Dynamic modeling of folliculogenesis signaling pathways in the presence of miRNAs expression

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Abstract

Background: TEK signaling plays a very important role in folliculogenesis. It activates Ras/ERK/MYC, PI3K/AKT/mTORC1 and ovarian steroidogenesis activation pathways. These are the main pathways for cell growth, differentiation, migration, adhesion, proliferation, survival and protein synthesis.

Results: TEK signaling on each of the two important pathways where levels of pERK, pMYC, pAkt, pMCL1 and pEIF4EBP1 are increased in dominant follicles and pMYC is decreased in dominant follicles. Over activation of ERK and MYC which are the main cell growth and proliferation and over activation of Akt, MCl1, mTORC1 and EIF4EBP1 which are the main cell survival and protein synthesis factors act as promoting factors for folliculogenesis. In case of over expression of hsa-miR-30d-3p and hsa-miR-451a, MYC activity level is considerably increased in subordinate follicles. Our simulation results show that in the presence of has-miR-548v and bta-miR-22-3p, downstream factors of pathways are inhibited.

Conclusions: Our work offers insight into the design of natural biological procedures and makes predictions that can guide further experimental studies on folliculogenesis pathways. Moreover, it defines a simple signal processing unit that may be useful for engineering synthetic biology and genes circuits to carry out cell-based computation.

Keywords: Dynamic modeling, Folliculogenesis, TEK signaling, Ras/ERK/MYC, PI3K/AKT/mTORC1

Background

In the previous study [1], transcriptome profiling of granulosa cells from small (< 5 mm) and large (> 10 mm) follicles using high-throughput data sets were analyzed and “Integrated miRNA–mRNA Bipartite Network” was reconstructed. According to the next analysis, the most important signaling pathways involved in folliculogenesis were detected. These pathways were included Ras/ERK/MYC, PI3K/AKT/mTORC1 and ovarian steroidogenesis. Based on these findings, Ras/ERK/MYC and PI3K/AKT/mTORC1 pathways represent regulatory networks and fundamental signaling transduction for the majority of cellular physiological processes in folliculogenesis, such as growth, differentiation, migration, adhesion, proliferation, survival and protein synthesis. These pathways are mostly activated by alterations in ANGPT1, TEK, MYC, MAPK1, PIK3R1, MCL1 and EIF4EBP1 genes [1]. Activation of such pathways is responsible of uncontrolled cellular physiological processes and can contribute to drug discovery. Combination therapies with pharmacological inhibitors or activators of these follicles pathways may have potential uses for the control of folliculogenesis process. Comparison of between dominant and subordinate follicles is a good model to investigate the activation of Ras/ERK/MYC and the PI3K/AKT/mTORC1 pathways as it is frequently affected by miRNAs that causes the activation or inhibition of pathways [1]. Folliculogenesis is thought to occur by sequential accumulation of molecular alterations and genetic [2]. Although, the mechanisms underlying follicle development are still unknown, several genes and metabolic pathways have been shown to carry molecular alterations in folliculogenesis. Folliculogenesis exhibit mutations in the Ras/RAF/mitogen activated protein kinase (MAPK) pathway. This mutation is known to play a key role in cell growth and proliferation of follicles,
through activation of the ERK pathway [1]. In particular, it occurs within the activation segment of the kinase domain and it results in an increased activity of the kinase itself. Constitutive activation of the kinase activity leads to unresponsiveness of negative feedback mechanisms within the ERK pathway [3]. Previous studies show that ERK and AKT pathways are activated in parallel and the evidence that PI3K/AKT and MAPK/ERK1/2 cascades are interconnected is described [4–6]. According to these findings, ERK and AKT pathways could represent targets for an otherwise devastating trait. Computational or dynamical modeling and Computer simulations are useful to analyze and to increase understanding of metabolic pathways [7], specially involved in folliculogenesis and their complex interactions. On the other hand, MYC activity must be controlled in response to different environmental cues. Previous studies have showed that MYC is regulated at multiple levels, including post-transcriptional regulation [8, 9] and autoregulation of MYC transcription [10]. More recent discoveries indicate that MYC is also regulated at the protein level by the Ras effector pathways [11–14]. The control of MYC dynamics by successive phosphorylation allows MYC to merge upstream signals from ERK and PI3K, which play key roles in controlling diverse cell fates [15, 16]. As well as, major advances in understanding cell survival and translation control have come from the discovery of the intimate relationship between growth factor, the mitogen and hormone-responsive phosphatidylinositol 3-kinase (PI3K) pathway and the energy-sensing target of rapamycin (TOR) pathway. The PI3K and TOR pathways converge on common downstream targets: the ribosomal S6 kinases (S6ks) and the translation initiation factor 4E-binding proteins (4E-BPs). Since the discovery that the TOR inhibitor rapamycin is effective in treatment of cell caused by misregulation of players in the PI3K pathway, the importance of the interplay between these two pathways is gaining greater recognition [17]. In this work we survey here recent advances in our understanding of how the PI3K pathway impinges on the PI3K/AKT/mTORC1 pathway, how levels of expression in the PI3K/R1 and EIF4EBP1 pathways affect cell survival and translation control, and how the downstream targets of mTORC1 regulate cell survival. We develop a dynamic model that simulates both Ras/ERK/MYC and PI3K/AKT pathways, cell growth, and its environments, such as ordinary differential equations (ODEs), Petri nets, Boolean network, agent-based model, the system biology modeling approach considering genetic variation, and linear programming (LP) based model [18]. Of these, because of modeling base on ODEs is able to predict precisely and continuously manner of each node in the network during the time, so in our work, it was considered. Overall, this model may be used for an in silico lab to study the effects of potential activators or inhibitors that may improve the response to standard treatments.

**Methods**

**Model description**

The related network for subordinate, dominant and dominant follicles in the presence of miRNAs were generated according to previous models and experimental observations that dealt with published signaling network dynamic behavior [19–38] (Additional file 1: Tables S1–S3). The network consists of two main pathways (Fig. 1) that play an important role in cell growth, proliferation, survival and protein synthesis. These pathways are activated by ANGPT1 and ANGPT2 attachment to TEK. After the interaction of ligand (ANGPT1 and ANGPT2 that these species were antagonist with each other) with the extracellular domain of TEK, the receptors undergo homodimerization that causes auto phosphorylation of certain tyrosine residues in the cytoplasmic end. Growth factor receptor-bound protein 2 (Grb2) binds phosphorylated tyrosines (pY1148, pY1086) and Src homology 2 domain containing transforming protein (Shc) binds pY1148 and pY1173 [39]. After C-terminal tail phosphorylation, the Shc adaptor binds its site and Grb2 attaches to Shc, and then SOS (GTP exchange protein for Ras) is employed by Grb2. In the next step SOS converts Ras-GDP into Ras-GTP which is the activated form of Ras [40, 41]. Activated Ras causes Raf phosphorylation and activation [42, 43]. There are three kinds of Raf in cells: Raf-1(C-Raf), A-Raf and B-Raf [44, 45]. Raf phosphorylates and activates the MEK (MAPK kinase). The activated MEK, phosphorylates and activates extracellular signal-regulated kinase 1 (ERK1) and ERK2. ERK1 and ERK2 phosphorylate MYC protein that leads to cell growth, proliferation and survival [1]. Two miRNAs are included hsa-miR-30d-3p and hsa-miR-451a suppress translation of MYC gene and result low concentration of MYC protein in dominant follicles. Phosphatidylinositol-3-kinase (PI3K) has two 85 and 110 KD subunits which are regulatory and catalytic subunits, respectively [46, 47]. After RTK phosphorylation, the regulatory subunit binds its phosphorylated Thyrosine site and then the catalytic and the regulatory subunits join together. Insulin receptor substrate-1(IRS1) phosphorylated. Under such conditions the, PI3K and IRS1 are activated and converts the membrane phosphatidylinositol 4,5biphosphate (PIP2) into phosphatidylinositol 3,4,5-3-phosphate (PIP3).
PIP3 makes Akt activation in a way that PDK1 binds membrane PIP3, then PDK1, phosphorylates and activates Akt. Akt is activated and stimulates several factors such as phosphorylation of CREB and mTORC1. The CREB phosphorylates and activates Mcl1. On the other hand, mTORC1 phosphorylates and inactivates EIF4EBP1. So, Akt indirectly or directly regulates cell growth, survival and translation control through phosphorylating its substrates. Akt deactivation and the PI3K negative regulation are done by PTEN. PTEN is a phosphatase that removes phosphates group from the phosphatidylinositol 3, 4,5-3phosphate that results Akt inactivation [48]. Based on literature mining, TEK signaling network was modified for subordinate, dominant follicles and dominant follicles in the presence of miRNAs and reconstructed network were designed for the folliculogenesis state which is accessible in Additional file 1.

**Dynamical model**

Systems biology involves network perspective of biological systems. The domain of systems biology could be from an individual reaction cycle of a cell, tissue, organ or even the whole organism. Computational modeling has an axial role in converting biological events to quantitative data for analyzing. For better understanding, such models should reflect the systems as they exist because by only this way, reliable predictions are practical about them. One of the most usually practical methods in order to model biological systems is Ordinary Differential Equations (ODEs). A differential equation is explained as an equation showing the relationship between a function such as $[S]$, i.e. the concentration of S, alters over the time. This is done by inter relating the rate of change with the concentration at the moment $[49–52]$. Our suggested subordinate follicles (SF) and dominant follicles (DF) models are based on ODEs and involve 105 species, 112 reactions, 170 parameters and 1 rule. Our suggested dominant follicles in the presence of miRNAs (DF + miRNAs) model is based on ODEs and involves 110 species, 116 reactions, 178 parameters and 1 rule. In Additional file 1: Table S1, reactions are shown in SF and DF. The Additional file 1: Table S1 also show the initial values for non-zero species in SF, DF and DF + miRNAs respectively. On the other hand, in Additional file 1: Table S3, reactions are
presented for DF + miRNAs model. The systems biology markup language (SBML) of our models is also presented in Additional file 2: SBML S1 (SF model), Additional file 3: SBML S2 (DF model) and Additional file 4: SBML S3 (DF + miRNAs model). SBML is a computer-readable format like XML for indicating models of biochemical reaction networks. For simulations, ODE15s routine from MATLAB 8.5 was used to solve ODEs. Here, a sample derivation regarding one of the ODEs (related to ANGPT1 and ANGPT2 and RTK binding) is reported; the reaction is considered as a second order one.

\[
ANGPT1 + TEK < \frac{K1}{Kr1} > ANGPT1 - TEK
\]

The reaction rate by which ANGPT1-RTK is produced is:

\[
v = K1[ANGPT1][TEK] - Kr1[ANGPT1 - TEK]
\]

Where K1 is the rate constant for the forward direction and Kr1 is for the reverse one.

**Results**

**ANGPT1 and TEK over-expression in DF in the comparison of SF**

Figure 2a compared TEK concentration between SF (blue line), DF (black line) and DF + miRNAs (red line). In this figure TEK concentration in presence of miRNAs during the time was higher than SF and DF as well as, in DF higher than SF. On the other hand, Fig. 2b compared phosphorylated and activated ANGPT1-TEK is lagging relative to the SF in comparison of DF and DF + miRNAs but its concentration decreases with a mild slope. ANGPT1 was first detected as a major activator of TEK, resulting in a downstream activation of the phosphatidylinositol 3-kinase/Akt survival pathway, thereby promoting cell survival.

**Comparison of Ras/ERK/MYC pathways between SF, DF and DF + miRNAs model**

Our simulation results denoted that up-regulation of TEK, enhanced TEK and Ras expression in dominant follicles had a deep impact on Ras-ERK. Figure 4 compares important species kinetics in Ras/ERK/MYC between SF, DF and DF + miRNAs model and shows that Ras/ERK/MYC pathway activation kinetics is vastly different between the SF, DF and DF + miRNAs. Figure 3a compared Ras-GTP formation kinetics. Ras-GTP formation peak in SF is lagging relative to the DF and DF + miRNAs, but its concentration decreases with a mild slope. Of course lines relate to DF and DF + miRNAs were overlapping approximately and there isn’t different between them. Raf1 activation kinetics is also like to Ras-GTP formation (Fig. 3b). MEK downstream factor kinetics and ERK between the SF, DF and DF + miRNAs model are completely different. Figure 3c represents the MEK phosphorylation peak in DF and DF + miRNAs (red line overlapped black line) in about 280 s while in the SF model it is about 340 s. Phosphorylated MEK (ppMEK) in DF and DF + miRNAs are much higher than the SF model showing that in DF and DF + miRNAs MEK activity is higher and the active state period is longer for MEK. Figure 3d illustrates that phosphorylation or activity level of ERK (ppERK) in SF and DF + miRNAs (red and black lines) is not comparable to the SF in a way that conversely, SF model ppERK concentration is few and the simulation predicts that DF and DF + miRNAs ERK phosphorylation peak is nearly 650 s and for SF is about 950 s. As well as, the time period and the concentration of phosphorylated ERK in DF and DF + miRNAs is much more that the SF model. Figure 3e represent different concentration of MYC between SF, DF

![Fig. 2](image-url)  Computational simulation of TEK and ANGPT1-TEK autophosphorylation and internalization. a Kinetics of TEK concentration in different models in 70s. b Kinetics of ANGPT1-TEK phosphorylation and internalization in 70 s (blue, black and red line representing concentration in the subordinate follicles, dominant follicles and dominant follicles in the presence of miRNAs, respectively)
and DF + miRNAs. Concentration of MYC in subordinate follicles significantly is higher than both dominant follicles and dominant follicles in the presence of two miRNAs (hsa-miR-30d-3p and hsa-miR-451a) which both of them were up-regulated during the entire period. Also, the comparison between DF and DF + miRNAs model showed that concentration level of DF was higher than DF + miRNAs. Figure 3f show that phosphorylation or activity level of MYC (pMYC) in SF (blue line), DF (black line) and DF + miRNAs (red line) is considerable, SF model pMYC concentration is higher than DF and DF + miRNAs model and the simulation predicts that SF MYC phosphorylation peak is nearly 850 s and for DF and DF + miRNAs is about 700 s.

Comparison of PI3K/AKT/mTORC1 pathways between SF, DF and DF + miRNAs model

Figure 4a, b and c compares IRS1, PI3K and Akt activation kinetics between SF, DF and the DF + miRNAs model. In Fig. 5a, b and c, it is showed that IRS1 (pIRS1), PI3K and Akt (pAkt) activity level in DF and DF + miRNAs (black and red line which lines were
overlapping) are higher than the SF model. These models show that over-expression of ANGPT1 and TEK enhanced levels of these species, so high expression levels have a considerable effect on PI3K/Akt/mTORC1 pathway. Figure 5a and b show that formation of cytoplasmic CREB (pCREB) and nucleic McI1 (pMcI1) in DF and DF + miRNAs significantly have different kinetics compared to the SF model and the concentration level of pCREB and pMcI1 is much higher than the subordinate follicles. As well as, Fig. 6a and b compares mTORC1 and EIF4EBP1 activation kinetics between SF, DF and the DF + miRNAs model. pmTORC1 and pEI-F4EBP1 formation peak in SF are lagging relative to the DF and DF + miRNAs. Of course lines relate to DF and

Fig. 4 IRS1, PI3K and Akt species kinetics comparison in of PI3K/Akt/mTORC1 pathway between SF, DF and DF + miRNAs models. a Kinetics of IRS1 formation. b Kinetics of PI3K activation. c Phosphorylation kinetics of Akt leading to pAkt double phosphorylation: SF model factor (blue); DF model factors (black); DF + miRNAs model factors (red)

Fig. 5 CREB and McI1 species kinetics comparison in of PI3K/Akt/mTORC1 pathway between SF, DF and DF + miRNAs models. a Kinetics of CREB formation. b Kinetics of McI1 activation: SF model factor (blue); DF model factors (black); DF + miRNAs model factors (red)
DF + miRNAs were overlapping approximately and there isn’t different between them.

**Discussion**

In the previous work, Bahrami et al. [1] analyzed the global transcriptome in both dominant follicles and subordinate follicles. Taking advantage of hierarchical clustering, they found the global gene expression is greatly altered, with 233 genes significantly upregulated and 44 genes significantly downregulated. In the KEGG pathway analysis of altered genes, they also found that Ras/ERK/MYC and PI3K/AKT/mTORC1 signaling pathways were highly enriched in deregulated genes. As well as ANGPT2 gene was downregulated. Besides the protein encoded by this gene is an antagonist of angiopoietin 1 (ANGPT1) and endothelial TEK tyrosine kinase [1]. Of the four identified angiopoietins (ANGPT1 to ANGPT4), the best characterized are ANGPT1 and ANGPT2. ANGPT1 and ANGPT2 have both been identified as ligands for TEK, a receptor expressed on endothelial cells, and they play important roles in angiogenesis, in concert [53–58]. There are a few studies about over-expression of TEK receptor in folliculogenesis process. In general, dynamics modeling of biological process in the reproduction events is rare and this study in the first report around computational modeling of ovary changes and biological procedures. In the previous study we reported over-expression of TEK receptor in dominant follicles. On the other hand, two miRNAs include has-miR-548v and bta-miR-22-3p which suppress RTK or TEK receptor, were down-regulated in dominant follicles. As well as, expression level of ANGPT1 and ANGPT2 were up-regulated and down-regulated, respectively [1]. ANGPT1 binding to TEK maintains and stabilizes mature vessels by promoting interactions between endothelial cells and surrounding extra-cellular matrix. However, ANGPT2 shows context-dependent, antiangiogenic and proangiogenic activities. ANGPT2 was first identified as a natural antagonist for TEK that disrupts in vivo angiogenesis. Studies of ANGPT2 knockout animals reported that ANGPT2 is not need to embryonic vascular development, but is needed for postnatal, angiogenic remodeling [59]. ANGPT2 is only up-regulated at sites of active vascular remodeling, which involves vessel regression and destabilization [56]. There are many studies on ERK roles in biological process especially in cancer study. Activated ERK has been elicited Form 91% out of 101 head and neck cancer samples, 84% of 60 prostate cancer samples, 72% out of 90 breast cancer samples and 67% out of 74 gastric cancer samples respectively [60, 61]. Bahrami et al. [1] have revealed that ERK activation is related to RTK signaling in folliculogenesis. The MYC protein is a nuclear phosphoprotein that plays a role in cellular transformation, cell cycle progression, apoptosis, cell growth and proliferation [62]. Because of subordinate follicles critical need to cell growth and proliferation thus MYC protein concentration was high in these follicles. Bahrami et al. [1] disclosed the activity of PI3K/Akt/mTORC1 pathway in folliculogenesis. Mcl1 protein that relates to cell survival and cell adhesion has high concentration in the DF, because of these follicles need to survival for the ovulating process. On the other hand, this protein is an anti-apoptotic protein, which is a member of the Bcl-2 family, so involved in the regulation of apoptosis versus cell survival, and in the maintenance of viability but not of proliferation. EIF4EBP1 protein which relate to PI3K/AKT/mTORC1 signaling pathway and protein synthesis has high concentration in the DF. This protein is one member of a family of translation repressor proteins. The protein directly interacts with eukaryotic translation initiation factor 4E, which is a limiting component of the multisubunit complex that recruits 40S ribosomal subunits to the 5’ end of mRNAs. Interaction of this protein with eukaryotic translation initiation factor 4E

![Fig. 6 mTORC1 and EIFEBP1 species kinetics comparison in of PI3K/Akt/mTORC1 pathway between SF, DF and DF + miRNAs models. a Kinetics of pmTORC1 formation. b Kinetics of EIFEBP1 activation: SF model factor (blue); DF model factors (black); DF + miRNAs model factors (red)](image-url)
inhibits complex assembly and represses translation. Since DF don’t need to additional growth, protein synthesis via high concentration of this protein was inhibited.

**Conclusion**

In this study, for the first time we have made mathematical and dynamical models representing TEK signaling in subordinate and dominant follicles as well as dominant follicles in the presence of miRNAs and explore the different behaviors and dynamics of these models. For the first time we analyze the TEK signaling pathway specially involved in folliculogenesis. As well as, for the first time we have done dynamic modeling in the presence of differential expression of miRNAs. Since, TEK signaling is crucial for cell growth, survival, proliferation and translation control (protein synthesis); it might be the main reason for biological progression in folliculogenesis. TEK signaling activates Ras/ERK/MYC and PI3K/AKT/mTORC1 activation pathways. These two pathways are the main routes for cell growth, survival and proliferation [1]. Therefore over or down expression that lead to excessive activation of these pathways may cause folliculogenesis. This differential expression causes kinetic changes in downstream factors in the above pathways. This study has denoted how computational and dynamical models can be used to explore different trait states and propose how drug treatment could be improved to better combat the effects of the trait. We believe that our model is a good delegation of the PI3K/AKT/mTORC1 and Ras/ERK/MYC pathways. And we need to more investigation about dynamic behaviors involved in folliculogenesis in the future.

**Additional files**

- Additional file 1: Tables S1 - S3. (DOCX 125 kb)
- Additional file 2: SBML S1 (SF model). (XML 158 kb)
- Additional file 3: SBML S2 (DF model). (XML 158 kb)
- Additional file 4: SBML S3 (DF+miRNAs model). (XML 163 kb)

**Abbreviations**

ANGPT1: Angiopoietin-1; ANGPT2: Angiopoietin-2; DF: Dominant follicle; EIF4EBP1: Eukaryotic Translation Initiation Factor 4E Binding Protein 1; MAPK: Mitogen-Activated Protein Kinase; ODE: Ordinary differential equation; PI3K: Phosphoinositide 3-kinase; PIP2: Phosphatidyl inositol 4,5bisphosphate; SF: Subordinate follicle

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**Availability of data and materials**

The authors declare that data supporting the findings of this study are available within the article and its Additional files.

**Authors’ contributions**

Conceived and designed the experiments: AB and AN; Analyzed the data: AB, AN, SRMA, MS and RR; Contributed reagents/materials/analysis tools: AB; Wrote the paper: AB and AN; All authors reviewed the results and approved the final version of the manuscript.

**Ethics approval and consent to participate**

Not applicable.

**Consent for publication**

Not applicable.

**Competing interests**

The authors declare that they have no competing interests.

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