



Bartolozzi, A., Viti, F., De Stefano, S., Sbrana, F., Petecchia, L., Gavazzo, P. and Vassalli, M. (2020) Development of label-free biophysical markers in osteogenic maturation. *Journal of the Mechanical Behavior of Biomedical Materials*, 103, 103581. (doi: [10.1016/j.jmbbm.2019.103581](https://doi.org/10.1016/j.jmbbm.2019.103581))

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# Development of label-free biophysical markers in osteogenic maturation

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## Abstract

The spatial and temporal changes of morphological and mechanical properties of living cells reflect complex functionally-associated processes. Monitoring these modifications could provide a direct information on the cellular functional state. Here we present an integrated biophysical approach to the quantification of the morphological and mechanical phenotype of single cells along a maturation pathway. Specifically, quantitative phase microscopy and single cell biomechanical testing were applied to the characterization of the maturation of human foetal osteoblasts, demonstrating the ability to identify effective label-free biomarkers along this fundamental biological process.

## Keywords

Human foetal osteoblasts; cell morphology characterization; cell mechanics investigation

## Introduction

Label-free approaches in tissues, cells and molecular biology are nowadays particularly appealing, as testified by many recent publications (Ray and Steckl 2019, Golfier et al. 2017, Stern et al. 2009, Chang et al. 2017), due to their minimal invasiveness, maintenance of living cells (being therefore available for reuse and time series experimental designs), user-friendliness and rapidity. To such an extent, the analysis of mechanical and morphological features of single cells is gaining more and more relevance, having shown a tight correlation with the functional state of the cell in several conditions (Chen et al. 2016; Muthukumaran et al. 2012). For example, it is well known that cell mechanical behaviour is related to cancer staging (Kawano et al. 2015, Fortier et al. 2016), metastatic condition (Kumar and Weaver 2009, Brodland and Veldhuis 2012) and differentiation (Engler et al. 2006). Analogously, cell morphometric phenotyping has been fruitfully exploited, among the others, in cancers (Wu et al. 2015), blood pathologies (Ford 2013), and cellular transdifferentiation evaluation (Petecchia et al. 2016).

Over the past decade, many strategies and techniques have been proposed to support the characterization of biological samples down to the single cell level, either from the mechanical or morphological point of view. Among proposed approaches (Darling and Di Carlo 2015), Atomic

Force Microscopy (AFM) is certainly the most adopted and effective to identify mechanical signatures of cellular systems (Lulevich et al. 2006). In this paper, a similar approach is proposed, based on a recent implementation of the cantilever-based force measuring method used in AFM, but exploiting an innovative in-fiber detection method (Chavan et al. 2012). This device, nowadays converted in a commercial product (Piuma Chiaro, Optics11, The Netherland), provides much higher versatility and throughput measurements than standard AFM, still retaining the force resolution required to finely assess mechanical properties of single cells.

For what concerns the assessment of morphological features, most of the commonly exploited techniques foresee cells treatment and labelling (Monteiro et al. 2015), while white-light phase contrast microscopy requires complex and cumbersome image processing steps to provide a sensible quantitative description of the cell morphology (Ambühl et al. 2012). To overcome these intrinsic limitations of standard approaches, label-free quantitative phase imaging has emerged as a potential tool for efficient cell morphometry, also providing the possibility of cells reuse for successive evaluations (Zhang et al. 2018). In this paper, morphological aspects of living cells were evaluated qualitatively and quantitatively, respectively, through a non-interferometric approach based on the transport-of-intensity equation (TIE, Teague 1983) and using digital holography (Gabor, 1948).

Ossification is a complex and delicate physiological process, whose impairment might lead to the onset of potentially severe disorders, such as calcification of heart valve or osteoporosis (Rutkovskiy et al. 2016). Moreover, bone is a tissue intrinsically subjected to wide mechanical stimuli which represent the driving force of its continuous remodelling (Thompson et al. 2012) and single cells involved in this process are expected to experience a drastic change in the mechanical and morphological phenotype during maturation. Here this aspect was addressed focusing on a specific cell line, hFOB 1.19 (ATCC® CRL11372™ from ATCC, Manassas, VA, USA, from now on referred to as hFOB). hFOB is an immortalized cell line of human foetal osteoblasts from limb tissue of a spontaneous miscarriage, transfected with the temperature sensitive expression vector pUCSVtsA58 which alternatively enables cell proliferation or differentiation according to culture temperature (respectively 33.5°C and 39.5°C) (Harris et al. 1995). According to Yen et al. (2007) hFOB cells show multilineage differentiation potential and share several surface markers with stem cells from the stroma of the bone marrow. Due to their potential, hFOB cell line is commonly used as a model system for studying human osteoblast maturation, osteoblast physiology, and effects of biomolecular and chemical factors on osteoblast function and maturation. In the described work, hFOB was considered as the model for identifying novel label-free biomarkers to monitor the transition between immature and mature osteoblasts, therefore representing valuable indicators of the osteoblasts developmental stage.

## **Materials and methods**

### ***Cell culture***

hFOB 1.19 cells are foetal osteoblasts taken from limb tissue and therefore already in a specific developmental pathway. Nevertheless hFOB cells show a multilineage differentiation potential and share several surface markers with stem cells from the stroma of the bone marrow (Yen et al. 2007). Being immortalized with a temperature-sensitive SV40 large T antigen, this cell line is able to proceed towards cell proliferation when cultured at 33.5°C in a humidified atmosphere of 5% CO<sub>2</sub>, or cell differentiation towards mature osteoblasts when grown at 39.5°C, without the need to add any chemical supplement in the cell culture medium. hFOB cell line was cultured in accordance with ATCC recommendations. The base medium was a 1:1 mixture of Ham's F12 Medium Dulbecco's Modified Eagle's Medium, (DMEM-Ham's F-12, GIBCO), with 2.5mM L-glutamine (Sigma-Aldrich, St Louis, MO, USA), without phenol red. To complete the growth medium, G418 (0.3 mg/mL) and 10% foetal bovine serum (FBS) (Sigma-Aldrich) were added to the base medium. Medium was changed every three days. For all experiments, the same passage, from 2 to 10, was used.

### ***Experimental design***

hFOB were monitored in the following conditions, each considered in at least three replicates: during the proliferation, with cells growing at 33.5°C, (case A); during maturation toward the osteoblastic phenotype, moving cells at 39.5°C for 15 days, and performing functional and morpho-mechanical tests at day 5 (case B), day 11 (case C), and day 15 (case D). The actual maturation of the cell line was assessed through several traditional indicators of the osteoblastic phenotype, and on top of this, morpho-mechanical experiments were carried out. In detail, for each experimental condition, the following readouts were evaluated: functional information related to osteogenic maturation towards bone phenotype including (1) *calcium deposition*, assessed by Alizarin Red S (Alizarin), (2) *osteocalcin expression*, evaluated through confocal microscopy (CM), and (3) *gene targets expression*, defined by means of qPCR; *quantitative and qualitative cell morphology*, assessed respectively by (4) digital holographic microscopy (DH) and (5) quantitative phase imaging based on transport of intensity equation resolution (TIE), to shed light on the morphological changes consequent to experimental conditions; and (6) *mechanical properties*, by single cell Chiaro nanoindenter, to reveal elastic response in difference culture condition. Functional assessments, together with TIE results, were performed to check osteoblasts maturation trend, while DH images and nanoindentations returned phenotypical modifications related to cellular evolution.

### ***Osteogenic staining***

Calcium deposition was assessed by staining with Alizarin Red S (Petcchia et al. 2015). Briefly, hFOB grown on glass cover slips were fixed in 4% paraformaldehyde (PFA), and stained with 2% aqueous solution of Alizarin red S (Sigma Aldrich) for 10 minutes at room temperature.

In order to evaluate osteocalcin expression, cells were grown on cell culture dish (format: Ø 35 mm) with glass bottom for microscopic applications (Greiner Bio-One International, supplier: Savatec SRL, Torino, Italy), fixed in 4% PFA at room temperature, permeabilised with 0.2% Triton X-100 and blocked with 20% normal goat serum (Vector, Labs Burlingame, CA, USA) for 1 h at room temperature. Then cells were incubated with mouse anti-human osteocalcin (1:50 Santa Cruz Biotechnology), washed and incubated with Alexa-fluor 633 (red) goat anti-rabbit (Molecular Probes) (Petcchia et al. 2015). Image cells were acquired by confocal laser scanning microscopy (TCS SL microscope; Leica; Mannheim, Germany), using a 20X PL APO N.A. 0.7 objective.

### ***RNA extraction and reverse transcription***

Total RNA was extracted from cell populations collected at relevant experimental stages using the NucleospinRNA XS Kit (Macherey Nagel) according to manufacturer instructions. 1 µg of purified RNA was retro-transcribed using the iScript Advanced cDNA Synthesis Kit (Bio-Rad), which employs both oligo-dT and random primers to trigger the transcription reaction. The cDNA obtained was ready for real time-qPCR.

### ***Real time-qPCR experiments***

cDNAs retro-transcribed from hFOB RNAs were amplified performing real time-qPCR and assessed by SYBR Green fluorescence using SsoAdvanced SYBR Green Supermix (Bio-Rad Laboratories, Inc.) in a CFX Connect instrument (Bio-Rad Laboratories, Inc.). Thermal protocol and data treatment were performed as previously described (Petcchia et al. 2017). The list of genes and the sequence of forward and reverse relative primers are reported in Table 1. To obtain robust statistics, acquisitions were performed in triplicate.

<b>Target name (function)</b>	<b>Primer forward</b>	<b>Primer reverse</b>
<b>Actin (housekeeping)</b>	<i>CTGGAACGGTGAAGGTGACA</i>	<i>AAGGGACTTCTGTAAACAAT</i>
<b>GADPH (housekeeping)</b>	<i>GAAGGTGAAGGTCGGAGTC</i>	<i>GAAGATGGTGATGGGATTC</i>
<b>KDM6A (epigenetic)</b>	<i>GAGGGAAGCTCTCATGCTG</i>	<i>AGATGAGGCGGATGTAATG</i>
<b>IBSP (osteospecific)</b>	<i>TCAGCCTCAGAGTCTCATCTTC</i>	<i>GGGAGTAGTGACTCATCCG</i>
<b>BGLAP/OC (osteospecific)</b>	<i>CTCACACTCCTCGCCCTATT</i>	<i>GTAGTGAAGAGACCCAGGCG</i>
<b>SPP1/OP (osteospecific)</b>	<i>CATCACCTGTGCCATACCAG</i>	<i>AGATGGGTCAGGGTTTAGCC</i>

Table 1. List of target genes used in RT-qPCR and sequence of the relative forward and reverse primers.

### ***Quantitative phase imaging***

Quantitative phase imaging (QPI) techniques were exploited to obtain quantitative reconstruction of the cell morphology and to provide high contrast images of label-free samples. In this work, two different QPI approaches were adopted: DH, which is based on interferometry, and a noncoherent method which relies on the TIE (Mir et al. 2012). DH is based on the use of a laser, a coherent source, to illuminate the sample; the phase map is recovered from the interferogram obtained by the interference of the light passing through the sample and a reference unmodified wavefront (Gabor 1948). In the present study, image acquisition has been performed through HoloMonitor M3 (Phase Holographic Imaging PHI AB, Sweden). The system is equipped with 10X magnification for phase contrast (CFI Achro, Nikon) and 20X and 40X objectives (both PlanC N, Olympus) for working with digital holography.

The method based on the TIE is non-interferometric and relies on the fact that amplitude and phase distributions are mathematically coupled in defocused images. By measuring several intensity images near the focal plane (recording a stack of defocused intensity images) it is possible to invert the equation and calculate the phase (Teague 1983). This technique benefits from operating with a standard bright-field microscope, as long as it can move along the Z axis either the sample or the objective, thus drastically reducing the cost of the optical setup. For this work, TIE resolution method has been enabled on a home-made acquisition system that integrates standard modular components (Optem FUSION, Qioptiq Photonics GmbH & Co KG), and is equipped with an Optem 10x high-resolution, long working distance (NA 0.45; WD 19 mm), infinity-corrected objective, mounted on a motorized Z-axis with a 0.01  $\mu\text{m}$  resolution step. The sample is scanned through a motorized X-Y stage with a 0.5  $\mu\text{m}$  resolution step. A modular, compact and freely programmable stepper motor controller (phyMOTION™, Phytron, Gröbenzell, Germany) drives three motorized axes. A LED lamp is integrated for the illumination and a Gig-E DMK 23G274 camera (The Imaging Source, Bremen, Germany) equipped with a CCD (b/w, 1600 x 1200 pixel) is used. The control software was developed in LabVIEW (National Instruments, Austin, Texas) and it automates all microscope functions such as movement of the three axes, acquisition of the 2D bright-field image and acquisition of a Z-stack of images. The number of images acquired in the Z-stack and the Z-step size can be both defined by the user. A LabVIEW module was developed for executing the algorithm producing the resulting quantitative phase images.

### ***Image analysis***

DH images were processed in order to extract shape features enabling the definition of cell populations morphological aspects. Acquired images were analysed by means of functionalities made available in M3 HStudio software (Phase Holographic Imaging PHI AB Sweden). For each image, single cells were isolated: Minimum Error Method algorithm (Kittler and Illingworth 1986),

where appropriate thresholds are selected by a minimum error criterion, demonstrated to be the most effective for segmentation, coupled to a post-processing manual error fixing. For each segmented cell the software calculates 37 different morphological parameters, including 2D morphometric indicators (such as area, perimeter, eccentricity, etc.) and parameters associated to the third dimension (such as volume, thickness, roughness, etc.). The full list of parameters extracted from cells images by HStudio software is reported in the Supplementary Materials section. In particular, according to HoloMonitor M3 image analysis software, roughness is defined as cell height irregularity and it is computed as  $\acute{R} = \frac{1}{n} \cdot \sum_{i=1}^n |w_i - \varphi_i|$ , where  $w$  is the waviness image, i.e. the smoothed version of the image,  $\varphi$  is the original image, all averaged over  $n$  pixels. Eccentricity measures cell elongation and it is defined as  $E = \sqrt{1 - \frac{b_b^2}{a_b^2}}$ , where  $a_b$  and  $b_b$  are cell diameters, respectively along major and minor axis. Irregularity measures how much cell shape deviates from a circle and it is computed as  $I = 1 - \frac{4 \cdot \pi \cdot A}{p^2}$ , where  $A$  is cell area and  $p$  cell perimeter. Raw data were exported, and analysed using custom Python software.

TIE images in phase-reconstructed formats were processed through a digital image correlation (DIC-like) approach in Gwyddion open-source software (Nečas et al. 2012) in order to obtain easily interpretable 3D-like images.

### ***Elastic modulus assessment***

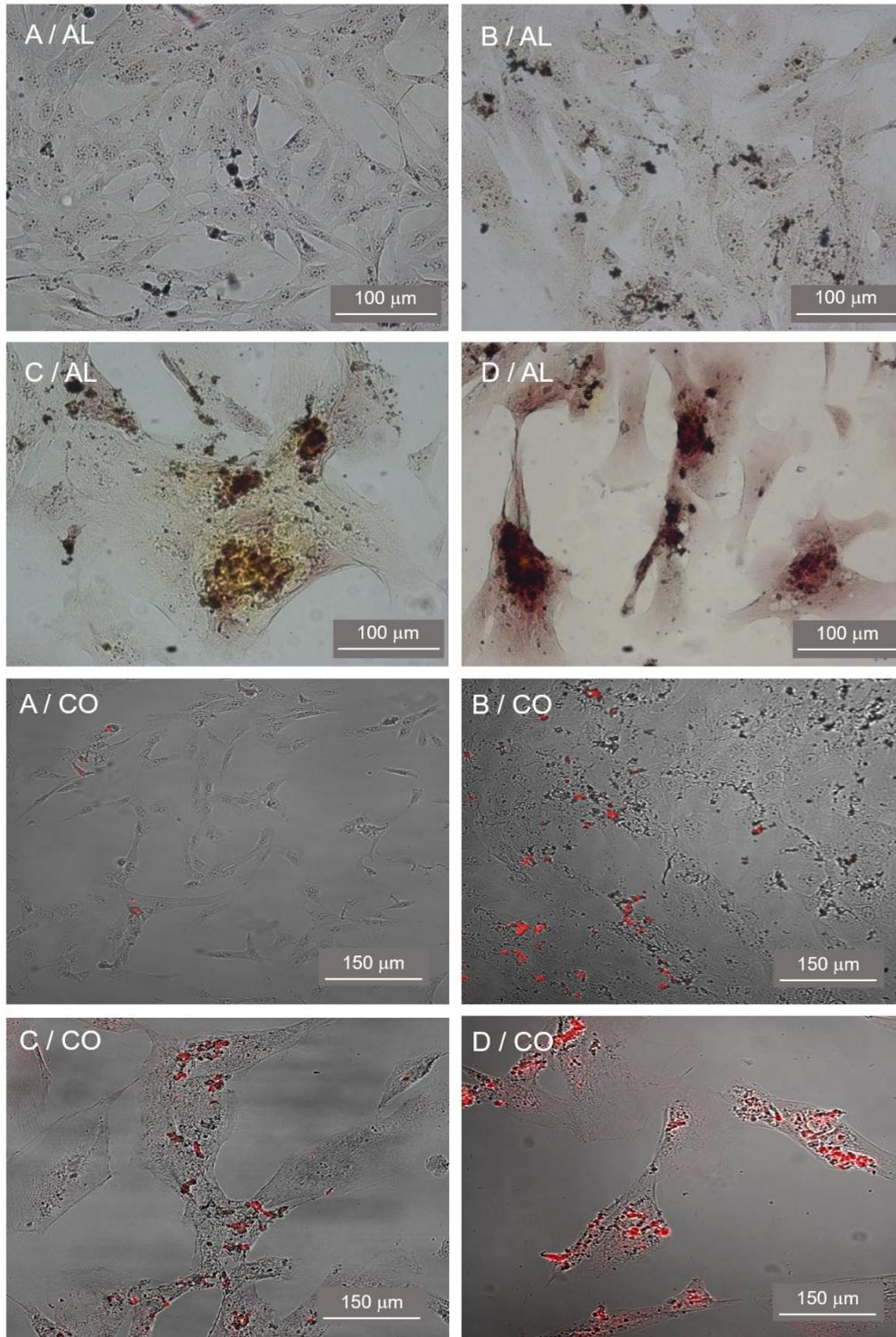
Mechanical properties indicate how a material responds to mechanical stimuli, in particular describing how it deforms in response to an applied stress, and how this deformation evolves over time. The scaling between stress and strain of a solid material can be expressed in terms of the so-called Young's modulus ( $E$ ), which indicates the ability of a material to sustain its shape under mechanical stress. A single-cell nanoindentation experiment aims at obtaining information on single cell elasticity, and is typically enabled by a calibrated force-sensing probe (commonly a  $\mu\text{m}$ -sized sphere) which is gently brought in contact with the cell at a constant speed. The resulting interaction force is measured as a function of the deformation of the cell, since the probe is glued to an elastic cantilever which bends proportionally to the force needed to deform the cell. Single cell elasticity value was obtained by using Piuma Chiaro (Optics11, NL), a commercial interferometric nanoindenter based on ferrule top technology (Chavan et al. 2012), adaptable to be anchored to various inverted microscopes in order to focus the cell, for this purpose attached to an Olympus IX71 inverted microscope. To measure the deflection, the Chiaro device integrates a laser interferometer, coupled to the instrument through an optical fiber. The tool can be programmed to automatically perform a grid of measurements over the cell. The probe used in this work was a glass sphere with a radius of  $10.5 \mu\text{m}$  and a stiffness of  $0.116 \text{ N/m}$ , suitable for measuring objects in the range of  $0.1-$

100 kPa. The same probe was exploited for all tested cells, in order to allow curves comparison at different time points and experimental conditions. Cells were measured in standard Petri dishes immersed in the growing medium suitable for hFOB (described in ‘Materials and methods’ section), and the exploited protocol foresees to enter in contact with cells in their center, which commonly is the wider and thicker area of the cell and allows a deeper indentation. Before starting each experiment session, the optical sensitivity and the geometrical factor were calibrated using the software provided with the instrument. After each experiment the probe was washed in 70% ethanol for approximately 10 minutes to remove eventually attached debris. Indentation curves were acquired at a speed of 2  $\mu\text{m/s}$ . Curves processing, aimed at extracting a parameter related to samples mechanical properties, required the adoption of a model describing the response of the cell to the compression of the indenter. In order to obtain a robust relative evaluation of cells elasticity, the direct fit method, DFIT (Vinckier and Semenza 1998) was applied, which considers the experienced force  $F$  obtained through a perfectly rigid and spherical indenter of radius  $R$ , modeled as  $F = \frac{4\sqrt{R}}{3} \frac{E}{1-\nu} \delta^{\frac{3}{2}}$  (Johnson et al., 1971) where  $\nu$  is the Poisson ratio of the sample, set to 0.5 as considered suitable for cells, which are approximated to incompressible material,  $E$  its Young’s modulus, and  $\delta$  the indentation depth, and fits it to the collected force-versus-indentation depth curves in order to get an estimate of the Young’s modulus for each cell. For excluding contribution of cell adhesion, the analysis was performed only on approach curves. All collected curves were analysed through Python custom scripts, relying on Numpy/Scipy Scientific Computing Stack (Millman and Aivazis 2011). Each dataset of curves acquired for the same experimental condition was pre-filtered to eliminate irrelevant curves (i.e. glass substrate or no substrate found) using a semi-automatic procedure. In the aim of removing the high-frequency noise and enable the determination of the tip-cell contact point, each force-indentation depth curve is filtered using Savitzky-Golay filter, and its derivative is computed. Since the tip-sample contact point is determined, the force indentation curve can be derived from the force curve. Then, force-indentation curves are fitted with DFIT on Hertz model, up to a threshold selected to match the suggested limit of the 10% of the cell thickness, and the Young’s modulus can be therefore calculated.

## Results

hFOB cells were induced to differentiate following the proposed protocol (see Materials and methods section) and the maturation stage towards osteoblasts was deeply characterized with reference biochemical approaches. Alizarin Red S staining was performed to visualize the amount of free calcium produced along the osteoblast mineralization process. hFOB cells in proliferation at 33.5°C were negative to the staining (Fig.1, A/AL), while a progressive increase of calcium-based

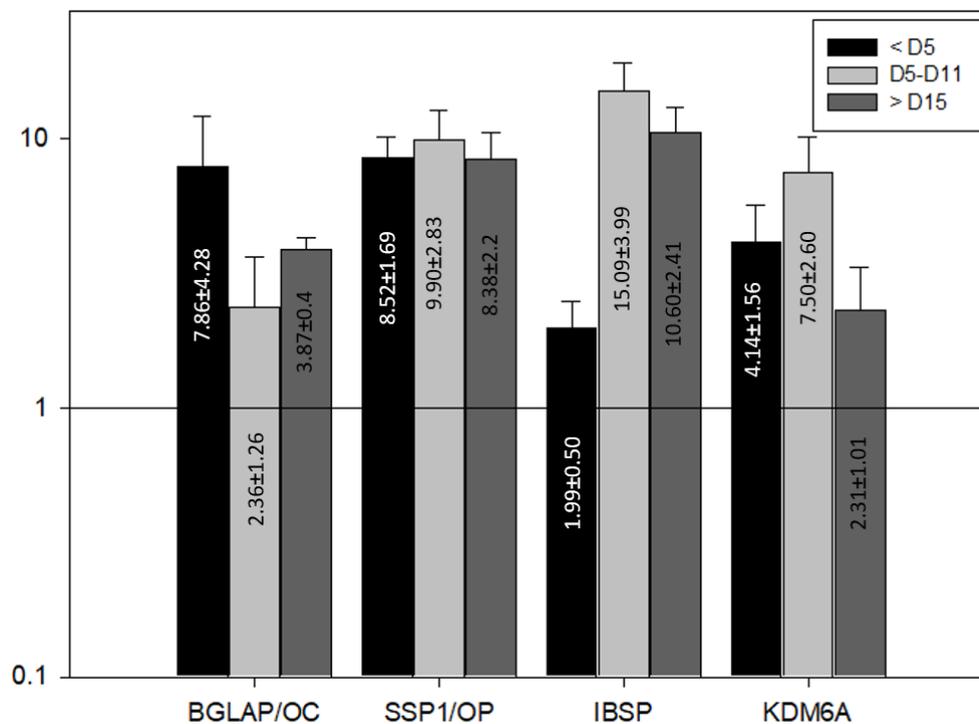
extracellular matrix was detected as sparse red spots after a few days of cell culturing at 39.5°C (Fig.1, B/AL) becoming even more evident in the following period (Fig.1, C/AL and D/AL), thus confirming the occurrence of cell maturation.



*Fig.1: Summary panel reporting imaging tests on hFOB 1.19 maturation. A to D: time steps for acquisition. AL: alizarin red staining; CO: confocal microscopy acquisition of osteocalcin fluorescent staining merged with absorbance images.*

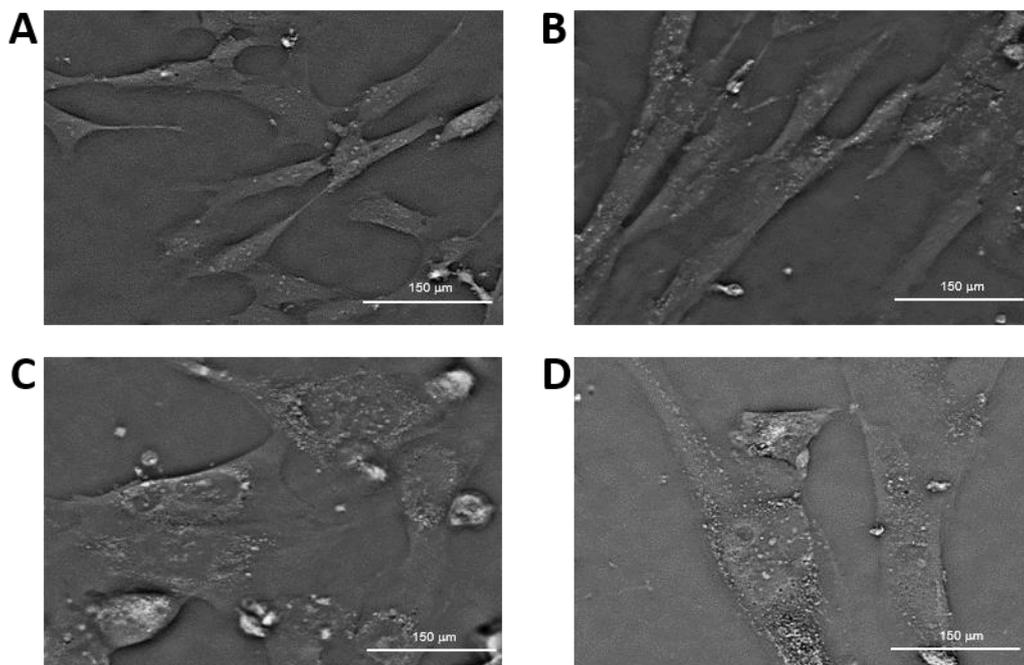
The progression of osteoblast maturation is usually accompanied by changes in osteocalcin production, therefore hFOB cells were labelled with antibodies against this protein. While osteocalcin staining is barely perceptible when cells actively proliferate at 33.5°C (Fig.1, A/CO), a red fluorescence starts to be detectable after 5 days at 39.5°C (Fig.1, B/CO) and has further increment after 11 and 15 days (Fig.1, C/CO and D/CO). The visual inspection of merged fluorescence and absorbance micrograph suggests that osteocalcin is mainly localized in the perinuclear region after 5 and 11 days, while has a wider cytoplasmic distribution after 15 days.

As further confirmation of the ongoing maturation process, the expression of several bone-related genes was assessed by real time qPCR experiments. Results are summarised in in Fig.2, where target expression levels, normalized both to the housekeeping genes and to the corresponding value in proliferating cells, are indicated on the histogram bars. Tested markers included Osteocalcin (BGLAP/OC), Osteopontin (SSP1/OP), integrin binding sialoprotein (IBSP) and the histone lysine demethylase 6A (KDM6A), a chromatin epigenetic regulator. According to the generally accepted model, the first three markers are up-regulated in the second stage of osteoblast maturation, when genes connected to proliferation are down-regulated and those correlated with extra-cellular matrix (ECM) maturation are activated (Stein et al. 2004). Indeed, independent of the considered time point, all cells grown at 39.5°C show a higher expression of these markers than cells grown at 33.5°C, confirming that a maturation process is occurring. The up-regulation of KDM6A, a gene connected with stromal MSCs osteogenesis (Hemmings et al. 2014), indicates an involvement of this protein also in hFOB maturation.



*Fig.2: Histogram reporting hFOB maturation through expression of main bone-related genes assessed by qPCR. Evaluation times: less than 5 days of growth at 39.5 °C (<D5), between 5 and 11 days of growth at 39.5 °C (D5-D11), more than 15 days of growth at 39.5 °C (>D15). Expression values of target genes are shown in histogram bars.*

For each growth condition, morphological and mechanical features were evaluated. These include qualitative shape knowledge from a Transport of Intensity Equation based approach (Fig.3), morphological quantitative data obtained by DH (Fig.4) and mechanical features from nanoindentation (Fig.5). Cell shape changes induced by maturation were qualitatively monitored using the 3D-like reconstruction based on QPI. This approach does not require any specialized cell treatment (no staining nor fixation) to obtain the 3D image of the cells, allowing characterization of the sample along time. Based on the acquired stack of images, a digital phase map is calculated and rendered in Fig.3 using a differential (DIC-like) filter which enhances finer details of cell morphology (McMahon et al 2002). Images reconstructed at different time steps show substantially different morphology of the cell body and appearance of the extra-cellular environment. hFOb grown at 33.5°C are smaller and narrower than mature osteoblasts grown at 39.5°C, which are wider, polygonal-shaped and whose environment shows the presence of grains in the ECM, putatively related to calcium-based release by cells.



*Fig.3: Images of cells morphology reconstructed using the Transport of Intensity equation during hFOB maturation (A=33,5°C, B=39,5°C for 5 days, C=39,5°C for 11 days, D=39,5°C for 15 days).*

The TIE-based imaging demonstrated to be effective for qualitative characterization of cellular samples, but DH is outperforming while addressing quantitative evaluation of the different cellular features. DH was deployed to image cells at different time points along maturation, and a further

step of image processing allowed to segment and measure single cells. A typical holographic cell image obtained from this kind of experiment is reported in Fig.4A, together with a schematization of checked morphological parameters detailed in the Materials and methods section (Fig.4B).

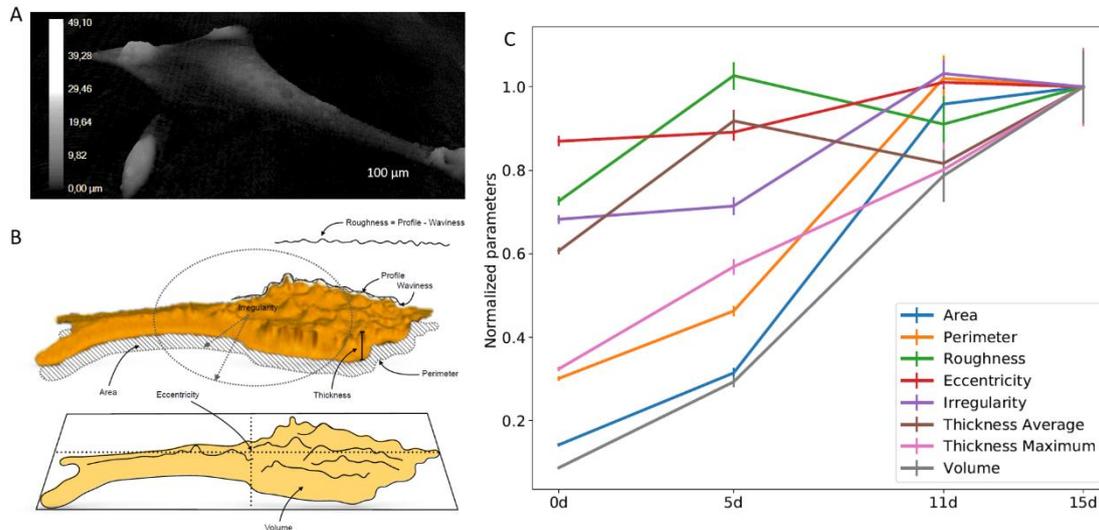


Fig.4: Digital holography. A) Example of a holographic hFOB image obtained from DH through a 20X magnification objective. B) Schematization of checked morphological parameters available through DH. C) Trends of morphological parameters changes in considered time steps (0d: proliferation condition, A; 5d: 39.5°C for 5 days, B; 11d: 39.5°C for 11 days, C; 15d: 39.5°C for 15 days, D).

Cells images were acquired in different conditions, and segmented to extract shapes' values for each cell. 296 cells from 14 images were processed for condition A; 122 cells from 21 images were processed for condition B; 20 cells from 10 images were processed for condition C; 27 cells from 14 images were processed for condition D. Average parameters, together with their standard deviation, were calculated for each condition and plotted in Fig.4C.

Consistently with knowledge from TIE, osteoblasts size increases with maturation; in each image only few, large mature cells were present when acquisition was done on samples grown for more days at 39.5°C, while a higher number of small cells was detected in each image acquired from proliferating cells.

Parameter	Conditions			
	A (33.5°C)	B (39.5°C/5days)	C (39.5°C/11days)	D (39.5°C/15days)
Area (μm <sup>2</sup> )	1365.60 ± 37.61	3039.30 ± 139.97	9038.27 ± 562.48	9702.94 ± 877.62
Eccentricity	0.80 ± 0.01	0.80 ± 0.18	0.90 ± 0.01	0.90 ± 0.01
Irregularity	0.40 ± 0.01	0.40 ± 0.01	0.65 ± 0.20	0.62 ± 0.02
ThicknessMax (μm)	3.77 ± 0.08	6.64 ± 0.28	9.53 ± 1.02	12.03 ± 1.38
ThicknessAvg (μm <sup>2</sup> )	1.21 ± 0.02	1.89 ± 0.07	1.62 ± 0.08	2.07 ± 0.19
Perimeter (μm)	174.55 ± 3.37	266.97 ± 8.78	591.77 ± 35.2	578.25 ± 34.01
RoughnessAvg (μm)	4.17 ± 0.08	5.98 ± 0.25	5.12 ± 0.41	5.71 ± 0.49
Volume (μm <sup>3</sup> )	1671.60 ± 56.89	5510.84 ± 273.50	14716.45 ± 1293.01	19351.20 ± 2149.43

Table 2: Average values of morphological parameters exploited in the analysis, and related SEM.

Cell elastic modulus was measured for each time point following the procedure previously described. For each experimental condition, around 150 indentation curves were acquired, repeating acquisitions at least 3 times for each cultured condition. Measurements were performed at room temperature and the indentations were collected for less than 1 hour for each cell plate, in order to preserve the stability of the sample. A typical force-displacement curve obtained from this experiment is reported in Fig.5A.

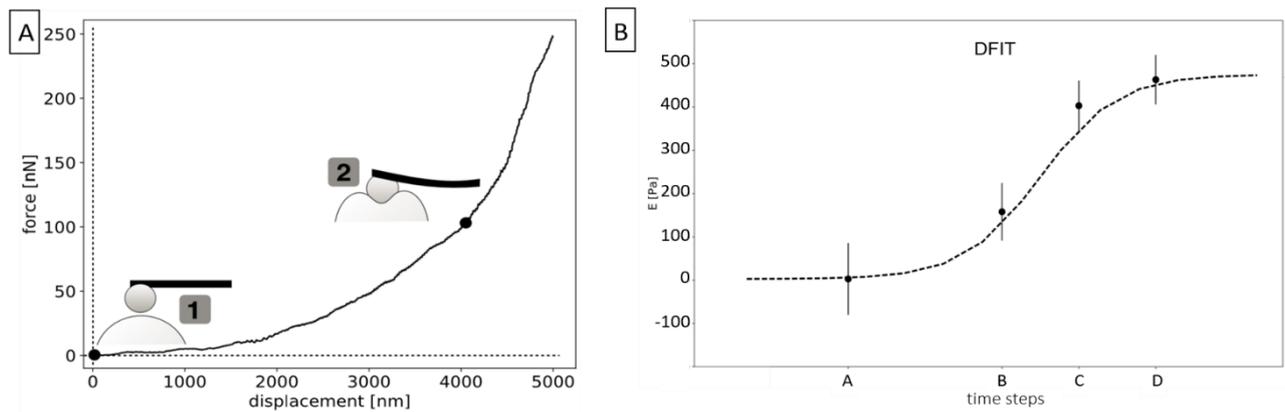


Fig.5: Nanoindentation. A) Force-displacement curve of the indentation phase. In step '1' Chiaro probe is approaching the cell; step '2' represents cell indentation and associated cantilever bending. B) Average elastic modulus measured during hFOB maturation. Results are obtained from DFIT approach applied to nanoindentation curves; the dashed curve represents a best fit with a sigmoidal trend and it is included as a guide to the eyes.

Fig.5B shows the average elastic modulus (E) measured during hFOB maturation and calculated using the DFIT approach. Relative values are shown, considering as reference value obtained for condition A ( $E=2400\text{Pa}$ ). The obtained results clearly show a tendency of cells to get more rigid with maturation, with a marked increase during the initial steps (points A, B, C), followed by a saturation phase (point D), where values are quite constant.

Trends of morphometric and elastic features during time steps have been correlated pairwise, using Pearson correlation index calculated using Python Pandas library (<https://pandas.pydata.org>): results are shown in a correlation matrix reported in Fig.6.

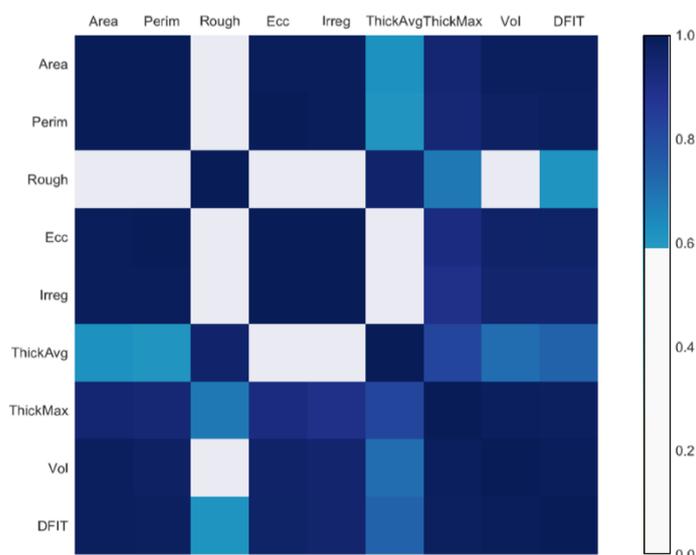


Fig.6: Correlation matrix of considered parameters over time: it measures the correlation of elasticity trend during hFOB maturation with trends of the analysed morphometric features. Only values showing  $r > 0.6$  are shown.

## Discussion and conclusion

Bone is a dynamic tissue, constantly being reshaped. Osteoblasts, which are the major cellular component of this tissue, are responsible for producing and secreting matrix proteins and transporting mineral into the matrix, which will be broken down by osteoclasts in the tissue remodelling process. Osteoblasts derive from mesenchymal stem cells (MSCs), which are also able to give rise to adipose or cartilage cells. When osteoblasts are buried and immobilized into extracellular matrix they become osteocytes. During their journey from MSCs to osteocytes, osteoblasts undertake diverse functional conditions, and monitoring their maturation is crucial for assessing bone tissue formation, which represent the basis for bone-related diseases studies and bone tissue regeneration approaches. A number of techniques are commonly exploited in order to evaluate osteoblasts stage of maturation. Most of them make use of chemical or biochemical labels, such as alizarin red for detecting  $\text{Ca}^{2+}$  production (Putchler et al. 1969) or fluorescent dyes for revealing osteocalcin (BGLAP) presence (Tsao et al. 2017), to detect the presence of specific molecules in the extracellular matrix (ECM) surrounding osteoblasts. Other approaches involve the use of qPCR to evaluate transcription levels of targets known to be differently expressed in diverse stages of osteoblast maturation, such as IBSP, OP and OC (Setzer et al. 2009). However, label-free techniques, which could represent faster and user-friendly approaches, did not enter yet the common lab practices.

In this paper a label-free morpho-mechanics approach for assessing osteoblasts maturation relying on morphological and mechanical modifications is presented. It is recognized that both morphological and mechanical properties of living cells are associated with their structural

characteristics (Yang et al. 2016; Atilgan et al. 2005) and their functional state, and their study contributes to understand and evaluate, among the others, drug treatment effects (Pasqualato et al. 2012), immune cell activation (Lin et al. 2015), cell differentiation (Petcchia et al. 2017) and cancer prognosis (Efremov et al. 2014). For example, alterations in cell cytoskeletal architecture have direct effects not only on cell morphology (Fletcher et al 2010) but even on organelles organization (Bershadsky and Vasiliev 1988), cytoplasmic trafficking (Willmann and Dringen 2018), cell adhesion (Parsons et al. 2010), migration (Tang and Gerlach 2017), polarization (Raman et al. 2018), and differentiation (Müller et al. 2013), thus modulating status and health. In this context, cell shape phenotype and mechanotype are emerging as valuable label-free biomarkers. In this work mechanical properties were assessed through single-cell nanoindentation, while morphological parameters were acquired and estimated through digital holography, an interferometric approach able to provide label-free 2D and 3D single-cell shape features. hFOB 1.19 cell line was chosen as model for bone cells, not only because it is a well characterized cell line (Harris et al. 1995, Setzer et al. 2009), but also for its unique and intrinsic feature of evolving from pre-osteoblasts to pre-osteocytes depending on temperature and time of culture. Its maturation state was assessed by traditional qualitative staining (Alizarin Red staining and osteocalcin fluorescent labelling) and quantitative gene expression (qPCR) systems. Using innovative technologies (DH, TIE, single cell nanoindentation), hFOB morpho-mechanical phenotypes were analysed at different time points of maturation, thus concurrently assessing their cellular functionality and morpho-mechanical characteristics.

Results show that hFOB morpho-mechanical features are highly correlated to their physiological stage (Fig.6), thus representing promising label-free tools able to check osteoblast maturation. It was expected that mechanical properties could play a relevant role in cell life and function: actually, basic processes such as cell growth, proliferation, migration, adhesion, and differentiation, all depend on and are regulated by mechanical properties (Titushkin et al. 2007). In recent studies the definition of living cell stiffness proved to be a biophysical fingerprint able to discern between cell phenotypes, to unravel processes in aging or diseases, to detect and diagnose pathological status of cell (Lim et al. 2011, Mescola et al. 2012, Lee et al. 2016, Northcott et al. 2018, Scholz et al. 2018, Harris et al. 2018, Zemła et al. 2018). Since elastic modulus enormously depends on exploited technology and chosen probe (Wu et al. 2018) and absolute data between different experiments are often highly incomparable between different experiments, the most appropriate way to show evolution of this parameter during hFOB maturation appeared to be the increment with respect to the initial condition value. hFOB stiffness, which appeared in line with literature findings (Docheva et al. 2008, Kelly et al. 2011), increases of about 450 Pa during cells maturation process, following a trend that can be fitted on a sigmoidal curve (Fig.5B).

For what concerns morphology features, scientific literature even shows that cell shape can be associated to functional and health status (Baba et al. 2007, Petecchia et al. 2017). Testing hFOB, both qualitative TIE and quantitative holographic images highlight important changes in cells morphology (Fig.3 and Table 2). Morphology changes can be associated to differentiation toward mature osteoblasts, thus expressing the normal osteoblast phenotype, characterized by large, flat and polygonally- shaped cells surrounded from calcium-based grains. It is worth noticing that some shape characteristics of hFOB evolve in temperature and time with trends analogous to the elastic behaviour, leading to hypothesize that consistent cytoskeletal and internal organelles remodelling impact not only on cell morphology and behaviour, but also on mechanical properties. In particular, the evolution of osteoblasts stiffness, acquired over time in proliferation and maturation modes, results to be highly correlated ( $r>0.9$ ) with modification in osteoblasts occupancy area, perimeter, eccentricity factor, surface irregularity, maximum thickness and volume (Fig.6).

In conclusion, the present work provides a detailed study of the changes of the morpho-mechanical phenotype of hFOB cells along their maturation from pre-osteoblasts to mature osteoblasts. Although the chosen cellular system is a genetically designed model, only partially recapitulating the complexity of the original bone tissue (Rutkovskiy et al. 2016), this work suggests that morphological and elastic properties can represent a relevant marker to follow osteoblasts during their maturation process. The presented study provides an indication of a connection between the physical properties of the cell and the ability to develop towards a mature bone tissue. Although the proposed measurements intend to highlight a statistical link between cell morpho-mechanics and its maturation stage, a biomolecular interpretation of findings can be hypothesised. In particular, recent results by Li et al. (2019) and Sun et al. (2019) shed additional light on the mechano-transduction processes activated in the bone (Shuaib et al. 2019, Alfieri et al. 2019), clearly demonstrating a central role of the mechanosensitive ion channel Piezo1 in bone formation and anabolism in animal systems. The knockout of Piezo1, a very peculiar molecule capable of directly sensing and transducing mechanical stimuli, thus determining the fate of differentiating mesenchymal stem cells (Sugimoto et al. 2017), severely impairs bone structure and strength. This mechano-transduction process influences and is potently influenced by the organization of the cytoskeleton (Cox et al. 2016) and the biomechanics of the cell and of the underlying environment (Bavi et al. 2019). The findings of the present paper can be interpreted in this scenario, directly connecting the cell mechano-type and its biological fate: the possibility to characterize morpho-mechanical aspects using advanced biophysical tools opens the way for a fine measurement of the maturation of osteoblasts in patho-physiologically relevant contexts.

## Acknowledgements

This work was supported by the Ministry of Education, Universities and Research of Italy, in the context of the FIRB 'Futuro in Ricerca 2013' project [grant number RBF13V4M2]. Thanks to Francesca Quartino for technical support.

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