



PROJECT DELIVERABLE REPORT



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Advanced personalised, multi-scale computer models preventing osteoarthritis SC1-PM-17-2017 - Personalised computer models and in-silico systems for well-being

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3 Summary

Osteoarthritis (OA) is one of most common types of musculoskeletal diseases worldwide and is categorized by the European Commission as one of the twenty-four priority diseases. The OActive project aims to improve the life's quality of people who suffer or would probably suffer from OA through personalized prevention and intervention. Over the last few years, new evidence has emerged suggesting that changes in the gut microbiome may also be linked to developmental factors in the onset of musculoskeletal disorders such as OA, but the possible role of the mechanisms involved remains unclear.

Computer modelling is going to be used for integrating various datasets such as lifestyle, organ and tissue level mechanistic, environmental and biochemical biomarkers among other. The miRNA exosome content and microbiome data generated during this task will be included in the computer modelling algorithm in order to help to improve the personalized prediction of OA onset.

Contributions from partners

For the fulfilment of this task, technical inputs have been received from the following partners:

- LEITAT will use the knowledge attained during the project in order to strengthen their research activities in preclinical assays, such as exosome studies and biomarkers investigation on miRNA and microbiome. Moreover, the project will enable to enlarge their expertise in microbiome techniques, expanding thus their portfolio in Technical Services. This will enable them to provide greater technological value long-term to their customers, maintain their principles and values and expand their business network.
- UNIC Plasma samples from OA patients
- HULAFE Plasma and stool samples from healthy volunteers and early OA patients.
- ANIMUS Plasma samples from high-risk OA athletes.

This report refers to Deliverable 4.2, which relates to the OACTIVE WP 4, "Biochemical modelling and inflammation biomarkers" led by LEITAT and specifically Task 4.3 "Qualification of OA-related exosomal and microbiome biomarkers", also led by LEITAT.

4 Introduction

Osteoarthritis (OA) is one of the world's most common types of musculoskeletal diseases and its prevalence is increased, especially since the middle of the 20th century (Biver et al., 2019). In Europe is the most common form of chronic pain (34%), representing a great economic and social cost to society. At present there are no recognized drugs to modify the disease state, the only alternative being palliation of symptoms. Different risk factors may play an important role in OA, including obesity, joint trauma, age, hormonal disorders, and genetic background (Abramoff and Caldera, 2020). There is increasing evidence for an inflammatory component to OA linked with joint tissue damage (Robinson *et al.*, 2016)Scanzello, 2017).

Knee osteoarthritis (OA), the most common arthritis, is a leading cause of disability in the elderly population mainly due to pain, which is the primary symptom of the disease (Neogi, 2013). It has been recently estimated that more than 250 million people suffer from knee OA (Ro, D.H., et al., 2019). Knee OA is characterized by structural modifications to primarily articular cartilage and the subchondral bone, but also Hoffa's fat pad, synovia, ligaments, and muscles, suggesting that knee OA should be observed as a whole joint disease (Loeser et al. 2012). Moreover, OA is a complex disease involving all the tissues of the joint as well as inflammation and thus it is a collection of heterogeneous pathologies that result in a

common outcome (pain and joint destruction), rather than being one homogeneous disease (Laslett., et al., 2016). Furthermore, the Osteoarthritis Research Society International (OARSI) has recently endorsed a new definition of OA: "Osteoarthritis is a disorder involving movable joints characterized by cell stress and extracellular matrix degradation initiated by micro-and macro-injury that activates maladaptive repair responses including pro-inflammatory pathways of innate immunity. The disease manifests first as a molecular derangement (abnormal joint tissue metabolism) followed by anatomic, and/or physiologic derangements (characterized by cartilage degradation, bone remodeling, osteophyte formation, joint inflammation, and loss of normal joint function), that can culminate in illness" (Kraus, V.B., et al., 2015). From this definition, and as aforementioned, can be easily deduced that OA is a heterogeneous disease with a variety of pathophysiologic drivers leading to multiple phenotypes.

Currently, there are no therapies and/or medication approved by regulatory authorities that alter the onset or progression of OA structural damage while the available treatments have only moderate effects, and therefore, patients are not satisfied with their efficacy (Zhang, et al., 2010). Due to the failure of the current medications to treat knee OA, the number of joint replacement surgeries is increasing by $\sim 10\%$ annually (Hunter, D.J., et al., 2014). The anatomical severity of OA is usually assessed by joint imaging using standard radiographs, (x-Ray, and magnetic resonance imaging), however, by the time OA is detected on the radiograph, significant cartilage degradation has already occurred (Ishijima, M., et al., 2011). Therefore, there is an urgent need for the identification of alternative methods for early detection and monitoring OA while the identification of OA before it becomes evident on radiographs remains a challenge (Cibere et al., 2009).

In the past few years, the prognostic and diagnostic utility of biomarkers in OA has been investigated. Specific molecules of the cartilage and bone turnover that are released in the blood or other biological samples could be used as biomarkers for identifying patients at greater risk for developing OA earlier than the time required to reveal radiographic changes and also to monitor the progression of the disease (Cahue, S., et al., 2007). Biomarkers are candidates that are now being used to detect and monitor cartilage, bone turnover, and synovial metabolism for the critical assessment of the pathophysiological processes that lead to joint failure and pain in OA patients (Ishijima, et al., 2011). Several biomarkers have been tested in samples from patients with varying severity of OA during the last year. Interestingly, specific biomarkers have also been found to predict OA progression in longitudinal studies (Cahue, S., et al., 2007). This work aimed to validate characteristic a new source of biomarkers, as exosomes content and gut microbiome in samples from OA patients. The main objectives of this work were to:

- Perform an exploratory analysis to determine the potential use of microRNA content of exosomes data as an OA prognosis biomarker.
- Perform an exploratory analysis to determine the potential use of microbiome data as OA prognostic biomarkers for OA.
- Generate the data needed in the correct format to be integrated into the computer models designed for personalized prevention of OA.

5 Exosome content - miRNA studies

The evaluation of new markers for clinical practice requires standardized pre-analytical, analytical, and postanalytical protocols aimed at providing accurate, reproducible, and consistent, preferably non-invasive, measures by controlling for all variables that may introduce biases in the detection of biomarkers (type of sample, manipulation, method of measurement). In the case of the analysis of circulating miRNAs as biomarkers in clinical practice, and research they show several advantageous features as they can be evaluated in human biofluids (e.g., urine, plasma, and serum). However, it is essential to standardize sampling and manipulation protocols to reduce the bias that may affect the pre-analytical phase of miRNA

validation. The selection of reference genes along with the standardization strategy applied for miRNA quantification represents the most difficult problem due to the absence of a standardized methodology. RT-qPCR data for miRNA expression can be normalized using single or multiple endogenous and/or exogenous reference genes, as well as by the mean expression value of all measured miRNAs. Taking all these aspects into account, it is necessary to carry out a careful validation of the results and a comparative work with the public databases in the studies of this field, in our case, of OA. For this reason, a data collection procedure has been developed that includes quality controls from each step, as expressed in the workflow diagram in Figure 1, to obtain robust results that can be applied to the early diagnosis and prognosis of OA in clinical practice and can serve also to contributing to the computer modelling of the OActive project.



Figure 1: Diagrammatic representation of the workflow. The thick arrow shows where is now the research.

5.1 Relation of plasma and urine samples received

HULAFE: 84 plasma and urine samples were received during 2019 and 2020.

UNIC: 51 plasma samples. July 2019.

ANIMUS: 25 plasma samples. November 2020.

Most of the plasma samples were 5 ml or less, and urine samples from 4 to 10 ml. HULAFE samples were split to leave 2ml of plasma and urine for ELISA assays. This small volume forces us to limit the study to miRNA analysis.

5.2 Exosome extraction

Exosomes from plasma samples were extracted with the Plasma/Serum Exosome Purification kit from Norgen Biotech. Plasma volume for all samples was from 2 to 4 ml. Quality control -size particle and concentration- was done with 12 samples in a Nanosight N300 nanoparticle analyser. As it is shown in figure 2, exosome suspension was homogenous and had the expected reported size, around 100 nm. Then, it was assumed that all samples extracted will have similar quality.

Exosomes from urine samples were extracted using an adapted protocol in Leitat that was the one that gave higher yields. Briefly, 5 ml of urine sample was centrifuged at 2000rcf, 30min at 4°C. The supernatant (sn) was recovered and centrifuged at 16000 rcf for 1h at 4°C. Then, the sn was filtered through a 0.22 μ m filter and the Total Exosome Isolation Reagent (for urine) (Invitrogen) solution was added at 1:1 and incubated for 1h at RT. The mixture was centrifuged at 10000 rcf for 1h at 4°C, and the pellet was resuspended in 500 μ l of PBS. Quality control was done in the same way as plasma samples, but with fewer samples because it was done before with volunteers' samples to set up the method. As it can be shown in Figure 2, there was more heterogenous composition and less concentrated samples. A)



Figure 2. Examples of quantification and size analysis of microvesicles extracted from A) plasma and B) urine samples. The analysis was done with Nanosight N300

5.3 miRNA purification and cDNA synthesis

RNA from all exosomes was extracted using the RNA extraction kit from Norgen. For quality control of the miRNA yield, spike miRNA (RNA Spike-In Kit Qiagen) was added to the lysis buffer. These spike miRNAs were used as a control of RNA purification in the miRCURY LNA SYBR® Green PCR amplification system using the QC PCR Panel, miRCURY LNA miRNA QC PCR Panel from Qiagen. It was not possible to quantify RNA concentration due to the low yield obtained in exome extraction. That is why it was necessary to include spike miRNA for quality control of the RNA extractions.

Two methods of cDNA synthesis will be used. The miRCURY LNA RT Kit method for quality control adding spike miRNA, and the Taqman Adv miRNA cDNA synthesis (Applied Biosystems) to use in the TaqManTM OpenArrayTM Human Advanced MicroRNA Panel, QuantStudioTM 12K Flex (Applied Biosystems).

5.4 Quality control analysis

5.4.1 cDNA synthesis

cDNA was synthetized with the miRCURY LNA RT Kit. As it was pointed out, it was not possible to measure RNA concentration, so for the quality control assay, it was necessary to synthetize cDNA using increasing volumes of the RNA samples. Table 1 shows the reaction settings.

	RNA reaction (µl)
Sample RNA	6
5X miRCURY SYBR Green RT reaction	2
RNAse free H2O	14
10x miRCURY RT enzyme mix	4
miR-39-3p	2
Template RNA	12
TOTAL reaction volume	40

Table 1 cDNA synthesis using different amounts of RNA to determine the right amount for the qPCR analysis.

5.4.2 qPCR Quality control

The miRCURY LNA miRNA QC PCR panel was used. Table 2 shows the reaction settings and Table 3 the composition of the qPCR panel. The qPCR panel includes all the Spike miRNA added in the purification and cDNA synthesis.

Table 2 – Reaction mix for qPCR control panel

COMPONENT	QC PCR PANEL, μl
2x miRCURY SYBR Green Master Mix	70
cDNA template	1.4
RNAse free water	68.6
TOTAL	140

The qPCR settings for a LightCycler® 480 Instrument II (Roche Life Science) were:

Activation 95°C 2 min

Amplification 45 cycles 95 °C, 10 sec

56 °C, 60 sec

Table 3 miRNA sequences included in the miRCURY LNA miRNA QC PCR 384 panel. Up to 32 samples can be run. These panels include the spike-in sequences included in the RNA extraction and cDNA synthesis and known miRNA sequences expressed in different tissues.

miRname (human)	microRNA target sequence	Corresponding LNA™ microRNA PCR primer set (Prod No)	Sample	Assay type
hsa-miR-103a-3p	AGCAGCAUUGUACAGGGCUAUGA	YP00204063	S1	GOI
hsa-miR-191-5p	CAACGGAAUCCCAAAAGCAGCUG	YP00204306	S1	GOI
hsa-miR-451a	AAACCGUUACCAUUACUGAGUU	YP02119305	S1	GOI
hsa-miR-23a-3p	AUCACAUUGCCAGGGAUUUCC	YP00204772	S1	GOI
UniSp6		YP00203954	S1	Spike
UniSp2		YP00203950	S1	Spike
UniSp4		YP00203953	S1	Spike
UniSp5		YP00203955	S1	Spike
cel-miR-39-3p		YP00203952	S1	Spike
UniSp3 IPC		YP02119288	S1	IPC

The quality analysis of samples was performed only in 96 samples (green box before the number) out of the 102 predicted by the analysis with the TaqManTM OpenArrayTM Human Advanced MicroRNA Panel. Table 4. The criteria for HULAFE and ANIMUS samples was to select by sex and body mass index (BMI) because all patients were early OA for HULAFE and risk volunteers for ANIMUS. The criteria for UNIC were sex and Kellgren-Lawrence (K&L) scores for the left (L) and right knee (R). Controls from HULAFE were also included.

 Table 4 - Relation of samples included in the miRNA analysis. Green boxes indicate the samples that were selected for the TaqManTM OpenArrayTM

 Human Advanced MicroRNA Panel

	HULAFE												
Se													
le													
ct					COM			А					
ed	Patient	Sex	Status	BMI	Р	HA	PIICP	ge	IL1β	TNFα	IL6	K&L L	K&L R
		Femal	Early										
	SPH089	e	OA	18.62	161.9	155.9	438.648	53	<1.95	21.7	<10,24	0=0 points	0=0 points
		Femal	Early										
	SPH069	e	OA	19.492	172.2	83.3	257.484	46	<1.95	30.8	<10,24	0=0 points	0=0 points
		Femal	Early										
	SPH082	e	OA	20.69	266.5	66.3	666.066	54	<1.95	22	<10,24	-	-
		Femal	Early										
	SPH067	e	OA	21.855	208.8	31.9	525.76	47	<1.95	34.5	<10,24	0=0 points	0=0 points
		Femal	Early		198.1	65.38					1.16460		
	SPH019	e	OA	23.01	77	2	589.123	58	2.81892	14.588	7	0=0 points	0=0 points
		Femal	Early										
	SPH072	e	OA	23.524	120.1	49.7	326.095	49	<1.95	52.7	<10,24	0=0 points	0=0 points
		Femal	Early		135.8	65.21			2.42328	5.73685	4.05492		
	SPH017	e	OA	23.533	8	2	834.375	44	2	5	7	0=0 points	0=0 points
		Femal	Early										
	SPH047	e	OA	23.533	382.7	150.8	625.208	55	<1.95	28.7	<10,24	0= 0 points	0= 0 points

	Femal	Early										
SPH071	e Esmal	OA	23.627	214	67.5	466.4	43	<1.95	33.9	<10,24	0=0 points	0=0 points
SPH022	e	OA	23.147	232.0 34	94.05 3	526.451	60	3.06619	21.8000	0.46584	2- 5-4 points	2- 3-4 points
	Femal	Early									r ·	r · ··
SPH044	e	OA	24.281	292.8	73.2	670.691	57	<1.95	26.6	<10,24	0= 0 points	0= 0 points
SDLIGES	Femal	Early	24 747	252.6	77	693 026	52	<1.05	27.1	<10.24	0 = 0 points	0 = 0 points
3PH036	e Femal	Early	24.747	232.0	11	085.020	55	<1.93	2/.1	<10,24	0– 0 points	0– 0 points
SPH091	e	OA	24.805	215.2	70	253.629	46	0.2	29.6	<10,24	0= 0 points	0= 0 points
	Femal	Early										
SPH053	e	OA	24.883	456.7	226.8	998.328	57	<1.95	31.8	<10,24	0=0 points	0=0 points
SPH001	Femal	DA	25 097	363.1 5	117.5 86	680 409	65	<1.95	<23.4	0.81522	0=0 points	0=0 points
0111001	Femal	Early	201077			0001107	00	1170	2011		o o pointo	o o pointo
SPH078	e	OA	25.796	178.9	80.2	653.731	51	<1.95	24.7	<10,24	0=0 points	0=0 points
		- ·					11					
SDH 05	Femal	Early	26 504	64	0.6	80	66	2	35.1	<10.24	1= 1-2	1= 1-2
5111_75	C		20.304	04	0.0	07	75	2	55.1	<10,24	points	points
	Femal	Early					3.				1= 1-2	
SPH_94	е	OA	26.915	-	0.4	34.1	6	2	34.2	<10,24	points	0=0 points
SDL1054	Femal	Early	27.716	104.1	201.0	(07.477	ΕC	<1.05	27.7	<10.24		
5PH054	e Femal	Early	27.710	194.1	201.9	607.477	50	<1.95	21.1	<10,24	-	-
SPH033	e	OA	27.84	276.9	58.9	434.793	56	<1.95	30.8	<10,24	0=0 points	0=0 points
	Femal	Early									-	-
SPH062	e	OA	28.172	180.6	132.5	313.76	54	<1.95	28.7	<10,24	-	-
SDH016	Femal	Early	28.65	270.8 57	38.63 2	833.03	60	1.73091	13.4406	0.19309	0 = 0 points	0 = 0 points
3111010	Femal	Early	20.05	57	2	055.05	00	0	5	2	0– 0 points	0– 0 points
SPH096	e	OA	29.261	-	-	-	53	-	-	-	0=0 points	0=0 points
	Femal	Early										
SPH064	e	OA	29.456	257.8	86	728.51	51	<1.95	29.8	<10,24	-	-
SPH038	Femal	DA	29 469	270.9	70.7	346 138	57	<1.95	29.2	<10.24	0=0 points	0=0 points
0111000	Femal	Early	271102	LITONS	/ 01/	0101200	01	1170	2712	10,21	o o pointo	o o ponito
SPH040	е	OA	31.398	355.6	81.8	856.481	52	<1.95	37.6	<10,24	0= 0 points	0= 0 points
0011000	Femal	Early	21.115	253.6	93.12	44.60 500	50	-14.05		1.90219		
SPH002	e Femal	OA Forly	31.445	630.2	43.60	1168.528	52	< 1.95	<23.4	3 00848	0=0 points	0=0 points
SPH004	e	OA	32.447	23		1345.354	55	<1.95	5.02272	4	0=0 points	0= 0 points
	Femal	Early		187.3	35.74			3.46183	34.7489	1.24224		
SPH023	e	OA	34.844	43	6	578.801	42	1	5	7	0= 0 points	0=0 points
SPH031	Femal	Early	35 330	90.3	94-5	430.6	46	2.0	12.5	0.2	0 = 0 points	0 = 0 points
5111051	Femal	Early	-55.557	- 70.5	- 74.5	450.0	10	2.9	12.5	0.2	1 = 1 - 2	1 = 1 - 2
SPH075	е	OA	35.725	202.8	147.9	421.688	46	<1.95	22.4	<10,24	points	points
	Femal	Early									1= 1-2	
SPH048	e Formal	OA	38.83	183.8	46.2	450.982	60	<1.95	28.7	<10,24	points	0=0 points
SPH030	e	Healthy	25.089	179.2	75.9	481.5	42	3.4	17.2	0.6	0=0 points	0=0 points
	Femal			92.09	48.57			3.21455	13.2767	0.77640		
SPH028	e	Healthy	25.113	2	8	321.474	47	8	2	5	-	-
SD11007	Femal	I.I. del	25 404	247.4	72.4	722.261	17	<1.05	24.7	<10.04	0-0.	0-0.
3PH087	Femal	Healthy	25.496	247.4	73.1	732.364	4/	<1.95	21.7	<10,24	0– 0 points	0– 0 points
SPH041	e	Healthy	25.97	305.9	163.9	633.688	53	<1.95	31.3	<10,24	-	-
	Femal			184.1	39.41			2.22546	10.8180	1.55280		
SPH027	e	Healthy	26.026	83	2	367.926	42	3	7	9	0=0 points	0=0 points
SDLIOPO	Femal	Health	26.272	225	42.0	776.206	50	<1.05	20.1	<10.24	0 = 0 exists	
311080	Femal	Healthy	20.272	555	43.2	770.306	39	~1.95	20.1	<10,24	0– 0 points	0– 0 points
SPH083	e	Healthy	26.396	270.2	115.9	471.026	53	<1.95	20.5	<10,24	0=0 points	0=0 points

		Femal											
	SPH079	e Esserel	Healthy	26.573	204.5	92.8	640.626	58	<1.95	19.4	6.9	0=0 points	0=0 points
	SPH018	e	Healthy	26.814	218.0 4	72.40 9	633.363	58	2.52219	0.90150	0.85404	0 = 0 points	0=0 points
		Femal			339.9	49.68				6.96163		· · · · ·	
	SPH008	e	Healthy	27.074	26	2	1002.46	57	<1.95	4	<10.24	0=0 points	0=0 points
	SPH045	Femal	Healthy	28.043	1547	237.8	272 902	48	<1.95	31.8	<10.24	0 = 0 points	0=0 points
	0111010	Femal	Treating	2010 15	10 117	20110	272.002		1170	5110	,21	o o pointo	o o pomito
	SPH065	e	Healthy	28.125	244.6	197.8	909.673	60	<1.95	30.3	<10,24	0=0 points	0=0 points
	SPH026	Femal	Healthy	28 214	190.5 03	101.2	968 109	54	3 65965	41.4692 7	0.85404	0 = 0 points	0 = 0 points
	0111020	Femal	Treating	20.211	187.3	54.11	,00.10	51	3.46183	16.3910	0.15528	o o points	o o pomito
	SPH025	е	Healthy	28.353	43	2	754.284	56	1	1	1	0=0 points	0= 0 points
	SPH035	Femal	Healthy	30.022	255	88.2	770 139	53	<1.95	32.4	<10.24	0 = 0 points	0 = 0 points
	0111035	Femal	Treating	50.022	233	00.2	110.137		-1.75	52.1	-10,21	o o points	o o points
	SPH097	е	Healthy	30.11	-	-	-	43	-	-	-	0=0 points	0= 0 points
	SDH052	Femal	Healthy	30.408	276.1	100.7	818 706	62	<1.95	34.5	<10.24	0 = 0 points	0 = 0 points
	3111032	Femal	Treating	30.400	270.1	177.7	010.700	02	<1.75	54.5	~10,24	0– 0 points	0– 0 points
	SPH093	е	Healthy	30.664	-	-	-	62	-	-	-	0= 0 points	0= 0 points
	SDH057	Femal	Healthy	31.033	170 /	67.2	578 182	60	<1.95	30.3	10.5		
	51 11057	Femal	Treating	51.055	179.4	07.2	576.102	00	<1.75	50.5	10.5		
	SPH060	е	Healthy	31.056	177.8	62.7	713.091	49	<1.95	30.8	<10,24	0=0 points	0=0 points
	SDLI036	Femal	Hoalthy	31 220	175.9	51.6	1200 754	52	<1.05	27.1	<10.24	0 = 0 points	
	5171050	Femal	Healthy	51.229	175.0	51.0	1299.734	52	<1.95	27.1	<10,24	0– 0 points	0– 0 points
	SPH086	е	Healthy	32.456	191.7	29	659.899	41	<1.95	25.8	<10,24	0=0 points	0=0 points
	SDI 1002	Femal	Lloghthur	24.965				47					
	5PH092	е	Healthy	34.003	-	-	-	4/	-	-	-	-	-
			Early										
	SPH039	Male	OA	22.14	249.4	133.7	467.171	58	<1.95	26.6	<10,24	0= 0 points	0= 0 points
			Early										
	SPH042	Male	OA	24.49	247 140.8	216.2	896.568	55	<1.95	25.1	<10,24	0 = 0 points $1 = 1_2$	0=0 points
	SPH007	Male	OA	25.136	46	4	607.796	46	<1.95	8	<10.24	points	0= 0 points
			Early										
	SPH063	Male	OA Early	25.295	234.7	68 72	834.895	45	<1.95	27.1	<10,24	-	-
	SPH009	Male	OA	25.69	92	9	821.601	54	<1.95	6	<10.24	0 = 0 points	0=0 points
					401.3	124.9			4.15419	15.2436	3.26089		3= 5-9
	SPH021	Male	Forly	25.617	2	25	523.501	51	8	4	9	0=0 points	points
	SPH051	Male	OA	26.908	193.7	85.6	532.699	57	<1.95	29.8	<10,24	0 = 0 points	0= 0 points
			Early										
	SPH077	Male	OA Early	26.919	207.4	77.9	816.393	48	<1.95	23.2	<10,24	0=0 points	0=0 points
	SPH032	Male	OA	26.99	92.1	100	492.5	48	2.9	15.4	0.5	0 = 0 points	0= 0 points
	ODU		Early	07.7	274.4					11.3706		0	
	SPH006	Male	OA Early	27.745	69	16.98	805.464	46	<1.95	7	<10.24	0=0 points	0=0 points
	SPH074	Male	OA	27.835	220.6	137.9	366.182	59	<1.95	28.1	<10,24	0 = 0 points	0=0 points
	-		Early										
<u> </u>	SPH073	Male	OA	28.076	449	120.4	569.702	48	<1.95	25.8	<10,24	0=0 points	0=0 points
			Early					2.					
	SPH_98	Male	OA	28.998	52	0.2	45.5	8	2	32.9	<10,24	-	-
	SPH076	Male	Early	31 161	292.5	27.6	947 448	52	<1.05	22	<10.24	0 = 0 points	0 = 0 points
	SDLIGZO	Mala	Hoghthr	25.04	215.2	60.2	707.605	52	<1.05	20.2	<10.24		
	3111008	wate	Healthy	25.04	213.2	08.5	707.695	55	<1.95	29.2	<10,24	0- 0 points	0- 0 points
	SPH088	Male	Healthy	25.112	208.6	72.8	506.488	53	<1.95	22	<10,24	0=0 points	0=0 points

SPH090	Male	Healthy	25.436	163.6	57.6	738.531	44	<1.95	25.8	<10,24	0= 0 points	0=0 points
				344.6					0.25317			
SPH003	Male	Healthy	25.912	91	9.898	474.673	46	<1.95	1	5.16309	0=0 points	0=0 points
SPH050	Male	Healthy	26.15	222	144.2	625.979	46	<1.95	25.6	<10,24	0=0 points	0=0 points
				194.1	48.98			2.37382	32.2555			
SPH013	Male	Healthy	26.504	14	6	1048.179	50	7	7	2.12401	0=0 points	0=0 points
SPH066	Male	Healthy	27.668	198.9	146.6	986.764	47	<1.95	26.6	<10,24	-	-
SPH070	Male	Healthy	27.932	158.7	69.6	739.302	41	<1.95	27.7	<10,24	0=0 points	0=0 points
									1.39323	1.16460		
SPH020	Male	Healthy	27.933	221.2	72.85	597.971	56	2.72001	6	7	0=0 points	0=0 points
SPH034	Male	Healthy	28.025	305.9	68.1	1106.255	47	<1.95	27.7	<10,24	0=0 points	0=0 points
				193.6	82.99			2.96728	7.65779			
SPH011	Male	Healthy	28.713	63	7	986.324	51	4	7	<10.24	0=0 points	0=0 points
				280.7	84.61				13.9232			
SPH005	Male	Healthy	29.297	89	1	656.205	52	<1.95	7	<10.24	0=0 points	0=0 points
SPH085	Male	Healthy	29.584	173.9	32.2	638.313	50	<1.95	24.7	<10,24	-	-
				177.3	49.86			3.46183	8.35941	1.08696		
SPH024	Male	Healthy	31.028	66	8	682.026	48	1	7	6	-	-
SPH056	Male	Healthy	35.834	171.5	112.1	1088.524	39	<1.95	54.8	<10,24	-	-

				UNI	С							
Selected				SECOND S	ORTIN	G FEMAL	ES					
					1	2	3	4	5			
	Sample	Sex	K&L L	K&L R	BMI	COMP	HA	PCIIP	Age	IL1b	TNFa	IL6
	CYU161	Female	2= 3-4 points	2= 3-4 points	24.61	240.34	52.8	100.62	79	<3.9	<23.4	<10.2
	CYU129	Female	2= 3-4 points	2= 3-4 points	27.04	229.19	21.03	26.5	50	<3.9	<23.4	<10.2
	CYU004	Female	2= 3-4 points	2= 3-4 points	27.68	122.94	16.87	179.11	81	<3.9	<23.4	<10.2
	CYU171	Female	2= 3-4 points	2= 3-4 points	28.52	142.96	212.59	152.07	69	<3.9	<23.4	<10.2
	CYU172	Female	2= 3-4 points	2= 3-4 points	29.3	227.94	255.66	168.62	80	<3.9	<23.4	<10.2
	CYU163	Female	2= 3-4 points	2= 3-4 points	29.38	255.32	43.45	95.61	73	<3.9	<23.4	<10.2
	CYU118	Female	2= 3-4 points	2= 3-4 points	29.72	342.64	40.29	214.22	66	<3.9	<23.4	<10.2
	CYU168	Female	2= 3-4 points	2= 3-4 points	31.22	282.46	20.35	17.88	63	<3.9	14.6	<10.2
	CYU175	Female	2= 3-4 points	2= 3-4 points	39.54	206.54	122.09	168.62	74	<3.9	45.43	<10.2
	CYU020	Female	2= 3-4 points	2= 3-4 points	40	187.5	32.51	176.56	64	<3.9	<23.4	<10.2
	CYU009	Female	1= 1-2 points	3= 5-9 points	20.78	123.45	52	163.16	65	<3.9	<23.4	<10.2
	CYU174	Female	3= 5-9 points	3= 5-9 points	24.24	162.81	29.13	124.48	60	<3.9	33.7	168
	CYU127	Female	3= 5-9 points	3= 5-9 points	24.97	393.31	285.96	28.47	79	<3.9	<23.4	<10.2
	CYU110	Female	3= 5-9 points	3= 5-9 points	25	291.36	88.74	271.6	76	<3.9	<23.4	<10.2
	CYU167	Female	3= 5-9 points	2= 3-4 points	25.39	384.62	69.92	86.58	73	<3.9	<23.4	<10.2
	CYU173	Female	3= 5-9 points	3= 5-9 points	26.67	226	147.16	224.72	71	<3.9	<23.4	<10.2
	CYU002	Female	3= 5-9 points	3= 5-9 points	26.95	265.46	176.5	220.56	70	<3.9	<23.4	<10.2
	CYU013	Female	3= 5-9 points	3= 5-9 points	26.95	191.37	20.86	140.79	51	<3.9	<23.4	<10.2
	CYU102	Female	3= 5-9 points	3= 5-9 points	27.34	557.99	158.38	110.69	84	<3.9	<23.4	<10.2
	CYU023	Female	0= 0 points	3= 5-9 points	28.25	403.57	52.59	146.14	64	<3.9	<23.4	<10.2

CYU006	Female	3= 5-9 points	2= 3-4 points	28.65	333.75	81.12	189.23	74	<3.9	<23.4	<10.2
CYU176	Female	3= 5-9 points	2= 3-4 points	28.72	248.37	93.54	146.55	66	<3.9	<23.4	<10.2
CYU012	Female	1= 1-2 points	3= 5-9 points	29.14	420.4	66.21	177.38	77	<3.9	<23.4	<10.2
CYU113	Female	3= 5-9 points	3= 5-9 points	30.49	326.39	16.84	38.97	65	<3.9	<23.4	<10.2
CYU166	Female	3= 5-9 points	3= 5-9 points	31.14	410.62	7.13	88.22	51	<3.9	<23.4	<10.2
CYU162	Female	3= 5-9 points	3= 5-9 points	31.22	210.05	52.56	94.46	58	<3.9	<23.4	<10.2
CYU119	Female	3= 5-9 points	3= 5-9 points	31.63	104.03	14.73	8.77	80	<3.9	<23.4	<10.2
CYU112	Female	1= 1-2 points	3= 5-9 points	32.44	344.45	28.33	67.62	75	<3.9	<23.4	<10.2
CYU114	Female	3= 5-9 points	3= 5-9 points	32.81	334.47	19.74	177.12	77	<3.9	<23.4	<10.2
CYU024	Female	1= 1-2 points	3= 5-9 points	33.2	478.98	81.26	175.82	67	<3.9	<23.4	<10.2
CYU157	Female	3= 5-9 points	3= 5-9 points	33.2	402.48	124.31	102.67	66	<3.9	<23.4	<10.2
CYU160	Female	3= 5-9 points	3= 5-9 points	33.98	421.83	54.32	11.23	79	<3.9	22.9	<10.2
CYU137	Female	3= 5-9 points	3= 5-9 points	34.96	315.97	172.72	168.5	74	<3.9	<23.4	<10.2
CYU125	Female	3= 5-9 points	1= 1-2 points	44.44	184.55	86.08	38.32	74	<3.9	<23.4	10.6
CYU003	Female	3= 5-9 points	3= 5-9 points	46.88	539.1	37.03	164.39	72	<3.9	<23.4	78.4
CYU159	Female	0= 0 points	4= 10 points	29.38	519.48	154.04	14.59	73	<3.9	<23.4	<10.2
CYU165	Female	4= 10 points	3= 5-9 points	31.22	727.33	127.94	179.5	67	<3.9	<23.4	<10.2
CYU136	Female	4= 10 points	3= 5-9 points	33.3	786.52	201.68	266.03	84	<3.9	<23.4	<10.2
CYU104	Female	4= 10 points	4= 10 points	37.04	492.09	147.57	104.61	60	<3.9	<23.4	<10.2
	1		SECOND	SORTIN	IG MALE	S					
				1	2	3	4	5			
 Sample	Sex	K&L L	K&L R	BMI	COMP	HA	PCIIP	Age	IL1b	TNFa	IL6
CYU170	Male	2= 3-4 points	3= 5-9 points	27.47	289.89	120.17	484.96	72	<3.9	<23.4	<10.2
CYU005	Mala										<10.2
	Male	1= 1-2 points	3= 5-9 points	28.09	586.12	131.7	212.99	81	<3.9	<23.4	<10.2
CYU106	Male	1= 1-2 points 3= 5-9 points	3= 5-9 points 3= 5-9 points	28.09 29.41	586.12 281.41	131.7 46.67	212.99 130.59	81 75	<3.9 <3.9	<23.4 <23.4	<10.2
CYU106 CYU117	Male Male	1= 1-2 points 3= 5-9 points 2= 3-4 points	3= 5-9 points 3= 5-9 points 3= 5-9 points	28.09 29.41 30.93	586.12 281.41 293.03	131.7 46.67 78.92	212.99 130.59 274.31	81 75 50	<3.9 <3.9 <3.9	<23.4 <23.4 <23.4	<10.2 <10.2 <10.2
CYU106 CYU117 CYU111	Male Male Male Male	1= 1-2 points 3= 5-9 points 2= 3-4 points 3= 5-9 points	3= 5-9 points 3= 5-9 points 3= 5-9 points 2= 3-4 points	28.09 29.41 30.93 31.14	586.12 281.41 293.03 546.05	131.7 46.67 78.92 142.11	212.99 130.59 274.31 58.92	81755084	<3.9 <3.9 <3.9 <3.9	<23.4 <23.4 <23.4 <23.4	<10.2 <10.2 <10.2 <10.2
CYU106 CYU117 CYU111 CYU109	Male Male Male Male Male	1= 1-2 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 2= 3-4 points	3= 5-9 points 3= 5-9 points 3= 5-9 points 2= 3-4 points 3= 5-9 points	28.09 29.41 30.93 31.14 31.96	586.12 281.41 293.03 546.05 323.62	131.7 46.67 78.92 142.11 43.72	212.99 130.59 274.31 58.92 194.2	8175508459	<3.9 <3.9 <3.9 <3.9 <3.9	<23.4 <23.4 <23.4 <23.4 <23.4	<10.2 <10.2 <10.2 <10.2 <10.2
CYU106 CYU117 CYU111 CYU109 CYU124	Male Male Male Male Male	1= 1-2 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 2= 3-4 points 1= 1-2 points	3= 5-9 points 3= 5-9 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 3= 5-9 points	28.09 29.41 30.93 31.14 31.96 36.9	586.12 281.41 293.03 546.05 323.62 556.97	131.7 46.67 78.92 142.11 43.72 95.49	212.99 130.59 274.31 58.92 194.2 43.24	 81 75 50 84 59 66 	<3.9 <3.9 <3.9 <3.9 <3.9 <3.9	<23.4 <23.4 <23.4 <23.4 <23.4 <23.4	<10.2 <10.2 <10.2 <10.2 <10.2 <10.2
CYU106 CYU117 CYU111 CYU109 CYU124 CYU019	Male Male Male Male Male Male	1= 1-2 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 2= 3-4 points 1= 1-2 points 3= 5-9 points	3= 5-9 points 3= 5-9 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 3= 5-9 points 3= 5-9 points	28.09 29.41 30.93 31.14 31.96 36.9 37.98	586.12 281.41 293.03 546.05 323.62 556.97 232.62	131.7 46.67 78.92 142.11 43.72 95.49 54.57	212.99 130.59 274.31 58.92 194.2 43.24 173.6	81 75 50 84 59 66 66	<3.9 <3.9 <3.9 <3.9 <3.9 <3.9 <3.9	<23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4	<10.2 <10.2 <10.2 <10.2 <10.2 <10.2 <10.2
CYU106 CYU117 CYU111 CYU109 CYU124 CYU019	Male Male Male Male Male Male	1= 1-2 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 2= 3-4 points 1= 1-2 points 3= 5-9 points	3= 5-9 points 3= 5-9 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 3= 5-9 points 3= 5-9 points	28.09 29.41 30.93 31.14 31.96 36.9 37.98	586.12 281.41 293.03 546.05 323.62 556.97 232.62	131.7 46.67 78.92 142.11 43.72 95.49 54.57	212.99 130.59 274.31 58.92 194.2 43.24 173.6	81 755 84 59 666 666	<3.9 <3.9 <3.9 <3.9 <3.9 <3.9 <3.9	<23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4	<10.2 <10.2 <10.2 <10.2 <10.2 <10.2 <10.2 <10.2
CYU106 CYU117 CYU111 CYU109 CYU124 CYU019	Male Male Male Male Male Male Male	1= 1-2 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 2= 3-4 points 1= 1-2 points 3= 5-9 points 3= 5-9 points	3= 5-9 points 3= 5-9 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 3= 5-9 points 3= 5-9 points 4= 10 points	28.09 29.41 30.93 31.14 31.96 36.9 37.98 25.95	586.12 281.41 293.03 546.05 323.62 556.97 232.62 995.47	131.7 46.67 78.92 142.11 43.72 95.49 54.57 57.05	212.99 130.59 274.31 58.92 194.2 43.24 173.6 21.82	81 75 50 84 59 66 66 66 71	<3.9 <3.9 <3.9 <3.9 <3.9 <3.9 <3.9	<23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4	<10.2 <10.2 <10.2 <10.2 <10.2 <10.2 <10.2 <10.2 <10.2
CYU106 CYU117 CYU109 CYU124 CYU019 CYU019 CYU123 CYU125	Male Male Male Male Male Male Male Male	1= 1-2 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 2= 3-4 points 1= 1-2 points 3= 5-9 points 3= 5-9 points 4= 10 points	3= 5-9 points 3= 5-9 points 3= 5-9 points 2= 3-4 points 3= 5-9 points 3= 5-9 points 3= 5-9 points 4= 10 points 4= 10 points	28.09 29.41 30.93 31.14 31.96 36.9 37.98 25.95 33.06	 586.12 281.41 293.03 546.05 323.62 556.97 232.62 995.47 510.81 	131.7 46.67 78.92 142.11 43.72 95.49 54.57 57.05 171.61	212.99 130.59 274.31 58.92 194.2 43.24 173.6 21.82 83.05	81 75 50 84 59 66 66 66 71 80	<3.9 <3.9 <3.9 <3.9 <3.9 <3.9 <3.9 <3.9	<23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4 <23.4	<10.2 <10.2 <10.2 <10.2 <10.2 <10.2 <10.2 <10.2 24.94 <10.2

ANIMUS									
Selected	Number	Sex	BMI	Age					
	GRA007	Female	20.1	19					
	GRA033	Female	20.3	22					

GRA069	Female	20.8	21
GRA028	Female	20.8	20
GRA019	Female	21.5	15
GRA043	Female	21.8	21
GRA056	Female	23.0	21
GRA031	Female	23.3	33
GRA039	Female	24.3	18
Number	Sex	BMI	Age
GRA004	Male	20.9	25
GRA027	Male	21.1	24
GRA052	Male	22.3	38
GRA021	Male	22.9	-
GRA037	Male	24.5	21
GRA034	Male	24.9	27
GRA030	Male	25.0	31
GRA067	Male	26.0	37
GRA074	Male	26.0	32
GRA046	Male	26.6	29
GRA063	Male	26.6	26
GRA075	Male	26.7	43
GRA024	Male	27.8	29
 GRA026	Male	31.1	26

For example, Table 5 shows the quality control Cq results from HULAFE plasma samples. miRNA control from RNA extraction, UniSp2, UniSp4, and UniSp5, that were at 160 fmol, 1.6 fmol, and 0.016 fmol concentrations, showed an increase of mean of Cq values of 15.1, 21.7, and 29.6, respectively, that was equivalent to a 64-fold increase among UniSp2 and UniSp4, and 234-fold among UniSp4 and UniSp5. Similar results were obtained for all plates (results not shown). Also, cDNA synthesis control, UniSp6 (concentration not supplied), and cel-miR-39-3p (0.16 fmol) gave a mean Cq of 16.3 and 24.7, respectively. Comparing values from the former UniSp4, UniSp5 and cel-miR-39-3p, the concentration was, as expected, 7.8-fold among UniSP4 and cel-miR-39-3p, and 29-fold among cel-miR-39-3p and UniSp5.

T	able .	5–	Partial	set oj	f S amples	included	in the	miRNA	quality	analysis.
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Sampl	mir10	mir19	mir45	mir23	UniSp	UniSp	UniSp	UniSp	cel-miR-39-	UniSp	mir12	miR30	
e	3	1	1	a	6	2	4	5	3р	3	4	с	most tissues
1	31.7	30.6	24.52	30.64	16.26	14.81	21.55	29.3	25.11	19.66	29.12	32.14	brain
2	30.66	26.89	24.09	28.14	16.57	15.03	21.86	29.16	25.1	18.56	33.05	31.67	kidney
3	29.01	29.2	21.46	28.49	16.49	16.16	22.95	30.04	25.36	15.79	24.32	29.83	blood mRNA
4	30.36	30.51	24.68	28.97	16.34	14.84	21.5	29.3	25.12	19.73	32.47	32.24	extraction
6	26.7	25.46	20.02	25.76	16.32	14.89	21.51	28.64	19.5	9.79	21.15	26.28	cDNA synthesis
7	30.02	27.53	22.68	29.19	16.65	14.83	21.49	29.14	25.04	19.78	33.33	32.12	PCR efficiency
9	31.52	30.22	23.16	23.82	15.89	12.88	21.44	28.77	24.48	11.48	25.1	27.87	
10	28.32	28.78	23.95	3 0	16.32	14.74	21.47	29.17	25.01	19.31	32.85	33.62	
11	30.62	24.76	23.87	23.96	15.5	14.7	22.07	28.45	24.8	10.96	26.27	28.94	
12	27.73	25.82	22.24	26.71	16.45	14.53	21.09	28.94	25.12	19.53	33.08	31.61	
13	30.56	29.01	23.56	24.23	16	14.72	16.01	28.95	23.06	11.08	25.12	31.89	

16	22.36	16.85	21.88	27.32	16.53	15	21.7	29.67	25.15	19.68	29.04	31.31
17	28.73	29.14	22.28	26.89	15.11	8.99	21.42	28.9	25.44	19.91	22.14	30.35
18	25.32	17.47	23.59	29.43	16.55	15.03	21.79	29.57	25.32	19.61	30.25	31.76
19	28.12	28.8	21.84	26.97	12.63	15.23	21.65	29.11	20.64	10.78	19.37	21.59
20	28	21.14	21.79	26.95		15.27	21.68	29.56	25.53	19.67	28.03	29.78
21	29.73	29.05	22.83	23.07	16.55	15.83	18.98	29.12	24.31	8.7	31.5	31.44
22	28.79	31.99	22.48	27.65	16.76	15.28	21.69	29.66	25.03	19.79	29.3	31.3
23	29.8	30.95	22.27	26.53	16.62	14.73	20.74		21.87	19.59	31.96	27.45
24	28.19	26.89	22.09	27.9	16.89	15.24	21.77	29.7	25.17	19.91	31.57	30.86
25	28.35	29.5	21.5	27.53	16.44	14.54	21.3	28.92	24.35	19.86	23.56	30.64
26	27.88	26.66	21.68	26.93	16.48	15.14	21.88	29.49	25.27	19.86	31.21	29.05
27	29.04	26.65	21.91	26.83	16.51	16.51	21.47	28.17	25.69	12.78	30.22	29.89
28	27.12	30.68	23.3	27.55	16.56	16.81	23.07	30.48	25.25	19.56	31.79	31.99
29	29.61	28.04	26.84	28.14	16.56	15.85	22.81	30.75	24.8	19.54	25.91	30.77
30	28.14	25.19		26.64	16.73	14.83	21.79	29.51	25.77	19.47	33.3	31.46
31	30.61	30.6	23.76	29.29	16.8	16.44	23.69	31.42	25.26	18.14	27.76	31.16
32	27.91	27.42	19.75	24.98	16.69	15.47	22	29.63	25.86	18.71	32.1	31.23
33	28.92	24.96	21.58	26.88	16.75	16.17	22.92	30.34	25.35	20.36		37.68
34	28.61	29.33	21.18		15.89	16.09	23.14	29.86	25.22	19.72	31.21	32.35
35	28.46	29.18	21.94	27.44	16.62	16.26	22.93	31.23	25.12	19.62	29.32	29.8
36	29.75	30.49	22.22	27.92	16.76	16.58	23.11	31.8	25.37	19.51	31.18	31.84
Mean	28.77	27.49	22.61	27.19	16.30	15.11	21.70	29.57	24.67	17.51	28.92	30.68
SD	1.8	3.6	1.4	1.8	0.8	1.4	1.4	0.9	1.4	3.7	3.9	2.6

miR-451 and miR-23a-3p are found in plasma and serum and serve as haemolysis and internal control markers, respectively. Even more, they have been described also as cancer biomarkers (Karimi et al., 2017; Bai H et al., 2019) and miR-451 was increased in the OA articular cartilage (BD, Schwartz Z. Regulation of inflammatory and catabolic responses to IL-1 β in rat articular chondrocytes by microRNAs miR-122 and miR-451 and miR-23a promotes OA (Kang L et al., 2016). Then, it is expected to detect both miRNAs in serum samples and urine samples at the same levels as indicated by the commercial supplier (Qiagen – miRCURY LNA miRNA QC PCR Panel Handbook). Taken all values together (males and females in all samples, and control vs early OA in HULAFE samples), no differences were observed for plasma samples, but in urine samples, miR-451 Cq was much lower and miR-23a slightly higher (Table 6).

Table 6 - Whole Cq mean values of miR-451 and miR-23a from the three sample origins. Male and females were pooled. Values from Qiagen were taken from the miRCURY LNA miRNA QC PCR Panel Handbook

		miR-451	miR-23a
OIACEN	Urine	23	30
QINGEN	Plasma	21	28
	HULAFE urine	30.1 ± 3.8	26.7 ± 2.9
O'Activo	HULAFE plasma	22.0 ± 1.4	29.0 ± 1.8
Oncuve	CYU plasma	22.8± 1.6	29.5 ± 1.1
	Animus plasma	24.2 ± 0.7	29.5 ± 0.6

As a preliminary analysis using values of the quality control analysis, miR-451 and miR-23A values from HULAFE healthy and Early OA, UNIC K&L L/K&L R 2, K&L L/K&L R 3 and K&L L/K&L R 4, and ANIMUS were compared. Males and females were analyzed separately. Figure 3 shows the results from males and females. Mir-23a was significantly overexpressed in males and females K&L L/K&L R 2 and K&L L/K&L R 3 compared to healthy volunteers and patients with early OA, and in males, also ANIMUS samples were overexpressed. However, miR-451 is underexpressed in K&L L/K&L R 4 in female patients, but not in male patients, where ANIMUS athletes showed an overexpression. Results from miR-23a are expected according to the literature, but not for miR-451, that it has been described its overexpression in OA (Bottani M et al. 2020). However, the type of samples -synovial liquid vs plasma in this study- could affect our results. Also, this is a preliminary study but a limited number of samples, so it is necessary to have complete data from all participants included in this study.



Figure 3 – Graphical representation of Cq values of miR-23a and miR-451 from healthy volunteers and patients with early OA (HULAFE), UNIC patients, and athletes with risk of OA (ANIMUS). Ordinary one-way ANOVA statistical analysis was done with Graph Prism 9.0.

5.5 TaqManTM OpenArrayTM miRNA analysis

This assay is designed for the analysis of miRNA expression levels by real-time PCR. Each array contains 754 different miRNAs. Three samples/array were run per plate and a total of 102 samples are under analysis. At this time, 90 samples were run and are under analysis. Figure 4 outlines the procedure. The protocol consists of 102 samples, representing males and females, all OA stages, and K&L L/K&L R scores, together with the BMI. miRNA that correlated with OA stages combined with BMI for males and females, will be selected for further validation with the rest of the samples.

Workflow Prepare cDNA templates Input RNA sample Perform the poly(A) tailing reaction (page 17) Perform the adaptor ligation reaction (page 18) Perform the reverse transcription (RT) reaction (page 19) Perform the miR-Amp reaction (page 20) Prepare and run assays with fixed-content Prepare and run assays with flexible-content and custom-configured OpenArray" plates OpenArray" plates Generate OpenArray plate layouts in the QuantStudio 12K Flex Software Generate 384-well sample plate layouts in the OpenArray Sample Tracker Software (page 31) (page 21) Set up the real-time PCR reactions in an Set up the PCR reactions in an OpenArray[~] OpenArray 384-well Sample Plate (page 22) 384-well Sample Plate (page 32) . Set up the AccuFill[~] instrument (page 24) Set up the AccuFill[~] instrument (page 33) Transfer reactions to the OpenArray plate Transfer reactions to the OpenArray plate (AccuFill[~] instrument) (page 25) (AccuFill[~] instrument) (page 25) Seal the OpenArray plate (page 26) Seal the OpenArray plate (page 26) Run the OpenArray plate(s) on the Run the OpenArray plate(s) on the QuantStudio 12K Flex instrument (page 36) QuantStudio 12K Flex instrument (page 27) Check the QC images (page 28) Check the QC images (page 28) Export and review data (page 40)

Figure 4. Workflow of miRNA analysis. From the handbook "TaqMan[™] Advanced miRNA Assays – User guide" Applied Biosystems – Thermo Fisher scientific.

5.6 Deviation from the expected work

There have been deviations from the proposed work in Task 4.3.

- It was expected to include in the study the protein and miRNA exosome content. However, due to the small amount of plasma and urine samples sent, it was decided to concentrate the study on miRNA, together with the standard biomarkers included in Task 4.2 for HULAFE samples.
- COVID-19 restrictions, mainly during the lockdown on March-June 2020 and mobility restrictions from the rest of the year, forced a delay both in recruiting patients and laboratory work, so it was not possible to follow the expected schedule. For this reason, at this time we have run all samples and we are in the process of analyzing the Taqman OpenArray results. During this second quarter, we plan to validate the selected miRNA and run all samples for the final correlation analysis with clinical and biomarkers data generated and transfer the whole data to the consortium partners involved in computer modelling.

6 Microbiome analysis

6.1 Introduction

The gut microbiota has a critical role in the immune system's development, function, and inflammatory reactions (Rooks and Garrett, 2016). Perturbations in the microbiome can activate the innate immune system and may lead to enhanced production of pro-inflammatory cytokines which could affect multiple organs, including the joint (Boer et al., 2019). The intestinal microbiome composition is also highly associated with a large number of pathological disorders like obesity (Isolauri, 2017), inflammatory bowel diseases (Serban, 2015), aging (Aleman and Valenzano, 2019) and metabolic syndrome (Collins *et al.*, 2015). Many of the pathologies listed are shared between OA and microbiota imbalance. Consequently, there is growing evidence suggesting that changes in the gut microbiome may also be linked to developmental factors in the onset of musculoskeletal disorders such as OA.

The study of the gut microbiome and its relationship with the development of OA is an emerging area of study. The few published research papers to date include mostly murine models fed with a high-fat diet in order to induce obesity and to affect joint integrity. Very few studies involving human microbiome intestinal samples have been carried out to date, and all of them have been conducted in subjects diagnosed with OA. For this reason, there is an interest in exploring the relationship between the gut microbiome and the onset of OA in order to investigate whether there are existing early indicators/markers for the initiation of the OA process. In the course of this experimental research, 82 stool samples were provided by volunteers from La Fe University Hospital, in Valencia. Twenty of them are part of the control group without OA and 62 samples belong to volunteers with an initial OA.

6.2 Material and methods

6.2.1 DNA extraction

Fecal human samples were collected in Stool Nucleic Acid Collection and Preservation Tubes (Cat. 45630, 45660) from Norgen Biotek. The samples were stored at room temperature and shipped to LEITAT facilities. Total genomic DNA was extracted from a 150 mg fecal sample aliquot using a repeated bead beating method from ZymoBIOMICS DNA/RNA Miniprep Kit following the supplier instructions. The DNA extraction concentration was measured using Qubit 4 Fluorometer from Thermo Fisher.

6.2.2 16S rRNA V3-V4 metagenomic sequencing

Metagenomic analysis was performed using Illumina Miseq sequencing platform according to standard protocols devised by Illumina. The V3-V4 region of 16S rRNA with 515F and 806R primers designed for

dual indexing genes was targeted to characterize and estimate bacterial communities present in faecal samples. Polymerase chain reaction (PCR) was performed in 25µl reaction volume containing 1µl of template DNA (5ng/µl), 5µl of each primer (1µM) and 12.5 µl of 2x KAPA HiFi HotStart ReadyMix. Miseq amplicons library preparation. PCR was carried out under the following conditions: initial denaturation for 47 min at 94°C, followed by 25 cycles of denaturation for 45 sec at 94°C, annealing for 60 sec at 52°C and elongation for 90 sec at 72°C, and a final elongation step for 10 min at 72°C. Duplicates were combined, purified with the NucleoSpin Gel and PCR Clean-up kit (Macherey-Nagel), and quantified using the Quant-iT PicoGreen dsDNA kit (Invitrogen). Equimolar concentrations along with Illumina sequence adapter and index primers were utilized in emulsion PCR to generate amplicon libraries followed by PCR clean up. For bioinformatic analysis, sequences were imported into QIIME2 q2cli v2019.10 (Bolyen et al., 2019). The overall quality was inspected with the demux plugin (https://github.com/qiime2/q2-demux), and the total of reads was processed with DADA2 (Callahan et al., 2016) in order to remove trim reads, correct errors, merge reads pairs and remove PCR chimeras and to finally obtain representative ASV sequences and the abundances of each ASV.

The representative sequences were classified taxonomically using the q2-feature-classifier classifyconseunsus-vsearch(<u>https://github.com/qiime2/q2-feature-classifier</u>). SILVA v132 16S-only reference sequences and taxonomy were downloaded from the SILVA website for QIIME. SILVA v132, clustered at 99% database was used as reference reads.

Finally, the feature table obtained was normalized with q2-gcn-norm (<u>https://github.com/Jiung-Wen/q2-gcn-norm</u>). This plugin normalizes sequences by 16S rRNA gene copy number (GCN) based on rrnDB database.

6.2.3 Statistical analysis

Statistical analysis was performed using R 4.0.3 (R: *The R Project for Statistical Computing*,) and QIIME2 (Bolyen *et al.*, 2019).

6.3 Results

This Chapter outlines all the results obtained regarding the OActive deliverable 4.2.

6.3.1 OA pilot volunteers profile

A total of 82 volunteers agreed to provide a stool sample for the OActive pilot study. The samples were collected at LaFe University Hospital in Valencia and were shipped to LEITAT's facilities where they were properly stored and processed. The female group is higher than the male group 67% vs 33%. These results are consistent with the higher incidence of OA in women compared to men.

 Table 7 Depicted are the mean and \pm the Standard deviation. P-values were obtained using the Mann-Whitney U test for comparison between groups.

 BMI – Body Mass index, WOMAC - Western Ontario and McMaster Osteoarthritis Index, LDL-C – Low-density lipoprotein cholesterol.

	1	Females	Males			
	Initial OA	Healthy	p-value	Initial OA	Healthy	p-value
Cohort participants	39	16	-	23	4	-
Age (years)	$52,17 \pm 5,71$	$52,75 \pm 6,32$	0,75	49,69 ± 3,30	49,69 ± 4,76	0,86
BMI (kg/m2)	$26,30 \pm 4,96$	$28,32 \pm 4,22$	0,194	$26,80 \pm 2,95$	$26,34 \pm 3,88$	0,78
WOMAC (pain score)	$2,05 \pm 1,86$	1,21 ± 1,39	0,032	$1,29 \pm 1,18$	$0,22 \pm 0,33$	0,0022
LDL-C (mg/dL)	$106,13 \pm 25,27$	99,68 ± 23,82	0,37	$111,34 \pm 28,14$	113,25 ± 23,04	0,95
Triglycerides (mg/dL)	$103,02 \pm 42,72$	$118,43 \pm 63,27$	0,38	$121,3 \pm 54,58$	83,75 ± 37,17	0,11

No differences were observed between the Healthy and Initial OA groups in age, BMI, lipoprotein levels, and blood triglycerides. However, a significant difference in the WOMAC score was observed. WOMAC is one of the scores used to diagnose OA. The difference in the male group is highly significant (p=0.0022) while in the female group, due to high variability, this difference is lower, but still significant (p=0.032).

6.3.2 Bacterial communities taxonomic profile

For the 82 faecal samples, genomic DNA extractions were performed to amplify the V3 -V4 hypervariable regions from the 16S rRNA gene. After quality control, 81 samples were included in the post-analysis and one sample was discarded due to not reaching the minimum number of sequences per sample required. 3,145,458 sequences were obtained after quality filtering, corresponding to 6195 unique ASVs (Amplicon Sequence Variant). The results of the bacterial taxonomy assignation classified 5945 ASVs to the genus level and 5188 to the species level.

Figure 4 illustrates the relative abundances at the phylum level for each of the individual samples as well as the mean composition for the Healthy and Initial OA groups. Section A of figure 5 shows the interindividual differences between the volunteers. The bacterial taxonomic composition obtained for this cohort is very similar to other studies with cohorts of European Caucasians. (Boer et al., 2019)(Gupta, Paul, and Dutta, 2017) (Jackson *et al.*, 2018).



Figure 5 Taxa plot at phylum level for A) each individual sample and B) collapsed for all samples.

The most abundant phylum for each group is Firmicutes with 86.66% of the total abundance for the Initial OA group and 90.29% for the Healthy group, followed by Bacteroidetes 7.7% and 2.7% respectively. These two phyla are the most abundant in faecal samples extracted from adult individuals.

At the genus level, Figure 5 highlights the 20 most abundant genera representing about 85% of the relative abundance of the bacterial communities. The two most abundant genera found are *Faecalibacterium* and *Blautia*. Both genera are widely described in the state of the art and their variation is associated with different pathologies such as Crohn's disease. (Fujimoto *et al.*, 2013) Colorectal Cancer (Tilg *et al.*, 2018) for *Faecalibacterium* and obesity (Ozato *et al.*, 2019) for *Blautia*. Nevertheless, no differences in the abundance of these two genera are observed between the Healthy and Initial OA groups.



Figure 6 Boxplot for the 20 most abundant genera detected.

6.3.3 Alpha and Beta diversity

Alpha diversity can be used to calculate the diversity of each sample in terms of the number of different ASVs and beta diversity calculate the similarity between two samples or a group of samples, taking into account the phylogenetic distance between the different ASVs obtained.

Figure 7 shows the alpha diversity (Shannon index) of the bacterial communities from the 82 fecal samples (OA and Control groups). No significant differences in alpha diversity were detected between the Healthy and Initial OA groups. These results are consistent with those obtained in other similar studies involving patients diagnosed with OA and Control groups (Boer *et al.*, 2019) (Coulson *et al.*, 2013).



Figure 7 Box plot with sample distribution for Healthy and Initial OA group. Mann-Whitney U test was used to compare the differences between groups.

Beta-diversity analyses were also performed to investigate composition differences among the fecal bacterial communities from the samples corresponding to the two groups under study (Initial OA and Healthy). Comparisons between the Initial OA and Healthy group were made using 4 different beta-diversity metrics. Figure 8 plots the beta-diversity distance between the various microbiome samples. For none of the four metrics chosen a separation into distinct clusters is apparent, as both Initial OA and Healthy group bacterial communities are largely mixed.



Figure 8 Beta-diversity plots labelled by Initial OA and Healthy subjects. A. Weighted Unifrac, B. Unwaghted Unifrac, C. Jaccard and D. Bray Curtis

To confirm the hypothesis of no compositional differences between the two groups, a PERMANOVA test was performed with 999 permutations for each of the beta-diversity metrics. The results can be found in table 8. None of the 4 metrics obtains a significant p-value result. This result confirms that there are no significant differences in the composition of the bacterial microbiome between the Healthy group and the Initial OA group.

Table 8 PERMANOVA results after 999 permutations for the four betadiversity metrics analysed.

PERMAN	JOVA	RESIL	TS
	$V \cup V \perp I$	MLSUL	110

				1110 0 110	
	Sample	Number of	Test	p-value	Number of
	size	groups	statistic		permutations
Weighted Unifrac	81	2	1,366	0,219	999
Unweighted Unifrac	81	2	1,4192	0,127	999
Jaccard	81	2	0,954	0,604	999
Bray Curtis	81	2	0,965	0,861	999

6.3.4 Biomarker analysis

The last analysis was focused on the search for possible biomarkers using the variables generated from the volunteers' microbiome. A good biomarker should be more abundant in the Initial OA group over the Healthy group and should be able to predict the onset of OA. This is one of the main objectives of the project. Two approaches have been used to carry out this study. On the one hand a machine learning algorithm (Random Forest) and on the other hand a multivariate statistical comparison (Welch t-test). (Alekseyenko, 2016). The relative abundances between the Healthy group and the Initial OA group were compared for the 240 bacterial genera identified.

Random Forest (Breiman, 2001) is a machine learning method based on a decision tree algorithm, that can be used for classification and regression-based analysis. Random Forest is widely used in the discovery of new microbiome-associated biomarkers (Li et al., 2020) (Ren et al., 2019) (Wang, Xu and Xia, 2016)(Sprockett et al., 2019). Typically, around two-thirds of a particular study sample is used for the model fitting or training while the remaining one-third is used for model testing.

Receiver Operating Characteristic (ROC) curves are a graphical representation of the classification accuracy of a machine-learning model. The top-left corner of the plot represents the "optimal" performance position, indicating an FPR of zero and a TPR of one. A greater area under the curve (AUC) indicates better performance. This can be compared to the error rate achieved by random chance, which is represented in figure 9B as a diagonal line extending from the lower-left to upper-right corners. Additionally, the "steepness" of the curve is important, as a good classifier should maximize the TPR while minimizing the FPR.



Figure 9 A. Random forest classification model accuracy. B. The ROC curve plots showing the relationship between the true positive rate (TPR, on the y-axis) and the false positive rate (FPR, on the x-axis) "Micro-averaging" calculates metrics globally by averaging across each sample; hence class imbalance impacts this metric. "Macro-averaging" is another average metric, which gives equal weight to the classification of each sample.

Figure 9A illustrates the classification by the Random Forest machine learning model. After using the microbiome variables as input, the model is not able to correctly classify the different samples into the group to which each of them belongs. None of the 20 samples belonging to the Healthy group were classified as Healthy, all of them were included in the Initial OA group. The same tendency is observed in figure 5B, where the Area under the curve (AUC) is 0.51 for both groups, very close to 0.5 which is the value obtained when the samples are randomly classified (discontinuous diagonal line).

After analyzing the Random Forest model results, we can conclude that assessing complete bacterial microbiomes does not allow us to differentiate specifically between the two groups.

Finally, to further investigate possible genera that may play a role as indicators of the onset of OA, we have used the Welch multivariate test, specifically designed to be applied to unbalanced data sets, as in this case. This test reduces type I error.

After comparing the relative abundances across the ASVs belonging to 240 different genera, a total of 9 markers with a p-value lower than 0.05 have been detected. Eight of these markers are enriched in the Initial OA group and only one in the Healthy group. However, after adjusting the p-values obtained by Benjamini & Hochberg's false discovery rate (FDR) the p-values become non-significant.

The relative abundance across the 9 potential biomarker candidates is shown in Figure 10.

Table 10 Welch test results between Healthy and Initial OA group. P values and p adjusted with False Discovery rate Benjamini & Hochberg for all the genera detected are highlighted.

	feature	enrich_group	diff_mean	pvalue	Padj (fdr BH)
marker1	[Eubacterium] hallii group	2. Initial OA	-0.00426	0.0146	Ns
marker2	Alistipes	2. Initial OA	-0.0044	0.0370	Ns
marker3	Anaerostipes	2. Initial OA	-0.00306	0.000940	Ns
marker4	Barnesiella	2. Initial OA	-0.000475	0.0141	Ns
marker5	Coprococcus 3	2. Initial OA	-0.00357	0.00480	Ns
marker6	Dialister	2. Initial OA	-0.00186	0.005881	Ns
marker7	Parabacteroides	2. Initial OA	-0.000551	0.00365	Ns
marker8	Paraprevotella	2. Initial OA	-0.000396	0.0107	Ns
marker9	Ruminococcus 1	1. Healthy	0.0191	0.0297	Ns



Figure 10 Box plot with relative abundance for the 9 marker genus detected after Multivariate Welch t-test.

6.3.5 Discussion

Microbiome studies about OA with samples from real individuals have a high degree of novelty. This is highlighted because most of the studies pointing to a possible cause-effect relationship between the microbiome and OA are conducted in mice, either germ-free (Ulici *et al.*, 2018) or mice fed with a high-fat diet (Rios *et al.*, 2019)(Collins *et al.*, 2015)(Schott *et al.*, 2018). Moreover, the studies correlating variation in the microbiome with OA using human samples include patients with advanced stages of OA. (Boer *et al.*, 2019) (Coulson *et al.*, 2013). Some other papers do not analyze the microbiome through 16S rRNA gene sequencing but establish possible correlations through LPS detected in blood as a metabolite generated by the microbiome and the WOMAC pain index. (Huang *et al.*, 2016).

The microbiome variables obtained were analyzed to identify differences between the Healthy and Initial OA groups. The microbiome diversity measured by the Shannon alpha diversity index shows no significant difference between the Healthy and Initial OA groups. These results are aligned with other studies that have looked at the alpha diversity obtained from OA patients. (Boer *et al.*, 2019).

Regarding beta diversity measures, none of the 4 indices studied (Weighted and Unweighted Unifrac, Jaccard, and Bray Curtis) differentiate the composition between the Healthy and Initial OA groups. This indicates therefore that the microbiome composition as a whole is not sufficiently distinct to differentiate between the Healthy and Initial OA groups. The same hypothesis is confirmed by applying the Random Forest classification model, yielding an 0,51 AUC for both groups. Such results of no significant difference in beta-diversity values are like those obtained by one of the most exhaustive studies performed to date in the Rotterdam cohort. In this study, 1444 participants were enrolled in the Rotterdam study-III with hip and/or knee osteoarthritis (Boer *et al.*, 2019). At first, no significant differences in beta diversity indices were detected, however, after choosing a subset of 256 taxa, they found an association between increased WOMAC score and the relative abundance of ASVs corresponding to the genus *Streptococcus*, widely reported as proinflammatory.

The study leads in the direction of not only looking for differences at the community level, but also for taxa that may be valid predictors of disease development. Following this approach, the 240 genera detected in the OActive volunteers' samples were compared using the Welch multivariate test. This model returned a series of 9 markers with predictive potential in the development of OA, 8 of them being enriched in the Initial OA group. None of these 9 markers corresponds to the *Streptococcus* genus detected in the Rotterdam cohort. (Boer *et al.*, 2019). However, after adjusting the p-values obtained by the false Discovery rate, significant differences were found.

Looking at the relative abundance distribution for the 9 genera detected by the model, a high proportion of outliers can be identified, mainly in the Initial OA group. Due to the high inter-individual sample variability, these variations in the abundances of the detected genera lead to a decrease in the discovery of new potential biomarkers, it is expected that, by increasing the N with more volunteers, these differences should decrease, more genera will be differentially detected as potential biomarkers and the adjusted p-values after FDR will continue to be significant.

Nevertheless, based on data analyzed so far, we are unable to assert that the microbiome, or taxa belonging to the microbiome, in our database alone can reliably predict the development of OA disease.

However, once the data generated will be introduced together with the other set of variables obtained during the project into the combined database, perhaps some of the microbiome taxa variables in combination with the other data sets can help the predictive algorithm to make a better decision in predicting the disease development.

Further research is needed to confirm the possible relationship between new biomarker candidates and the onset of OA.

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