al. 2011), arguing that maintaining genomic integrity is also essential for the progress of OB conversion. In our data we found XPC, known member of pluripotency-maintaining complex (Cattoglio et al. 2015), to be phosphorylated at Ser883 and Ser884. The LIG1 and Lig3 proteins, reported to be involved in human neural crest stem cell differentiation (Newman et al. 2017), were phosphorylated at Ser51 and Thr244, respectively, as well (Supplemental Table S2). Moreover, substrates for the group II of the PAK kinases were increasingly phosphorylated at the 24 h...
time point. These kinases are at least partially activated by the small GTPases, CDC42 and RAC1. CDC42 has been shown to induce chondrogenesis, but not OB differentiation, downstream from TGFβ stimulation through the group I PAK member PAK1 and MAPK14 (Wang et al. 2016). It is therefore conceivable that the diverging groups of this family of kinases integrates upstream signaling events and balances osteoblast versus chondrocyte lineage commitment.

As a follow up on the global changes during OB differentiation, we identified a mechanistic involvement of PRKD1. At first, two complementary approaches that predict kinases responsible for the dynamic changes in the phosphoproteome at the onset of OB differentiation revealed pivotal roles for MAPK14 and PRKD1 kinases. Members of the MAPK family, and notably MAPK14, are well known for their involvement in OB lineage commitment (Rodríguez-Carballo et al. 2016). PRKD1 has also at several instances been described as a key mediator of the bone morphogenetic protein (BMP)-induced OB differentiation, mainly through activation of the MAPK14 (Lemonnier et al. 2004; Celli and Campbell 2005; Kim et al. 2011). In our study, the addition of a PRKD1 inhibitor did not affect the level of the activating phosphorylation of MAPK14, suggesting independent effects of the two kinases within the first hours of OB differentiation. In agreement with a previous study focused on BMP signaling (Jensen et al. 2009), we observed a PRKD1-dependent nuclear exclusion of the RUNX2-repressive HDAC7. It is therefore likely that part of the stimulatory effect of PRKD1 on OB differentiation is mediated through the relief of RUNX2 from HDAC7-mediated repression.

Our study shows that lineage commitment of hMSCs can be conceptualized as coordinated successive waves of interacting signaling pathways, where the “choir is more important than the solo” (Kassem and Bianco 2015). From a translation perspective, our findings suggest a possibility for a more flexible, multiple protein targeting approach to better control hMSCs lineage commitment to osteoblastic cells and bone regeneration, an important application in regenerative medicine.

**Methods**

**Cell culture, SILAC labeling, and differentiation**

As a model for human primary MSCs, we used hMSC-TERT cell line that has been immortalized by ectopic expression of telomerase reverse transcriptase gene (hTERT) (Simonsen et al. 2002). For simplicity, we will refer to the hMSC-TERT as hMSCs throughout the paper. For SILAC experiments, cells were labeled with L-arginine (Arg0) and L-lysine (Lys0) for the unstimulated cells, L-arginine-13C6 Arg6 and L-lysine-2H4 Lys4 for the 30-min and 1-h conditions, or L-arginine-13C6-15N4 (Arg10) and L-lysine-13C6-15N2 (Lys8) for the 6- and 24-h conditions (Cambridge Isotope Laboratories) at concentrations of 28 and 42 mg/L, respectively. Differentiation of hMSCs cells into the osteoblastic lineage was performed as described previously (Kratchmarowa et al. 2005). For PRKD1 inhibition, CRTC066101 (Abcam) was added 30 min before the induction of OB differentiation at a final concentration of 10 µM. For PRKD1 activation, phorbol myristate acetate (PMA) (Sigma-Aldrich) was used at a final concentration of 100 ng/mL.

**Sample processing for LC-MS/MS analysis**

Two triple-encoding SILAC experiments were performed in parallel for each replicate (total of three biological replicates). Cells were stimulated for 30 min, 1 h, 6 h, or 24 h, rinsed twice with cold PBS and lysed in 50 mM Tris, pH 7.5, 150 mM NaCl, 1% NP-40, 0.1% sodium deoxycholate, and 1 mM EDTA supplemented with phosphatase and protease inhibitors (Roche). Nuclei were pelleted by centrifugation for 10 min at 10,000g at 4°C. Cytoplasmic proteins in the supernatant were precipitated using aceton and resuspended in 10 mM Tris, pH 7.5, with 6 M urea and 2 M thiourea. The same buffer, supplemented with Benzonase, was used for resuspension of nuclear proteins. Protein concentrations were determined using Bradford (Bio-Rad) according to the manufacturer’s instructions. Samples were combined in a 1:1:1 ratio (L:M:H). Proteins were reduced with 10 mM dithiothreitol and alkylated with 55 mM iodoacetamide. For protein identification, 100 µg protein of each SILAC pool were separated using 1D SDS-polyacrylamide gel electrophoresis on NuPAGE Novex gels (Life Technologies). The gel was stained with Colloidal Blue (Life Technologies), cut into pieces, and digested with trypsin (Promega) overnight at 37°C. Resulting peptide mixture was extracted and desalted using modified Stage tips as described elsewhere (Akimov et al. 2011). The remainder of the protein lysates was used for phosphopeptide enrichment before MS analysis. Proteins were digested in solution for 4 h at room temperature with LysC (Wako Chemicals) followed by trypsin overnight (Akimov et al. 2018), used in a ratio of 1:160 and 1:200, respectively, to relative protein content. Peptide solutions were acidified with trifluoroacetic acid to a final concentration of 0.1% and cleared by centrifugation (20,000g, 5 min). Phosphopeptides were enriched and fractionated using TiO2 as described earlier (Rigbolt et al. 2011a).

**LC/MS/MS analysis**

Peptides were separated using an in-house packed fused silica column with a length of 15 cm, an inner diameter of 75 µm and reverse phase material (3 µm C18, ReproSil-Pur C18 AQ) (Dr. Maisch HPLC GmbH, Ammerbuch Germany). EASY-nLC (Proxeon), Agilent 1100, or Agilent 1200 systems were used, generating linear gradients of 90 min (solvent A was 0.5% acetic acid in water, and solvent B was 80% Acetonitrile and 0.5% acetic acid) at a constant flow of 250 nL/min. Mass spectrometry was performed in positive ion mode using LTQ-Orbitrap XL for total protein coverage (data-dependent acquisition for isolation and fragmentation of the 10 most intense ions, CID fragmentation) and LTQ-Orbitrap Velos and Q-Exactive for phospho-enriched samples (data-dependent acquisition for isolation and fragmentation of the 10 most intense ions for Velos and the 12 most intense for Q-Exactive, HCD fragmentation) (all from Thermo Fisher Scientific).

Data analysis is described in detail in Supplemental Methods. Briefly, all MS raw files from the three biological replicates were processed together using the MaxQuant with FDR of 1% set for both protein and site identification. Protein with shared peptides were combined in protein groups, and both unique and razor peptides were used for quantitation.

**Clustering of protein and sites based on their temporal profiles**

Phosphosite SILAC ratios were normalized where possible by dividing the phosphosite ratio with total protein ratio to ensure that changes were caused by the dynamics of the site and not the protein. Resulting values were log2 transformed and subjected to unsupervised clustering with the fuzzy c-means algorithm implemented in the GProX software (Rigbolt et al. 2011b). Only sites with a complete profile, $P < 0.05$ (Significance A, MaxQuant) and a ratio width of at least a twofold increase or decrease in at least one time point were considered. Enrichment analysis for kinase substrates within the clusters were performed using Fisher’s exact test with
Benjamins-Hochberg adjustment for multiple testing (Benjamini and Hochberg 1995). Protein ratios were subjected to the same method, designating as changing proteins with $P < 0.05$ (Significance A, MaxQuant) with at least 1.5-fold increase or decrease in at least one point. Details for additional bioinformatic analyses are provided in Supplemental Methods.

Immunocytochemistry and western blotting

Immunocytochemistry was performed as previously described (Sanchez-Quiles et al. 2017). Briefly, cells were fixed with 4% paraformaldehyde, permeabilized with 0.1% Triton X-100, and blocked with horse serum. Cells were incubated with mouse anti-HDAC7 antibody (Santa Cruz Biotechnology sc-74563), rinsed, incubated with anti-mouse/Cy3 secondary antibody (Jackson Immunoresearch 715-165-150) plus DAPI (Sigma-Aldrich), and rinsed before mounted on cover slides. Images were acquired using an Observer Z1 microscope (Zeiss). See Supplemental Methods for western blotting details.

Quantitative real-time polymerase chain reaction (qRT-PCR) and siRNA transfections

Relative mRNA levels were measured by qRT-PCR as previously described (Hallenborg et al. 2016; Supplemental Methods). A reverse transcription protocol using Lipofectamine 2000 (Invitrogen) was used for the siRNA transfections. hMSC cultures were trypsinized, and 18,000 cells/cm² were reverse-transfected with two independent siRNAs targeting the kinases (Supplemental Methods) in two biological replicates for each siRNA. For PRKD1, the experiment was performed in triplicate for each of its siRNAs.

Quantitative of alkaline phosphatase (ALP) activity

To quantify alkaline phosphatase activity, p-nitrophenyl phosphate (Fluuka) was used as substrate and to correct for differences in cell number, ALP activity data were normalized to cell viability measurements as described previously (Jafari et al. 2017). Briefly, CellTiter-Blue reagent (Promega) was added to the normal culture medium, incubated for 1 h at 37°C, and fluorescent intensity (560EX/590EM) was measured using FLUOstar Omega multimode microplate reader. Cells were then washed with Tris-buffered saline (TBS), fixed in 3.7% formaldehyde, 90% ethanol at room temperature for 30 sec, incubated with substrate (1 mg/mL of p-nitrophenyl phosphate in 50 mM NaHCO3, pH 9.6, and 1 mM MgCl2) for 20 min at 37°C, and the absorbance measured at 405 nm, using FLUOstar Omega multimode microplate reader.

Data access

The mass spectrometric data generated in this study have been submitted to the ProteomeXchange Consortium (http://proteomecentral.proteomexchange.org) via the PRIDE partner repository with the data set identifier PXD011001.

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