ABSTRACT

Since the introduction of in vitro fertilization (IVF) in clinical practice of infertility treatment, the indicators for high quality embryos were investigated. Cumulus cells (CC) have a specific gene expression profile according to the developmental potential of the oocyte they are surrounding, and therefore, specific gene expression could be used as a biomarker. The aim of our study was to combine more than one biomarker to observe improvement in prediction value of embryo development. In this study, 58 CC samples from 17 IVF patients were analyzed. This study was approved by the Republic of Slovenia National Medical Ethics Committee. Gene expression analysis [quantitative real time polymerase chain reaction (qPCR)] for five genes, analyzed according to embryo quality level, was performed. Two prediction models were tested for embryo quality prediction: a binary logistic and a decision tree model. As the main outcome, gene expression levels for five genes were taken and the area under the curve (AUC) for two prediction models were calculated. Among tested genes, AMHR2 and LIF showed significant expression difference between high quality and low quality embryos. These two genes were used for the construction of two prediction models: the binary logistic model yielded an AUC of 0.72 ± 0.08 and the decision tree model yielded an AUC of 0.73 ± 0.03. Two different prediction models yielded similar predictive power to differentiate high and low quality embryos. In terms of eventual clinical decision making, the decision tree model resulted in easy-to-interpret rules that are highly applicable in clinical practice.

Keywords: AMHR2 gene; Cumulus cells (CC); Embryo prediction; LIF gene, In vitro fertilization (IVF)

INTRODUCTION

Since the introduction of in vitro fertilization (IVF) in the clinical practice of infertility treatment, indicators of implantation potential of embryos have been researched. Despite significant improvements in assisted reproductive technology (ART), the success of IVF remains low. Although most of the oocytes retrieved after ovarian stimulation with gonadotropins in combination with gonadotro-pin-releasing hormone (GnRH) analogues are capable of fertilization, only half of them develop into embryos and even fewer implant [1]. Therefore, to increase pregnancy rate, more than one embryo is usually transferred, which can lead to multiple pregnancies and increased fetal and maternal morbidity and mortality [2]. Consequently, there is a need for identifying biomarkers that would serve as reliable indicators of high implantation potential of the embryos available for transfer [3].
The selection of embryos for transfer is currently based on the evaluation of subjective morphological criteria. These include evaluation of pronuclei in the zygote and early cleavage. On day 3 after oocyte retrieval, the embryo fragmentation, number and morphology of blastomeres, and the presence of multiple nuclei are evaluated; on day 5, blastocyst morphology is evaluated considering the Gardner and Schoolcraft system [4]. The presence of a mature (MII) and high-quality oocyte plays an essential role in the development of a high-quality embryo [5]. This means that the selection of high-quality embryos begins at the time of oocyte selection. The oocyte selection for fertilization is currently also based on morphological evaluation of the polar body, meiotic spindle, zona pellucida and cytoplasm [6]. There is increasing evidence that morphological evaluation is not a reliable predictor of oocyte competence and embryo implantation potential [7]; that is why there is a need to discover new, noninvasive, objective and reliable indicators of oocyte and embryo quality. Having reliable biomarkers for oocyte and embryo selection could be of special importance in selective embryo transfer to avoid a twin pregnancy. Lately, the most intense research is being carried out on genome analysis of cumulus cells (CC) and granulosa cells (GC) in order to discover biomarkers that would be predictive of oocyte and embryo developmental potential [8-10].

It is well known that there is intense bidirectional communication between oocytes and their surrounding CC and GC through gap junctions and paracrine signaling during folliculogenesis [11]. This communication is crucial for the development of a mature, developmentally competent oocyte. Instead of being a passive recipient of nutrients and regulatory signals from its surrounding CC and GC, the oocyte plays an active role in the secretion of paracrine factors that maintain an appropriate micro environment for the acquisition of its developmental competence [12]. This leads to functional changes in CC and GC which are crucial for the development of a quality oocyte [13]. In clinical practice, this means that these cells can serve as an indirect marker of oocyte quality. In IVF procedures, these cells are separated from oocytes and then discarded. They are easily accessible and plentiful, which makes them a perfect material for gene expression analysis in order to identify reliable and objective biomarkers of oocyte quality and embryo development potential [11].

Cumulus cells have been the subject of many studies in order to test whether oocyte quality is related to the expression of some of the growth differentiation factor 9 (GDF9)-dependent genes (HAS2, PTGS2 in PTX3) [3,14, 15]. Furthermore, CC have been analyzed in terms of gene expression related to the quality of embryo development. van Montfoort et al. [9] proposed a set of the following genes: CCND2, CXCR4, GPX3, CTNNB1, DHCR7, DVL3, HSPB1 and TRIM28 that have proven to be most variably expressed among the CC of the follicles with zygotes that underwent a rapid division, and the CC of those follicles the zygotes of which underwent a slow division [9]. Hamel et al. [16] proposed the following set of genes: FDX1, CYP19A1, CDC42, SERPINE2 and 3βHSD1 as those having the most variable expression among the GC from the follicles that resulted in pregnancy and those that did not.

In our previous study [10], we identified CC expression of AMHR2, LIF, SERPINE2, VEGFC and FSHR to be associated with blastocyst formation. In that study, LIF did not pass correction for multiple hypothesis testing, but due to its previous implication for oocyte maturation [17], we included it in our further analyses of CC expression. In this study, we used these genes to construct an embryo quality outcome model according to CC gene expression from oocytes that resulted in either high or low quality embryos.

MATERIALS AND METHODS

Patients and In Vitro Fertilization Treatment.

In this study, 17 patients undergoing the classical IVF cycle at the Department of Obstetrics and Gynecology, University Medical Centre, Ljubljana, Slovenia, were included. The study was approved by the Republic of Slovenia National Medical Ethics Committee (http://www.kme-nmec.si/) and patients signed a written consent form prior to study inclusion. It included patients who were less than 35 years old and with body mass index (BMI) between 17 and 26 kg/m². They attended the IVF program because of tubal factor infertility. The spermograms of their partners were normal, according to the World Health Organization (WHO) criteria.

As our previous study did not expose any differences in CC gene expression between patients who were treated with either GnRH agonists or antagonists in combination with recombinant follicle-stimulating
hormone (rFSH) [10], we used both GnRH analogs in the present study. Ten patients were administered GnRH agonist buserelin acetate (Suprefact; Hoechst AG, Frankfurt/Main, Germany) starting from day 22 with a daily dose of 0.6 ml (600 pg) subcutaneously. When criteria for ovarian desensitization were fulfilled (estradiol <0.05 nmol/L, follicles <5 mm in diameter), patients were subcutaneously administered 225 IU of gonadotropin follicitropin α (Gonal F; Industria Farmaceutica Serono S.p.A, Bari, Italy). The other seven patients received 225 IU of gonadotropin follicitropin α, subcutaneously administered on day 2. When the dominant follicle measured ≥14 mm in diameter, the GnRH antagonist cetrorelix acetate (Cetrotide; Asta Medica AG, Frankfurt, Germany) in a dose of 0.25 mg, was administered subcutaneously. Afterwards, all patients received 10,000 IU of the human chorionic gonadotropin (hCG) (Pregnyl; N.V. Organon, Oss, the Netherlands) when at least three follicles were ≥17 mm and serum oestradiol was ≥0.40 nmol/L per follicle; 34-36 hours later, ultrasound-guided transvaginal oocyte retrieval was performed.

**Cumulus Cells Collection and Oocyte Follow-Up.** Oocytes were removed from the follicular fluid. Immediately after oocyte retrieval, a small sample of CC of each oocyte was removed using a needle and a glass denudation pipette (Swemed, Göteborg, Sweden). Oocytes were not denuded by this technique. Obtained CC samples were washed in phosphate-buffered saline (PBS), snap frozen in liquid nitrogen and stored at −80 °C in vials until RNA isolation. The oocytes were further inseminated (classical IVF) and cultivated individually. After 24 hours, oocyte fertilization status was assessed. Fertilized oocytes were further cultured to the blastocyst stage in the Universal IVF Medium followed by the BlastAssist System (M1 and M2; Origio, Målov, Denmark) for 5 days. On day 5, at most two embryos at the blastocyst or morula stage were transferred into the uterus. Supernumerary blastocysts were cryopreserved.

**Experimental Design.** The models were built on the CC expression level values of five genes (AMHR2, LIF, SERPIN2, VEGFC and FSHR) of two kinds of embryos: high quality embryos (n = 26), represented by morula and blastocyst stage embryos on day 5, and low quality embryos (n = 36), represented by embryos which arrested in development any time within 5 days of cultivation after fertilization. The decision trees probabilistic model was used to estimate sensitivity, specificity and area under the curve (AUC) of a proposed model.

**Quantitative Real Time Polymerase Chain Reaction Analysis.** Quantitative real time PCR (qPCR) was used for CC gene expression using TaqMan Gene Expression pre designed assays (Applied Biosystems, Foster City, CA, USA). Peptidylprolyl isomerase B (PPIB) and 18s rRNA were added for normalization. Genomic DNA contamination was eliminated by DNAse treatment using DNAase I (F. Hoffmann-La Roche Ltd., Basel, Switzerland). cDNA for qPCR assays was prepared from 200 ng DNAsed RNA using SuperScript RT III (Invitrogen, Carlsbad, CA, USA) in a final volume of 20 µL. Following cDNA synthesis, RNAs-free water was added to increase the sample volume to 30 µL. Measurements were performed using a LightCycler 480 System (Roche Applied Science, Penzberg, Germany). Normalized mRNA levels were obtained by dividing the averaged, efficiency corrected values for mRNA expression by a normalization factor calculated from peptidylprolyl isomerase B (PPIB) and 18s RNA values and are expressed in arbitrary units. The resulting values were log2 transformed (log2-fold change) for comparison with microarray data.

**Statistical Analysis and Decision Tree Model Construction.** A Mann-Whitney U test and logistic regression model were performed with the Statistical Package for the Social Sciences (SPSS) version 17.0 (SPSS Inc., Chicago, IL, USA). For the decision tree model construction Orange® (www.orange.biolab.si) was used. Orange® is a data mining tool that works through preprogrammed widgets. Designing and testing decision trees is one of its functions.

**RESULTS**

The baseline characteristics between the two embryo groups (high quality vs. low quality) were not altered significantly by age, BMI and patients’ distribution tests in a binary logistic regression model (Table 1). The pregnancy rate for the observed group of patients was 0.53 and delivery rate was 0.47. Because RNA level from four CC samples of low quality embryos was too low, only 32 CC samples were analyzed, therefore, 58 CC samples were included in the predictive model construction. The Mann-Whitney U test showed significant difference between CC gene expression of oocytes resulting in high quality and
low quality embryos only in AMHR2 ($p = 0.030$). The data were then split according to AMHR2 expression (high or low, where the median was the cut-off value) and genes were further tested in each group (Table 2). In the AMHR2 gene, high expression group LIF was shown to differ significantly between high quality and low quality embryos ($p = 0.033$). Therefore, AMHR2 and LIF were taken for the construction of the embryo quality outcome prediction model.

**Binary Logistic Regression Model.** The AMHR2 and LIF CC expression values were used to construct three different binary logistic regression models for predicting a high quality embryo outcome. First, AMHR2 and LIF CC expression values were used in separate models, and their prediction values yielded an AUC of $0.69 \pm 0.08$ and $0.63 \pm 0.08$, respectively. Then, both genes CC expression values were combined into one model in which the prediction value yielded an AUC of $0.72 \pm 0.08$.

**Decision Tree Model.** The same procedure was used in a data mining protocol for constructing three decision trees with AMHR2 and LIF CC expression values. A simple data discretization was used for node splitting in the decision tree where expression values were stratified into two equal frequency intervals (high and low CC expression values). The decision tree model was tested using 50.0% of the data for learning and 50.0% data for testing. Testing was then repeated 100 times; median AUC and standard deviation were calculated. First, AMHR2 and LIF CC expression values were used to construct separate decision tree models, and their prediction values yielded an AUC of $0.67 \pm 0.01$ and $0.57 \pm 0.02$, respectively. Combining both genes resulted in a decision tree (Figure 1) with an AUC of $0.73 \pm 0.03$.

**DISCUSSION**

In the present study, we analyzed CC expression of AMHR2, LIF, SERPINE2, VEGFC and FSHR from oocytes that developed into either high quality or low quality embryos. In our previous study [10], selected genes were shown to be well differentiated between immature MI and mature MII oocytes, according to CC expression. In this study, we use them as potential biomarkers of embryo quality. Only AMHR2 and LIF were shown to be significant and were used in our prediction models. In either the binary logistic model or decision tree model, the predictive power was the best when both genes were used simultaneously.
In recent years, many studies have been performed to analyze CC gene expression in association with various endpoints: oocyte maturity, embryo development and pregnancy [8,9,16,18]. Since CC is an easily accessible material that is normally discarded during the IVF cycle, it represents a good biological material for research and hopefully, someday, also for diagnostic purposes. In previous studies on CC gene expression, many genes were shown to be differentially expressed between observed groups of oocytes, but not many were tested for predictive power [19]. As a main characteristic of a good diagnostic test is high predictive power, the AUC value (defined by high sensitivity and specificity), only genes which would yield a good predictive power, whether alone or in combination with other genes in repeated trials, would be suitable for potential CC gene expression diagnostic testing.

The AMH works through its receptor AMHR2, being the highest in the preantral follicle for their recruitment, and during follicle maturation it gradually diminishes [20-22]. It has also been proved that AMHR2 CC level decreases with the level of oocyte maturity and in the same manner is expressed by AMH in CC [10,23]. In this study, the CC expression of AMHR2 was significantly negatively correlated with embryo quality. This again indicates that oocyte maturity is a prerequisite for high quality embryo development. In this study two models were constructed with AMHR2 for high quality embryo prediction, and both showed similar predictive power. In comparison to models based on LIF CC expression, AMHR2-based models show better predictive power, which can be well-explained by higher CC expression differences between high and low quality embryos in AMHR2 compared to LIF.

The connection of LIF with reproduction was shown in the study where LIF was abundantly expressed in the uterine endometrial glands on day 4 of pregnancy [24]. Namely, the p53 protein regulates the LIF expression and sufficient LIF levels are crucial for embryo implantation [25,26]. In addition, the role of LIF in reproduction is not only in implantation but also in CC expansion. In the study of LIF function in in vitro matured human and mice cumulus-oophorus complex, it was proven that LIF supple-mentation induced cumulus expansion in both settings [17]. LIF also plays a role in blastocyst formation, where the group of bovine cumulus-oocyte complexes that were incubated with LIF yielded higher blastocyst development compared to the control group without LIF [27]. In our study LIF CC expression showed sig-

![Figure 1. A decision tree model diagram with AMHR2 and LIF. The model first separates CC samples upon AMHR2 expression (high or low) and then upon LIF expression (high or low). The blue color represents leaves with predominantly CC of low quality embryos and red color represents leaves with predominantly CC of high quality embryos. A combination of high AMHR2 and low LIF CC expression leads to an 82.6% possibility of developing a low quality embryo, and combination of low AMHR2 and low LIF CC expression leads to 74.6% possibility of developing high quality embryos.](image-url)
nificance between CC of high quality and low quality embryos only when a subgroup with high AMHR2 CC expression was observed. Therefore, using LIF alone in the model results in lower predictive power than AMHR2 but when used together, LIF improves the prediction value of AMHR2. Logistic regression represents the gold standard for constructing prediction models in biomedical studies. As its statistic is based on logistic regression, for each attribute the model computes a coefficient and combines them into one prediction variable. These are used for the receiver-operator curve (ROC) and computing AUC values, but as such, a prediction variable has no informational value for to user (i.e. clinician). Predictive data mining has become one of the essential tools for the researcher in medicine [28]. One of these techniques is also the decision tree, where a decision is made in each node, according to the value and predictive power of the variable. At the end (leaves) the probability of an event is given. Decision trees are usually represented with diagrams but can also be with “if sentences.” Diagrams make the decision tree model easy to interpret, therefore, a model with more informational value for the user.

In our study, both types of models, binary logistic and decision trees, resulted in similar AUC values, indicating that the type of the models used does not improve the predictive power. Analyzing the decision tree diagram of the model with AMHR2 and LIF leads to the conclusions that first, when both AMHR2 and LIF are low, there is a high possibility of the development of high quality embryos; second, when AMHR2 is high and LIF is low, there is a high possibility of developing low quality embryos, and third, all other combinations of AMHR2 and LIF expression result in the equal possibility of developing of high quality or low quality embryos. An equal chance of developing a high quality or low quality embryo upon AMHR2 and LIF expression actually means that the model is unable to predict the outcome, and this group contained exactly half of all observed embryos. Additional biomarkers would probably improve prediction for this group of embryos, or some other factors exist which we currently did not take into consideration, e.g. the quality of spermatozoa.

A logistic regression of CC expression was also used for constructing predictive models in the study by McKenzie et al. [29]. In their study, the expression values of hyaluronic acid synthase 2 (HAS2), cyclooxygenase 2 (PTGS2) and gremlin (GREM1) were used for constructing regression models for oocyte maturity, oocyte fertilization and embryo quality. Regression models for embryo quality yielded an AUC of 0.76, 0.76 and 0.81 for HAS2, PTGS2 and GREM1, respectively. Combining PTGS2 and GREM1 only slightly improved the predictive power (AUC 0.82 vs. 0.81). Besides PTGS2, the study by Wathlet et al. [30] also used six other genes and tested them for predictive power of cleavage stage embryo prediction and pregnancy prediction. Among the tested genes, the best cleavage stage embryo prediction relied on TPRM7 and ITPKA, but the AUC value was not calculated. Another prognostic model for pregnancy was published by Lager et al. [31], where 12 genes previously recognized by microarray, were tested by qPCR for their predictive power. They used a “signal to noise” ratio to assess the predictive value of a gene using weighted voting. The AUC value for pregnancy prediction was 0.76 ± 0.08.

CONCLUSIONS

In this study, AMHR2 correlated with high quality embryos, and its predictive power was higher when combined, with LIF. A set of highly predictive genes would probably result in a good prediction model where the decision tree model seems to have high clinical applicability.

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