Supplementary Materials

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Supplementary Materials for

Merging Systems Biology with Pharmacodynamics

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Merging systems biology with pharmacodynamics

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MODEL DESCRIPTION

An operational ePD model for an epidermal growth factor receptor (EGFR) inhibitor was developed based on published mathematical models of the EGFR signaling pathway and biochemical knowledge of the pathway reported in literature. We used the Brightman and Fell model¹ as a framework model for the EGFR stimulation of MAPK/ERK signaling pathway. We simplified the model by consolidating some of the similar reactions and/or by applying Michaelis-Menten kinetics. The modified MAPK/ERK model was extended to incorporate the following key signaling modules: the PLCγ-mediated activation of PKC and its inhibition of RKIP, which limits the catalytic activity of the Raf-1 protein; the downstream ERK-mediated expression of target genes, specifically *CCND1* gene (cyclin D); $p27^{kip}$ -mediated inhibition of Cyclin D-CDK4/6 activity- (called active Cyclin D); and mir-221-mediated reduction of $p27^{kip}$ levels. To model drug activity, we assumed that the oncogene was the mutated activated EGFR and that the EGFR inhibitor was a specific tyrosine kinase inhibitor of EGFR that bound directly to EGFR that resulted in a constant fractional inhibition of 80%. To assess tumor size changes (measured in mg) to EGFR inhibition, the amplitude and duration of active Cyclin D was used as an indicator of the rate of cell proliferation measured as the normalized mass doubling time to the growth fraction of the tumor. The key components of the overall ePD model are depicted in Figure 1 in the main text.

To construct the model, we assumed a well-mixed cellular compartment with a cell volume of approximately 1.0×10^{-12} L. Therefore, the resulting model is a system of 34 non-linear ordinary differential equations (ODEs) with 69 kinetic parameters arising from chemical reactions represented mostly by the law of mass action and Michaelis-Menten kinetics. The complexity of the EGF-driven *CCND1* gene expression was approximated by using the Hill function. Many of the initial concentrations and rate constants used in the model were obtained from the previous models including the Brightman and Fell model¹⁻⁷. For the remaining unknown parameters for our proposed molecular reactions, we estimated them by comparing to similar molecular events in EGFR or other signaling pathway models and by choosing arbitrarily parameter values that gave qualitatively similar dynamic responses to those observed from other models⁸. The complete set of the model equations was integrated using the *ode15s* routine of MATLAB R2011b (7.13.0.564) (The MathWorks, Inc.). Complete set of model equations and parameters are shown in Table S1 and2.

The cell signaling model provided with values of active Cyclin D concentration that we used as marker for entry into the cell cycle as suggested by Stacey⁹. We hypothesize that the total area under the active Cyclin D is related to the mass doubling time given the growth fraction by a constant factor, c as 4.8 x 10^{-6} , which we empirically determined through hypothesizing that the "normal" variation has a rate of 0.1. We then estimated tumor size (mg) from mass doubling time given the growth fraction of the tumor according to Table 3-2 found in Ref #10¹⁰ using the Hill equation with V_{max} of 6.98 x 10³, K = 0.4884, and n of 1.1802. The resulting tumor sizes for the various conditions are shown in Figure 2 of the main text. This procedure is described as equation 38-39 in Table S3. Various simulation conditions for multiple genomic and epigenetic alterations are specified in Table S4.

Figure S1: Temporal Profiles of Active Signaling Components of the EGFR pathway in the Standard Patient (SP) Levels of activated (*) Ras, Raf and Mek and Cyclin D levels in the standard patient receiving EGFR antagonist therapy

Figure S2: Temporal Profiles of Active Signaling Components of the EGFR pathway in Patient B

Levels of activated (*) Ras, Raf and Mek and Cyclin D levels in Patient B who has hypermethylated *RASAL1* gene resulting in lower levels of RasGAP and a SNP in *RKIP/PEBP1* gene that renders it unresponsive to PKC regulation and is receiving EGFR antagonist therapy

Figure S3 Temporal Profiles of Active Signaling Components of the EGFR pathway in Patient A

Levels of activated (*) Ras, Raf and Mek and Cyclin D levels in Patient A who has hypermethylated *RASAL1* gene that decreases the levels of RasGAP and is receiving EGFR antagonist therapy

Figure S4

Figure S4 Temporal Profiles of Active Signaling Components of the EGFR pathway in Patient C.

Levels of activated (*) Ras, Raf and Mek and Cyclin D levels in Patient C who has a SNP in *RKIP/PEBP1* gene that renders it unresponsive to PKC regulation and decreased levels of miR-221resulting in increased levels of the cell cycle inhibitor p27kip and is receiving EGFR antagonist therapy.

Table S1. **CHARACTERISTICS OF DIFFERENT TYPES OF COMPUTATIONAL MODELS.**

- **Toy Models**: Arbitrary parameter values are used for both reactant concentrations and reaction rates. Generally toy models are used to make a theoretical point.
- **Plausible Models**: Parameter values that are experimentally measured or estimated from experiments are used to construct these models. Models are generally not specific for any cell type and use parameter values obtained from different cell types and species. These are the most common type of models in Systems Biology.
- **Operational Models**: These models use a combination of experimentally measured or estimated parameters as well as arbitrary parameters to develop a system wherein we can explicitly connect genomic/epigenomic and posttranslational changes to model parameters. For example in the model described here, we have arbitrarily assumed that a two fold increase in promoter region DNA methylation results in a 2 fold decrease in protein levels. This assumption (toy parameter) will have to be experimentally verified in the specific system of interest.
- **Identifiable Models:** These are models built to explain experimental observations. They are system-specific and fully constrained by or fitted to experimental data. These models are the type most commonly used in current PK/PD studies. Often the molecular basis for the experimentally identified parameters is not understood in current PD models.

Table S2. EQUATIONS FOR THE ODE MODEL

Table S3. ODE MODEL PARAMETERS. First-order (*kb* and *kc*) and second-order (*ka*) rate constants are given in $[\text{min}^{-1}]$ and $[\text{molecule}^{-1} \cdot \text{min}^{-1}]$, respectively, except for $ka(1)$ given in $[\text{nM}^{-1} \cdot \text{min}^{-1}]$ and dimensionless $kc(31)$. *Vm* and *Km* are expressed in [molecule⋅min⁻¹], and [molecule], respectively.

Table S4. RELATIONSHIP BETWEEN FREE CYCLIN D, MASS DOUBLING GIVEN GROWTH FRACTION, AND TUMOR SIZE.

Table S5. SIMULATION CONDITIONS FOR MULTIPLE GENOMIC/EPIGENOMIC ALTERATIONS IN THE EGFR NETWORK.

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