Gene expression profile of whole blood cells differs in pregnant women with positive screening and negative diagnosis for gestational diabetes

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ABSTRACT

Objective: To evaluate the gene expression profile of whole blood cells in pregnant women without diabetes (with positive screening and negative diagnosis for gestational diabetes mellitus (GDM)) compared with pregnant women with negative screening for GDM.

Research design and methods: Pregnant women were recruited in the Diabetes Perinatal Research Centre—Botucatu Medical School-UNESP and Botucatuense Mercy Hospital (UNIMED). Distributed into 2 groups: control (n=8), women with negative screening and non-diabetic (ND, n=13), with positive screening and negative diagnosis of GDM. A peripheral blood sample was collected for glucose, glycated hemoglobin, and microarray gene expression analyses.

Results: The evaluation of gene expression profiles showed significant differences between the control group and the ND group, with 22 differentially expressed gene sequences. Gene networks and interaction tables were generated to evaluate the biological processes associated with differentially expressed genes of interest.

Conclusions: In the group with positive screening, there is an apparent regulatory balance between the functions of the differentially expressed genes related to the pathogenesis of diabetes and a compensatory attempt to mitigate the possible etiology. These results support the ‘two-step Carpenter-Coustan’ strategy because pregnant women with negative screening do not need to continue on diagnostic investigation of gestational diabetes, thus reducing the cost of healthcare and the medicalization of pregnancy. Although not diabetic, they do have risk factors, and thus attention to these genes is important when considering disease evolution because this pregnant women are a step toward developing diabetes compared with women without these risk factors.

INTRODUCTION

Hyperglycemia is one of the most common medical conditions that women face during pregnancy. The occurrence of gestational diabetes mellitus (GDM) is increasing globally in parallel with the increased prevalence of impaired glucose tolerance, obesity, and type 2 diabetes mellitus (T2DM).1

Despite the high prevalence of hyperglycemic disorders in pregnancy and long-term maternal effects, the most appropriate diagnostic criteria to be used to diagnose GDM are still under discussion.2 The ‘Pragmatic guide for diagnosis, management, and care’ (2015), International Federation of Gynecology and Obstetrics (FIGO), shows that, despite the efforts of numerous health organizations, including national and international associations in the areas of diabetes, endocrinology, and gynecology, to establish protocols, cut-offs, and algorithms for the diagnosis of GDM, current evidence is still lacking. These recommendations are criticized because of their lack of validation and because the expert opinions are often biased due to economic considerations or convenience,3 creating confusion and uncertainty among users. An underlying problem is that the cut-offs considered in the diagnosis of GDM take into account the risk of future development of T2DM; the results of a Hyperglycemia and Adverse Pregnancy
Outcome (HAPO) Study showed that the risk of maternal and perinatal adverse outcomes is associated with continuous hyperglycemia, without clear inflection points.4,5

The American Diabetes Association (ADA)9 provides two strategies for GDM diagnosis, namely a ‘one-step’ strategy using a 75 g oral glucose tolerance test (OGTT) and a ‘two-step’ strategy using a 50 g screening followed by a 100 g OGTT, and presents the recommendations with the respective evidence levels. The ‘one-step’ strategy assesses the fasting glucose 1 and 2 hours after glucose overload in pregnant women who are between 24 and 28 weeks of gestation without previous diagnosis of overt diabetes. Threshold values for blood glucose levels are as follows: fasting (92 mg/dL), 1 hour (180 mg/dL), and 2 hours (155 mg/dL). Any value equal to or above these values confirms the GDM diagnosis. In the ‘two-step’ strategy, a pregnant woman first takes 50 g of OGTT between 24 and 28 weeks of gestation with a limit value of 140 mg/dL, provided she has not previously been diagnosed with diabetes. Pregnant women with blood glucose levels that equal or exceed the 140 mg/dL limit during the first test go onto the second step involving 100 g of OGTT with the following limit values: fasting (95 mg/dL), 1 hour (180 mg/dL), 2 hours (155 mg/dL), and 3 hours (140 mg/dL), as defined by Carpenter-Coustan, or fasting (105 mg/dL), 1 hour (190 mg/dL), 2 hours (165 mg/dL), and 3 hours (145 mg/dL), as defined by National Diabetes Data Group (NDDG).7 During the second test, two or more values that are equal to or above the threshold values confirm the GDM diagnosis. The ADA concludes that different diagnostic criteria identified different degrees of maternal hyperglycemia and maternal and fetal risks, intensifying the debate about the best criteria to be used.

The Diabetes Perinatal Research Centre—Botucatu Medical School—UNESP diagnoses hyperglycemia in pregnancy using screening, with fasting blood glucose ≥90 mg/dL, and risk factors (personal, obstetric and family). Women positive for the screening diagnostic phase with 75 g OGTT and glycemic profile. Classifying the pregnant women in four groups identified by Rudge, including pregnant women with GDM and mild gestational hyperglycemia (MGH).

The literature describes that there are several genes related to diabetes. Moreover, it is known that the pathophysiology of GDM and T2DM is also related to genetic abnormalities, which are widely studied. In healthy individuals, as well as non-diabetic (ND) and non-pregnant populations, one-third of the variation in fasting glucose is genetic, and common genetic variants in multiple loci are robustly associated with fasting glucose, type 2 diabetes, and glycemic traits. Thus, genetic factors are likely contributing to the variation in glucose levels during pregnancy. However, these variants were not analyzed extensively in large studies with pregnant women.9 Genomics approaches have changed the way we do research in biology and medicine. It is possible to measure the majority of mRNAs, proteins, metabolites, protein–protein interactions, genomic mutations, polymorphisms, epigenetic alterations, and micro-RNAs in a single experiment.10 Developed molecular biology techniques lend themselves to the study of both normal physiology and pathophysiology,11 which brought great contributions of studies involving diabetes, pregnancies, and their complications. The study of gene expression on a large scale (microarray) makes it possible to monitor thousands of genes using a single test.12 The gene expression profile can capture daily changes caused by environmental factors and lifestyle, as well as permanent changes caused by structural variations in DNA.

In the current discussion about the best strategy for the diagnosis of GDM, particularly the strategy proposed by the ADA, which includes the ‘one-step’ and ‘two-step’ tests, one of the discussion points is whether pregnant women with positive screening results for GDM present important differences compared with pregnant women with negative screening results, a subject that is scarce in the literature. Knowing that GDM has been correlated with genetic alterations and changes in gene expression, the evaluation of the gene expression profile in pregnant women with positive screening results for GDM compared with pregnant women with negative screening results is extremely important. This information can separate two populations, where only the results of the screening have changed, and contribute to the current discussion focused on evaluating the best criteria for GDM diagnosis. Thus, the aim of this study is to evaluate the gene expression profile in whole blood cells of pregnant women without diabetes (with positive screening results and negative diagnosis for GDM) compared with pregnant women with negative screening results for GDM.

**RESEARCH DESIGN AND METHODS**

**Study design and study populations**

This study was approved by the Research Ethics Committee—Brazil Platform (CAAE: 14489013.0.0000.5411, number 291638). All patients were informed about the purpose of the study and signed a consent form before recruitment. Pregnant women were recruited between 2012 and 2015 at 34 weeks of gestation in the Diabetes Perinatal Research Centre—Botucatu Medical School—UNESP and Botucatuense Mercy Hospital (UNIMED). The women were divided into two groups: group 1—control (n=8), women with negative screening; group 2—ND (n=13), women with positive screening and negative diagnosis of GDM (normal OGTT and glycemic profile).

A questionnaire about personal information (age, smoking, alcohol consumption, contact with chemicals, radiation exposure) and medical history (intercurrent diseases, habitual use of drugs) was applied to all study participants. The risk factors present in groups with positive screening were as follows: ND group must have one or more risk factors for diabetes such as: fasting glucose levels >90 mg/dL, prior obesity, family history of
diabetes, maternal age over 25 years, obstetric history of previous GDM, fetal macrosomia, previous perinatal death and prior fetal malformation.

The inclusion criteria in the study were as follows: (a) pregnant woman classified into one of the study groups; (b) the ND group needs to present one or more risk factors for diabetes; (c) prenatal care and childbirth received at Botucatuense Mercy Hospital—UNIMED or the Diabetes Perinatal Research Centre—Botucatu Medical School-UNESP; (d) signed consent form; (e) fasting at the time of blood collection; (F) OGTT and glycemic profile between 24 and 28 weeks and (g) not in labor at the time of collection. Criteria for non-inclusion were as follows: (a) multiple pregnancies; (b) smoking; (c) alcoholic, (d) diabetes type 1 and (e) mental retardation. The exclusion criteria were as follows: (a) pregnant women with chronic and infectious diseases; (b) fetal malformations and (c) delivery before the 34th week.

**Anthropometric and biochemical measurements**

A peripheral blood sample was collected for glucose, glycated hemoglobin, and gene expression analyses. Plasma glucose was measured by the glucose oxidase method (Glucose—Analyzer II Beckman, Fullerton, California, USA), and the glycemic mean was calculated using the arithmetic mean of plasma glucose measured in all glycemic profiles performed at diagnosis (ND group); glycated hemoglobin was assayed by high-performance liquid chromatography (D10TM Hemoglobin Testing System, Bio Rad Laboratories, Hercules, California, USA). Body mass index was calculated by body weight divided by the square of height at the beginning and end of pregnancy. Part of the blood sample (2.5 mL) was collected in syringes and transferred immediately to a Blood RNA Tube (PAXgene), homogenized, stored at room temperature for 24 hours, and frozen gradually.

**RNA processing**

RNA extraction was performed using the PAXgene Blood RNA Kit (Qiagen) according to the manufacturer’s instructions. The concentration was assessed using NanoVue equipment. The concentration’s means and the RNA contamination rate were satisfactory, average yield 0.5 µg/µL, and purity index (ratio 260/280 and 260/230) above 1.8. The sample quality and integrity were evaluated by examining the bands corresponding to the 18S and 28S ribosomal subunits. Further, analysis using Bioanalyzer (Agilent) capillary electrophoresis equipment was performed to check the RNA integrity number (RIN), and samples with an RIN ≥ 7 were considered acceptable for microarray analysis.

**Microarrays**

The gene expression profile was evaluated using a single-color microarray. Glass slides were used (GE SurePrint G3 Human 8x60K Microarray Kit) and made by converting the RNA into complementary RNA (cRNA), which was labeled with cyanine (Cy3) using the 1-Color Low Input Linear Amplification Kit (Agilent). The labeled cRNA was purified with the RNeasy Kit (Qiagen) and subsequently eluted in ribonuclease-free water and quantitated. The cRNA fragmentation and hybridization steps (to the SureHyb hybridization chamber for 17 hours at 65°C) were made on slides using a Gene Expression Hybridization Kit. Following hybridization, the slides were washed with specific solutions. Agilent’s Stabilization and Drying solutions were used to protect the cyanine probes from ozone-induced degradation. Analysis of microarray slides was performed using Agilent Microarray Scan Control. Data extraction was performed using Agilent Feature Extraction (FE) and all parameters were evaluated, as shown by the Quality Control (QC) report.

**Statistical and bioinformatics analysis**

To evaluate the characteristics of the study population, a Student t-test was used. For microarray analysis, data quantification and QC were performed using FE software, V15.5 (Agilent Technologies, Life Sciences and Chemical Analysis Group, Santa Clara, California, USA). The filter, normalization, and analysis of expression data were loaded into the R-statistical environment (http://www.r-project.org), V3.0.0. The background adjustment was performed by subtracting the median background values from the median expression values. Data were processed using log2 and then normalized using the quantile function aroma.light package. The differentially expressed genes were identified using the F-test with Benjamini-Hochberg correction in order to compare between groups. These analyses were performed using the multtest package. All clusters of coregulated genes were subject to functional analyses using the database for annotation, visualization and discovery Integrated (DAVID), V6.7. Values of p<0.05 after Benjamini-Hochberg correction were considered significant.

After bioinformatics analysis, a literature review was performed that focused on all differentially expressed genes and developed biological networks identified in this study. We present a discussion of the genes that were directly or indirectly related to diabetes and its pathophysiology.

**Gene interactions networks evaluation**

The gene interaction networks were made using STRING: functional protein association networks (String-db.org) by inputting each differentially expressed gene of interest that presented interactions within each comparison. Confidence scores of 0.7 (high confidence) were used, and networks did not contain more than 50 interacting genes. In addition, we used both the experiments and databases as prediction methods.

**Gene validation (real-time quantitative PCR (qRT-PCR))**

One altered gene (EEF2K) suggested by the microarray, was validate using qRT-PCR using qRT-PCR. Total RNA of
Peripheral blood samples were isolated using the RNeasy Mini Kit (Qiagen) according to the manufacturer’s protocol. RNA characteristics were determined using a NanoDrop spectrophotometer (Thermo Fisher). We synthesized the complementary DNA using the High Capacity Kit (Applied Biosystems, USA) according to the manufacturer’s instructions. EEF2K (Hs00179434_m1) gene expression level was evaluated using the TaqMan system (Applied Biosystems, Foster City, California, USA). β-Actin was used as a housekeeping gene. The relative gene expression data were analyzed using the 2−ΔΔCT method.

Results
Table 1 shows the clinical characteristics of the pregnant women involved in the study. The initial and final BMI was higher in ND group (p<0.05). Personal history, obstetric and/or familial of diabetes was present in ND group (p<0.05). The evaluation of gene expression profiles showed significant differences between the control group and the ND group (p<0.05), with 22 differentially expressed sequences (7 upregulated and 15 downregulated) (table 2 and figure 1). Of the differentially expressed genes, three genes are of particular interest,
two upregulated (EEF2K and CDR2) and one downregulated (AKAP5), based on the results of the gene networks (figures 2 and 3) and interaction tables (tables 3 and 4) and because of their involvement in key biological processes. The elaborated networks showed 49 interactions with other genes: EEF2K presented interactions with 20 different genes, CDR2 with only 1 gene (figure not shown) and AKAP5 with 28 different genes. The validation of microarray analysis was performed by qRT-PCR. The overexpression of the EEF2K gene in the ND compared with the control group suggested by the microarray was similar to changes in relative gene expression levels measured by qRT-PCR (figure 4).

Conclusion
This study was designed to evaluate the gene expression profile in ND pregnant women (positive screening result and negative diagnosis) compared with control pregnant women with a negative screening result for GDM. The gene expression analysis involving 66,000 genes showed 22 differentially expressed genes in ND pregnant women (7 upregulated and 15 downregulated). These results are important because they show that two populations, which differ only by a positive screening result for GDM, may be distinguished by different gene expression results. Further, the results of this study supports the ‘two-step Carpenter-Coustan’ screening strategy recommended by the American College of Obstetricians and Gynecology.

The literature contains extensive discussion about the best GDM diagnostic strategy able to detect adverse perinatal outcomes. Questions relating to the cost-effectiveness and benefits of GDM detection and treatment are growing in national and international publications and range from publications that deny their importance to those that conclude that screening, diagnosis, and treatment of GDM are cost-effective. The results of the HAPO Study showed that less severe glucose intolerance than is present in GDM is associated with adverse perinatal outcomes and that screening should be universal with use of 75 g of OGGT. The American College of Obstetrician and Gynecology (ACOG, 2013) defines the ‘two-step’ strategy, which includes screening followed by glucose overload for
patients with a positive screening result, because it is cost-effective. The latest publication of the ADA (2015) defines two screening strategies (‘one-step’ and ‘two-step’) as alternatives. Many studies have been published, and others are currently underway, evaluating whether the ‘one-step GDM’ screening strategy advocated by International Association of the Diabetes and Pregnancy Study Groups (IADPSG) is more efficient than the ‘two-step Carpenter-Coustan’ screening strategy. However, after a search of the National Center for Biotechnology Information (NCBI) PubMed Databases, we find no account of a study showing differential gene expression between these populations. For GDM screening, we use fasting glucose >90 mg/dL and risk factors for diabetes. Although different from that described by the ADA, where screening only is considered cost-effective and enough to separate the two populations, our findings provide a guide for a new ‘two-step’ screening that includes tracking, followed by a positive diagnosis.

Bonomo et al and Bevier et al showed that treating pregnant women with only positive screening (altered 50 g OGTT) reduces the occurrence of newborns who are large for their gestational age. The gene expression resulting between pregnant women with negative and positive screening detected in this study suggests a possible relationship between the treatment of altered gene expression and the improvement of adverse perinatal outcome in this population with positive screening for GDM.

Of the 22 differentially expressed genes, we highlight 3: 2 upregulated (EEF2K and CDR2) and 1 downregulated (AKAP5). The interaction networks built with these three genes resulted in the involvement of 49 genes. The EEF2K gene overexpressed in the ND group encodes a highly conserved protein kinase involved in signaling pathways mediated by calmodulin, which activates surface receptors for cell division. This kinase is involved in regulating protein synthesis and phosphorylates the eukaryotic elongation factor 2 (EEF2), inhibiting its function. The activity of this kinase is increased in many cancer types and may be a valid target in cancer treatment. EEF2 plays an essential role in protein synthesis because it catalyzes the translocation of a ribosomal Messenger RNA (mRNA) subunit and two transfer RNAs (tRNAs) after peptide transfer. Phosphorylation of EEF2 to EEF2K blocks translation of the mRNA. It was found that EEF2 knockdowns showed a more pronounced decrease in total insulin content than the decrease in insulin caused by hyperglycemia, suggesting that downregulation of one or more isoforms of this protein plays an important role in the regulation of insulin biosynthesis. Furthermore, long-term attenuation translation may also contribute to glucotoxicity in pancreatic β-cells. In the evaluation of the gene networks, it was found that EEF2K interacts with 20 different genes and is involved in biological processes related to proteins of metabolic processes, cell cycle, and gene expression, insulin receptor signaling pathways, and cellular response to insulin and insulin stimulation. The EEF2K gene and its interactions play an important role in the regulation of insulin biosynthesis. This suggests that this group, even with positive screening and negative diagnosis for diabetes, and this altered gene, may, in future circumstances...
requiring increased insulin synthesis, develop hyperglycemia favoring the emergence of factors related to metabolic syndrome and diabetes.

The CDR2 gene, also upregulated in the ND group, is a tumor antigen expressed on a high percentage of breast tumors and in ovarian cancer. It is also a cell cycle regulatory protein in tumor cells, and its overexpression is responsible for cellular proliferation in tumors. The CDR2 protein is present in 62% of ovarian cancers and is not present in normal tissues. The CDR2 gene has only one gene interaction with MYC (myelocytomatosis oncogene), which has a function related to tumorous processes. Vrachnis et al. showed a possible correlation between diabetes, breast cancer, and pathogenesis of endometrial carcinoma by inflammatory pathways and a possible correlation with ovarian carcinoma. In addition, several studies show a relationship between obesity and breast cancer and ovarian cancer. Women in the ND group, in addition to presenting with overexpression of this gene, also have a history of obesity from before pregnancy. This reinforces the attention that this group requires.

Figure 3 Gene interaction network of the AKAP5 gene. Confidence score of 0.7 (high confidence) and not more than 50 interactions per gene. Prediction methods: experiments (purple lines) and databases (blue lines).

Table 3 Genetic interactions table of the EEF2K gene

| Gene ontology      | Biological processes                                   | Number of genes involved* | p Value     |
|--------------------|--------------------------------------------------------|---------------------------|-------------|
| GO:0019538         | Protein metabolic processes                            | 16                        | 3.2e−6      |
| GO:0007049         | Cell cycle                                             | 15                        | 5.3e−11     |
| GO:0010467         | Gene expression                                         | 15                        | 2.9e−5      |
| GO:0008286         | Insulin receptor signaling pathway                     | 8                         | 2.4e−9      |
| GO:0032889         | Cellular response to insulin stimulation               | 8                         | 2.7e−8      |
| GO:0032868         | Insulin response                                        | 8                         | 1.0e−7      |

Shows the biological processes of interest involved in the network.

p<0.05—Significance of the biological process in the network.

*Number of genes in the interaction which are related to the biological process described.
because, despite having a negative diagnosis for diabetes, these women present positive screening and inflammation due to obesity. This, along with overexpression of the CDR2 gene, may increase the probability of the development of breast cancer and ovarian cancer in women in this group.

AKAP5 is an anchoring kinase A protein (AKAP) family member. AKAPs are a group of structurally diverse proteins that bind the regulatory subunit of protein kinase A (PKA) and confine the holoenzyme at various locations within the cell. PKA is cAMP-dependent and interferes with T cell activation through the expression of the inhibition receptor of the α chain interleukin 2 (IL-2) and IL-2 production. Genetic and immunological studies highlight deficiencies in the IL-2 receptor and its signaling pathway as a central defect in the pathogenesis of type 1 diabetes mellitus. Lack of IL-2 in the pancreas can impair the action of Treg cells and lead to pancreatic β-cell destruction. Prior intervention studies in animal models indicate that the increase in IL-2 signaling can prevent and reverse the disease, especially with the protection conferred by the restoration of Treg cell regulatory function. In obesity, a factor present in this group, Treg cells recruit cytokine-secreting macrophages, which are directly related to insulin resistance. In addition, T cells are directly related to chronic inflammation, and blockage of these cells can improve insulin resistance. This cascade of events is of great interest because downregulation of the AKAP5 gene, which leads to a reduction in Treg cells, may be a compensatory attempt to avoid insulin resistance onset due to obesity and may explain the probable pre-diabetic state present in this group. Furthermore, in the interaction gene networks, AKAP5 was found to interact with 28 genes, several related to immune response, immune system processes, and regulation and secretion of insulin and proteins in general. This finding suggests that this gene may indirectly influence insulin levels and may possibly be related to diabetes progression in this group of pregnant women presenting, at this point, with only risk factors for the disease. The conjunct analysis of these upregulated and downregulated genes suggests that this group of ND women with positive screening is in a moment of balance between insulin resistance and production. The presence of personal obstetric and/or family risk factors, associated or not with impaired fasting glucose, is related to the balanced gene expression profile. However, this is a matter of concern because environmental and personal factors, especially being overweight, can disrupt this balance. These results require confirmation with other studies; however, there is evidence that dietary guidance coming from nutrigenomics can interfere, delay, or even prevent the development of T2DM in this group.

The limitations of this study are related to the small sample size because of the difficulty in finding women with a negative screening result who fall within the inclusion criteria, especially the obesity requirement. Another limitation of this study is the lack of similar studies in the literature for comparison and discussion of the results. However, despite the limitations mentioned, our results show that pregnant women with positive screening for GDM show significant changes in their gene expression profile, displaying 22 differentially expressed genes. Furthermore, our results show that these genes are involved in gene networks and biological processes related to biosynthesis and regulation of insulin secretion and insulin pathways, which are processes implicated in the pathogenesis of diabetes in pregnancy. Despite the small number of differentially expressed genes, our findings genetically separate the two populations of normoglycemic pregnant women differentiated only by a positive screening result. In the group with a positive screening result, there is an apparent regulatory...
balance between the functions of the differentially expressed genes related to the pathogenesis of diabetes (upregulated) and a compensatory attempt to mitigate the possible etiology (downregulated). These results support the ‘two-steps of Carpenter-Coustan’ screening strategy because pregnant women with negative screening do not need to continue onto diagnostic investigation of gestational diabetes, reducing the cost of healthcare and the medicalization of pregnancy.

Our study provides new perspectives for a better understanding of the specific biological processes involved in the pathogenesis of diabetes in pregnancy. Although they are not diabetic, these pregnant women have risk factors. Thus, attention to these genes is important to the timeline of disease evolution and shows that this pregnant women group is a step forward toward diabetes compared with women without these risk factors.

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Contributors
RBG researched the data, wrote, discussed and reviewed/editied the manuscript. DMFS contributed to the discussion and reviewed/editied the manuscript. IMPC and MVCR contributed to the discussion and reviewed/editied the manuscript. FP, DCM and RAAC contributed to the researched data.

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REFERENCES
1. Ayres-de-Campos D, Arulkumaran S. FIGO Intrapartum Fetal Monitoring Expert Consensus Panel. FIGO consensus guidelines on intrapartum fetal monitoring: introduction. Int J Gynaecol Obstet 2015;131:3–4.
2. American Diabetes Association. (2) Classification and diagnosis of diabetes. Diabetes Care 2015;38(Suppl):S8–16.
3. Hod M, Kapur A, Sacks DA, et al. The International Federation of Gynecology and Obstetrics (FIGO) initiative on gestational diabetes mellitus: a pragmatic guide for diagnosis, management, and care. Int J Gynaecol Obstet 2015;131(Suppl 3):S173–211.
4. HAPO Study Cooperative Research Group. Hyperglycemia and Adverse Pregnancy Outcome (HAPO) Study: associations with neonatal anthropometrics. Diabetes 2009;58:453–9.
5. Jensen DM, Damm P, Serensen B, et al. Clinical impact of mild carbohydrate intolerance in pregnancy: a study of 2904 nondiabetic Danish women with risk factors for gestational diabetes mellitus. Am J Obstet Gynecol 2001;185:413–19.
6. Carpenter MW, Coustan DR. Criteria for screening tests for gestational diabetes. Am J Obstet Gynecol 1982;144:768–73.
7. National Diabetes Data Group. Classification and diagnosis of diabetes mellitus and other categories of glucose intolerance. Diabetes 1979;28:1039–57.
8. Rudge MVC. Perfil glicêmico e teste de tolerância oral à glicose no diagnóstico do diabetes gestacional. (Livro-Docência). Botucatu: Faculdade de Medicina, Universidade Estadual Paulista, 1983.
9. Freathy RM, Haynes MG, Urbanek M, et al. HAPO Study Cooperative Research Group. Hyperglycemia and Adverse Pregnancy Outcome (HAPO) study: common genetic variants in GCK and TCF7L2 are associated with fasting and postchallenge glucose levels in pregnancy and with the new consensus definition of gestational diabetes mellitus from the International Association of Diabetes and Pregnancy Study Groups. Diabetes 2012;59:28–92.
10. Trevino V, Falcião F, Barreira-Saldanha HA. DNA microarrays: a powerful genomic tool for biomedical and clinical research. Mol Med 2007;13:527–41.
11. Ingelfinger JR, Jung F, Tang SS. Molecular biology techniques and their applicability to the study of diabetes and hypertension: the renin-angiotensin system as an example. J Am Soc Nephrol 1992;3(4 Suppl):S18–26.
12. Golub TR, Slonim DK, Tamayo P, et al. Molecular classification of cancer: class discovery and class prediction by gene expression monitoring. Science 1999;286:531–7.
13. Bolstad BM, Irizarry RA, Astrand R, et al. A comparison of normalization methods for high density oligonucleotide array data based on variance and bias. Bioinformatics 2003;19:185–93.
14. Pollard KS, Sudhir S, van der Laan MJ. Multiple testing procedures: R multtest package and applications to genomics. In: Gentleman R, Carey V, Huber W, Irizarry R, Dudoit S, eds. Bioinformatics and computational biology solutions using R and Bioconductor, Chapter 15. Springer, New York, NY, 2005:249–71.
15. Huang W, Sherman BT, Lempicki RA. Systematic and integrative analysis of large gene lists using DAVID bioinformatics resources. Nat Protoc 2009;4:44–57.
16. Ayach W, Calderon IMP, Rudge MVC, et al. Associação glicêmica de jejum e fatores de risco como teste para rastreamento do diabetes gestacional. Rev Bras Saúde Matern Infant 2005;5:329–35.
17. American College of Obstetricians and Gynecologists. Reading the medical literature: applying evidence to practice. Washington DC: American College of Obstetricians and Gynecologists, 1998.
18. Metzger BE, Gabbe SG, Persson B, et al. International association of diabetes and pregnancy study groups recommendations on the diagnosis and classification of hyperglycemia in pregnancy. Diabetes Care 2010;33:676–82.
19. Feldman RK, Tieu RS, Yasumura L. Gestational diabetes: U.S. Preventive Services Task Force recommendation statement. Ann Intern Med 2008;148:759–65.
20. Scott DA, Loveman E, McIntyre L, et al. Screening for gestational diabetes: a systematic review and economic evaluation. Health Technol Assess 2002;6:1–161.
21. National Collaborating Centre for Women’s and Children’s Health (UK). Diabetes in pregnancy: management of diabetes and its complications from prepregnancy to the postnatal period. London, UK: RCOG Press, 2008.
22. Metzger BE, Lowe LP, Dyer AR, et al., HAPO Study Cooperative Research Group. Hyperglycemia and adverse pregnancy outcomes. N Engl J Med 2008;358:1901–0202.
23. Bonomo M, Corica D, Mion E, et al. Evaluating the therapeutic approach in pregnancies complicated by borderline glucose intolerance: a randomized clinical trial. Diabet Med 2005;22:1536–41.
25. Bevier WC, Fischer R, Jovanovic L. Treatment of women with an abnormal glucose challenge test (but a normal oral glucose tolerance test) decreases the prevalence of macrosomia. *Am J Perinatol* 1999;16:269–75.

26. Niemann B, Pan R, Teschner M, et al. Age and obesity-associated changes in the expression and activation of components of the AMPK signaling pathway in human right atrial tissue. *Exp Gerontol* 2013;48:55–63.

27. Jørgensen R, Merrill AR, Andersen GR. The life and death of translation elongation factor 2. *Biochem Soc Trans* 2006;34(Pt 1):1–6.

28. Xie CM, Liu XY, Sham KW, et al. Silencing of EEF2K (eukaryotic elongation factor-2 kinase) reveals AMPK-ULK1-dependent autophagy in colon cancer cells. *Autophagy* 2014;10:1495–508.

29. Szklarczyk D, Franceschini A, Wyder S, et al. STRING v10: protein-protein interaction networks, integrated over the tree of life. *Nucleic Acids Res* 2015;43:D447–52.

30. O’Donovan KJ, Diedler J, Couture GC, et al. The onconeural antigen Cdr2 is a novel APC/C target that acts in mitosis to regulate c-myc target genes in mammalian tumor cells. *PLoS ONE* 2010;5:e10045.

31. Balamurugan K, Luu VD, Kaufmann MR, et al. Onconeural cerebellar degeneration-related antigen, Cdr2, is strongly expressed in papillary renal cell carcinoma and leads to attenuated hypoxic response. *Oncogene* 2009;28:3274–85.

32. Vrachnis N, Iavazzo C, Iliodromiti Z, et al. Diabetes mellitus and gynecologic cancer: molecular mechanisms, epidemiological, clinical and prognostic perspectives. *Arch Gynecol Obstet* 2016;293:239–46.

33. Huang Z, Hankinson SE, Colditz GA, et al. Dual effects of weight and weight gain on breast cancer risk. *JAMA* 1997;278:1407–11.

34. Wolk A, Gridley G, Svensson M, et al. A prospective study of obesity and cancer risk (Sweden). *Cancer Causes Control* 2001;12:13–21.

35. O’Flanagan CH, Bowers LW, Hursting SD. A weighty problem: metabolic perturbations and the obesity-cancer link. *Horm Mol Biol Clin Investig* 2015;23:47–57.

36. AKAP5 gene. http://www.genecards.org/cgi-bin/carddisp.pl?gene=AKAP5 (accessed 5 Nov 2015).

37. Ramstad C, Sundvold V, Johansen HK, et al. cAMP-dependent protein kinase (PKA) inhibits T cell activation by phosphorylating ser-43 of raf-1 in the MAPK/ERK pathway. *Cell Signal* 2000;12:557–63.

38. Hartemann A, Bourron O. Interleukin-2 and type 1 diabetes: new therapeutical perspectives. *Diabetes Metab* 2012;38:387–91.

39. Hulme MA, Wasserfall CH, Atkinson MA, et al. Central role for interleukin-2 in type 1 diabetes. *Diabetes* 2012;61:14–22.

40. Lumeng CN, Maillard I, Saltiel AR. T-ing up inflammation in fat. *Nat Med* 2009;15:846–7.