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The NF-κB multidimer system model: A knowledge base to explore diverse biological contexts

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Abstract

The nuclear factor κB (NF-κB) system is critical for various biological functions in numerous cell types, including the inflammatory response, cell proliferation, survival, and differentiation, and pathogenic responses. Each cell type is characterized by a subset of 15 NF-κB dimers whose activity is regulated in a stimulus-responsive manner. Numerous studies have produced different mathematical models that account for cell type–specific NF-κB activities. However, whereas the concentrations or abundances of NF-κB subunits may differ between cell types, the biochemical interactions that constitute the NF-κB signaling system do not. Here, we synthesized a consensus mathematical model of the NF-κB multidimer system, which could account for the cell type–specific repertoires of NF-κB dimers and their cell type–specific activation and crosstalk. Our review demonstrates that these distinct cell type–specific properties of NF-κB signaling can be explained largely as emergent effects of the cell type–specific expression of NF-κB monomers. The consensus systems model represents a knowledge base that may be used to gain insights into the control and function of NF-κB in diverse physiological and pathological scenarios and that
describes a path for generating similar regulatory knowledge bases for other pleiotropic signaling systems.

**Introduction**

A key goal of systems biology is to represent molecular mechanistic knowledge that is informed by a plethora of experimental observations in mathematical models so that it can be used for predictive simulations. Indeed, constructing mathematical models involves summarizing accumulated knowledge into wiring diagrams and encoding these in equations whose constants are specified by biophysical, biochemical, or cell biological measurements. As such, models may function as knowledge bases of molecular mechanisms. Common model repositories (for example, Biomodels (1)) are fueled by the vision that such mathematically encoded knowledge bases will drive quantitative systems biology studies broadly. Indeed, models of metabolic networks in bacteria, yeast, or humans have enhanced research of microbial growth and metabolic control (2, 3). Yet, in the fields of signaling, gene expression, and regulatory biology, the adoption and application of models produced by one laboratory by other researchers has been disappointingly rare.

The reliability of a model as a knowledge base is evaluated by its ability to recapitulate experimental observations and make accurate predictions that can be experimentally confirmed. However, models are necessarily an abstraction of the true molecular regulatory network. One consequence is that an abstracted model built to explore a functional characteristic in one cell type may not capture the functional characteristics observed in another cell type. To capture diverse biological phenomena in different cell types, we suggest that the mathematical model must be
sufficiently detailed that the genetically encoded biochemical mechanisms are represented explicitly, such that nongenetic, cell-type differences may be accounted for by altering merely expression or synthesis (not interaction or catalytic) rate constants based on readily accessible mRNA or protein abundance measurements.

The NF-κB transcription factor family plays a critical role in the regulation of inflammation, immunity, cell development, cell survival, and proliferation. The dynamic control of the ubiquitous NF-κB family member RelA has been the subject of numerous mathematical models, and these have been reviewed previously (4, 5). However, much NF-κB-associated biology and pathology is mediated by other NF-κB family members, such as RelB, cRel, p50, and p52, which, together with RelA, may potentially form 15 distinct dimers whose activities are controlled by two stimulus-responsive signaling pathways (6, 7). Several studies involving models of the multi-dimer NF-κB system have explored diverse biological scenarios, ranging from B cells (8-10) to dendritic cells (DCs) (11) and fibroblasts (11-13). It remains unclear whether the knowledge base of molecular mechanisms encoded by these models of the multi-dimer NF-κB system may be represented by a single consensus model that may then be applied to diverse physiological and pathological contexts.

Here, we reviewed the accumulated molecular mechanistic knowledge contained in published mathematical models of the NF-κB multi-dimer system to produce a consensus model. Our analysis and synthesis were guided by the following principles: (i) that biochemical interactions mediated by protein surfaces are encoded by the genome and hence are conserved between cell types, and (ii) that the abundances of NF-κB monomers are a function of the epigenome and hence
are cell type–specific. We thereby showed that the NF-κB multi-dimer system consensus model is a knowledge base that may be used to explore NF-κB control and function in diverse physiological and pathological contexts.

**Components and topology of the NF-κB signaling network in different cell types**

**Canonical and noncanonical NF-κB signaling**

Two NF-κB activation pathways have been described: the canonical, NEMO-dependent pathway and the noncanonical, NF-κB-inducing kinase (NIK)–dependent pathway (14) (Fig. 1). Generally, pathogen and inflammatory signals activate the canonical pathway, whereas developmental signals activate the noncanonical pathway (15). The canonical activation of NF-κB in response to immune threats plays a critical role in acute and reversible immune and inflammatory responses, whereas noncanonical NF-κB activation by developmental signals may promote long-lasting or irreversible cell survival, cell differentiation, or organogenesis. The canonical NF-κB pathway can reach its peak response to inflammatory stimuli within 30 min, whereas the noncanonical pathway responding to pro-survival stimuli can remain active beyond 12 hours (15-17).

The molecular mechanism of canonical pathway activation by Toll-like receptors (TLRs), tumor necrosis factor receptors (TNFRs), B cell receptors (BCRs) and T cell receptors (TCRs), involves activation of the NF-κB essential modulator (NEMO)-IKK complex, which phosphorylates inhibitory NF-κB proteins (IκBs) and targets them for proteasomal degradation (18). Upon IκB degradation, the NF-κB dimers that were associated with them, such as RelA:p50, RelA:RelA, cRel:p50, RelA:cRel, and RelB:p50, are released and translocate to the nucleus to activate target gene transcription.
Activation of the noncanonical NF-κB pathway occurs after the ligation of receptors such as LTβR, CD40, BAFFR, RANK, TNFR2, or CD27. Upon receptor ligation, proteasomal degradation of the E3 ubiquitin ligase TRAF3 results in the accumulation of NIK (19), and hence activation of the IKK1-containing, NEMO-independent IKK complex. Once activated, IKK1 phosphorylates p100 at specific C-terminal serine residues, promoting ubiquitylation and proteasomal degradation of its C-terminal inhibitory domain (20). This enables two activities to result from noncanonical NF-κB activation. First, degradation of the C-terminal inhibitory domain enables the release of trapped NF-κB dimers, such as RelA:p50 and RelB:p50. Second, p100 processing leads to the generation of p52, enabling the generation and nuclear translocation of the RelB:p52 dimer (15).

**Cell type–specific NF-κB dimer repertoires**

Although the canonical and noncanonical signaling pathways exist in many cells, they stimulate different sets of NF-κB dimers in each cell type. A number of previous studies have established the prevalence of specific NF-κB dimers in specific cell types, including resting dendritic cells (DCs), which containing a high abundance of RelB:p50 (11), whereas resting mouse embryonic fibroblasts (MEFs) primarily have RelA:p50 with RelA:RelA homodimers upon activation (21, 22), and both resting and activated B cells have cRel:p50 in addition to RelB:p52 as their predominant NF-κB dimers (23-25). Different cells in the B-lineage exhibit differential abundances of NF-κB dimer, with RelA:p50 as a major dimer in pre-B cells, cRel:p50 as a major dimer in mature B cells, and RelB:p52 as a major dimer in plasma cells (8, 26). Upon entering the germinal center (GC), antigen-selected B cells are dependent on cRel and RelB:p52 to traffic from light zones to dark zones. The dynamic switch from cRel-containing to RelA-containing NF-κB
dimers controls the switch from proliferative GC B cells to differentiated antibody-secreting B cells (8, 27). A RelA:cRel dimer may also play a role in macrophages, increasing the expression of genes encoding proinflammatory factors, such as TNF (28).

**A single NF-κB signaling network**

The biochemical interactions that constitute the NF-κB signaling network are conserved across cell types; however, this network is also capable of generating the cell type–specific repertoires of NF-κB dimers and their activation. The NF-κB network consists of three types of biochemical processes (29): (i) molecular mechanisms that control NF-κB dimer generation; (ii) the binding of IκB inhibitors which block NF-κB dimers from activating transcription; and (iii) enzyme-mediated phosphorylation and degradation of inhibitors to release NF-κB dimers so that they can translocate to the nucleus for DNA binding and transcriptional activation. Mathematical modeling of these processes using appropriate binding affinities and enzymatic kinetic rates has recapitulated myriad experimental results and provided insight into previously uncharacterized regulatory interactions and signal transduction in health and disease (5, 6, 8, 14, 30-32).

Review of these diverse experimental studies has produced a self-consistent knowledge base of the interaction network that enables NF-κB signaling in different cell types (fig. S1). This standard signaling network has been mathematically encoded, parameterized, and tested iteratively with experiments in multiple studies and multiple experimental systems. Thus, the interaction network, its network topology, and the biochemical interaction parameters represent a knowledge base that applies generally to mammalian cells. Many biochemical interactions and kinetic rate constants are expected to remain unchanged by cellular context. For example, one would not expect the
affinity of IκB for a particular NF-κB dimer to vary between cell types. Where mammalian cell types differ is in their chromatin epigenome, which controls the expression of genes, including those that encode NF-κB monomers. Indeed, analysis of the relative amounts of NF-κB monomers in various cell types shows substantial differences in RelA, RelB, and cRel (fig. S2) (33). This cell type–specific difference in gene expression may propagate through a cell type–independent molecular interaction network to confer cell type–specific responses. We will now review whether cell type–specific cell expression of NF-κB constituents may thus account for the cell type–specific functionality of the NF-κB signaling network.

Predicting the cell type–specific and stimulus-specific activation of NF-κB dimers

Basal cell type–specific NF-κB and IκB repertoires

Combining published measurements of the expression of genes encoding NF-κB monomers in MEFs, B cells, T cells, and DCs (fig. S2) (33) with the standard NF-κB signaling network model, results in cell type–specific NF-κB signaling systems that produce cell type–specific NF-κB dimer repertoires (Fig. 2A; for a description of the modeling methodology, see file xxx). Analyzing the steady-state abundances of NF-κB dimers reached by each cell type–specific model recapitulates many of the cell type–specific NF-κB dimer abundances observed in experimental studies. Consistent with experimental observations, simulated MEFs and T cells almost exclusively contain RelA:p50, whereas in silico, naïve B cells contain both cRel:p50 and RelA:p50, with DCs containing predominantly RelB bound to either p50 or p52 (Fig. 2B) (8, 9, 11, 34). The cell type–specific dimer repertoire is maintained in both the inhibited (latent) NF-κB dimers bound by IκBs and the much smaller proportion of unbound NF-κB proteins under unstimulated, steady-state conditions (Fig. 2, B and C).
The cell type–specific dimerization repertoire results in a cell type–specific repertoire of inhibitors, to which NF-κB dimers are bound under unstimulated conditions. For p50-containing dimers (RelA:p50, cRel:p50, and RelB:p50), IκBα is the primary inhibitor, which enables the rapid activation [based on its scaffolding function (35)], which is a hallmark of canonical NF-κB activation. IκBε is predicted to play a more important role in B cells and DCs, due to the increased abundance of cRel:p50 in these cells (Fig. 2B). Indeed, IκBε has a functionally important role in B cell proliferation (9). The noncanonical NF-κB dimer RelB:p52 is predicted to be at low abundance in all cell types except for DCs, which exhibit high amounts of Relb and nfkB2 mRNAs and increased NIK stability compared to other cell types (11). RelB:p52 is predicted to be predominantly inhibited by higher molecular-weight complexes of p100 (IκBδ) (Fig. 2, B and C). Cell type–specific NIK stability in DCs may be attributed to the altered abundances of the mediators of NIK degradation, such as TRAF3, which were not explicitly modelled here. Increasing the stability of NIK in MEFs, through reducing TRAF3 abundance, results in “DC-like” NF-κB signaling in MEFs (11). In silico modeling demonstrates substantial differences in NF-κB dimer abundance, IκB inhibitor repertoire, and basal NF-κB activity, as measured in multiple experimental studies, which may largely be an emergent behavior of the cell–type specific basal amounts of NF-κB monomers.

Cell type–specific inducible NF-κB responses

Different cell types exhibit the activation of different NF-κB dimers in response to the same stimuli. For example, RelA:p50 is the near-exclusive dimer in fibroblasts that is activated in response to inflammatory stimuli (22, 36), whereas RelB:p50 is strongly stimulated during DC
activation (11). In response to stimuli, cell-type differences in the different NF-κB dimers that are activated may be influenced by several factors, including receptor abundances, kinase activity levels, and the abundances of components of the NF-κB signaling network before stimulation. However, if in a given cell a specific receptor signaling pathway is functional, the NF-κB dimers that are activated through the NF-κB signaling network depend on the cell type–specific basal NF-κB and IκB repertoires. We simulated canonical IKK activation and noncanonical NIK activation using the cell type–specific models of the NF-κB signaling network (Fig. 3A). In response to canonical IKK activation, the NF-κB response in MEFs primarily shows dynamic RelA:p50 nuclear activity, as expected (Fig. 3B). Virtual B cells show a transient peak of activation of all of the canonical NF-κB dimers, which is followed by the increased late-phase activation of RelA:p50 and cRel:p50. This is consistent with studies showing a role for cRel:p50 in controlling B cell proliferation, as well as for RelA:p50 in differentiation (8-10). In simulations of DCs, both RelB:p50 and RelB:p52 play more substantial roles, consistent with studies identifying an important role for RelB in DC activation (11). In response to noncanonical activation through NIK, all cell types respond similarly, with RelB:p52 being stimulated with similar dynamics (Fig. 3C). This indicates that the architecture of the NF-κB signaling network enables relative monomer abundance to tune cell type–specific canonical NF-κB responses, while maintaining reliable and conserved responses to noncanonical, developmental stimuli (Fig. 3, B and C).

**Canonical and noncanonical NF-κB crosstalk**

Whereas the two NF-κB pathways are often described as separate, there are numerous interconnections between them. Studies have indicated that higher molecular weight complexes of p100 (IκBδ) and p105 (IκBγ), termed the IκBsome complex, are capable of inhibiting RelA:p50
activity (37). Thus, NIK-dependent p100 processing enables RelA:p50 activation after simulation of the noncanonical pathway (38). These studies suggest that chronic inflammatory signals should not give rise to increased RelA activity, as IκBα controls excess RelA:p50. However, given an environment in which there are tonic developmental signals in addition to chronic inflammatory signals, enhanced and sustained RelA activity may result. Additionally, previous work established the role of canonical NF-κB activity in providing substrates for noncanonical NF-κB activation (6, 39). The synthesis of RelB and p100 is dependent on RelA activity, and reconstitution of RelB in RelA-deficient MEFs restores noncanonical NF-κB activation (6, 39). These studies have implicated p100 as an important mediator between the two activation pathways, with important functions in regulating basal and activated NF-κB. However, overexpression of p100 in RelA-deficient MEFs fails to restore noncanonical RelB activity (6). These studies suggest that whereas RelB activation is regulated by basal RelA activity, p100 synthesis induced by RelA activation is a critical determinant of activation of the noncanonical pathway (6). Despite this substantial potential for crosstalk, competition between the roles of NIK in p52 processing and p100 release limits the effect of canonical signaling on the noncanonical NF-κB pathway in inflammatory environments (40). Pathway insulation may also be mediated by other, receptor-proximal mechanisms (13).

**Cell type–specificity of canonical and noncanonical NF-κB crosstalk**

In principle, the mechanisms reviewed earlier together with cell type–specific NF-κB monomer expression may lead to distinct crosstalk between the two NF-κB signaling pathways in different cell types. Modeling that encapsulates this knowledge enables exploration of how developmental stimuli affect signaling through the canonical, inflammatory NF-κB pathway. This crosstalk is
investigated in multiple cell types by simulating the response to IKK activity (maintained from Fig. 3A) in the context of different basal amounts of NIK (Fig. 4, A and B). Increasing NIK activity amplifies the dynamic response of canonical NF-κB dimers to IKK activation (Fig. 4C). The first peak of the canonical NF-κB dimer response is amplified by increasing NIK activity together with moderate amplification of late-phase activity, particularly for cRel:p50 and RelB:p50 in B cells (Fig. 4C). Simultaneous, noncanonical activation extends cRel responses in B cells (12). The activation of RelB:p52 is predicted to be largely independent of canonical pathway activity across multiple levels of basal NIK activity (Fig. 4C).

Conversely, simulations with a consistent physiological NIK activation profile (consistent with Fig. 3A) at varying levels of basal IKK activation (Fig. 5A) enable investigation of the crosstalk from inflammatory status on the response of a cell to developmental, noncanonical stimuli. As expected, the predominant dimer stimulated by NIK activation is RelB:p52, and this activation is largely unaffected by canonical pathway activity in both B cells and MEFs (Fig. 5, B and C). An increase in cRel:p50 abundance in response to NIK activation through BAFF is amplified at later times (6, 12, 24, and 40 hours) by the addition of IKK-activating IgM (12), which is recapitulated in our simulations (Fig. 5B). Whereas p52 is thought to primarily bind to RelB, simulations in MEFs with high NEMO activity and increased NIK stability predict the formation of RelA:p52 dimer at amounts comparable to those of RelA:p50 (fig. S3). This is consistent with experiments performed in MEFs, intestinal epithelial cells, and macrophages, showing concurrent activation of canonical and noncanonical signaling (41-43). Indeed, in both MEFs and B cells, the model predicts that the canonical dimers RelA:p50 and cRel:p50 can be moderately activated by noncanonical stimuli in environments with high basal inflammation (high tonic IKK activity) (Fig.
compare the response to increasing NIK activity in the context of 10 vs. 1% IKK activity). Aberrant stimulation of cRel activity is oncogenic; after all, it was named for being the cellular homolog of the oncogene v-Rel, which is encoded by the avian reticuloendotheliosis virus. The predicted activation of cRel in response to noncanonical (NIK-mediated) pathway activation may lead to oncogenic activation of cRel in response to noncanonical signals, such as CD40 receptor-ligand binding on the surface of GC B cells.

Predicting the regulatory consequences of cancer-associated NF-κB mutations

Mutations that affect NF-κB signaling system components are associated with various diseases, including multiple types of cancer, atherosclerosis, lupus, and Hodgkin’s disease (44, 45). Particularly in lymphoid malignancies, it is well established that NF-κB dysregulation is involved (46, 47); however, it is not always clear which of the NF-κB dimers is affected or dysregulated (48, 49).

Hodgkin’s B cell lymphoma

Hodgkin’s disease is one of the most frequent lymphomas and is caused by Hodgkin/Reed-Sternberg (H/RS) cells that are derived from B lymphocytes. Many H/RS cell lines show multiple coexisting mutations, and strong NF-κB activity results from many of these mutations, playing a causal role in the pathogenesis of these cells (50). Some 10 to 20% of classical Hodgkin’s disease cases have a mutation in the nfkbia gene, which encodes IκBα, a primary inhibitor of NF-κB in the canonical pathway (51). Adapting the mathematical B cell model with an IκBα deletion enables examination of the signaling effect of this mutation within the context of the knowledge base of NF-κB signaling. IκBα knockout simulations predict increased amounts of all canonical
dimers under basal conditions, which is consistent with canonical pathway activation being consistently observed in H/RS patient samples and cell lines (51). Furthermore, the model predicts a marked increase in nuclear RelB:p50 NF-κB abundance, recapitulating results showing that nuclear RelB:p50 is ubiquitously detected in Hodgkin’s lymphoma (HL) cell lines (Fig. 6A) (52). In addition to basal increases in canonical NF-κB dimer activity in IκBα knockout cells, the model also predicts that the activation of dimers is considerably affected, with RelA:p50 and cRel:p50 being largely nonresponsive to canonical pathway stimulation (Fig. 6A). HL cell lines also show increased nuclear RelB:p52 abundance, which is not predicted by simulations (52). Some of these discrepancies can be attributed to the EBV-positivity of some of the lines tested (for example, L591 cells) because EBV infection mimics CD40 activation through the viral protein LMP1, which activates noncanonical signaling (52, 53). However, even EBV-negative HL cell lines and almost all patient samples show constitutive increases in NIK abundance, and many of these cases can be attributed to the loss of TRAF2 or TRAF3 (negative regulators of NIK) or copy number gains of the gene that encodes NIK (MAP3K14) (54, 55). Although these NIK-activating mechanisms are not considered in our simulations (Fig. 6), the model provides a tool with which to disentangle the predicted effects of the recurrent loss of IκBα from the co-occurring mutations found in many HL cells. These simulations (Fig. 6A) show that the IκBα mutation alone can result in a substantial shift toward RelB-containing NF-κB dimers, even without concurrent NIK activation.

**NFκB2-truncation cancers**

Various lymphomas and leukemias, such as B cell chronic lymphocytic leukemia, multiple myeloma, and adult T-cell leukemia/lymphoma, have mutations in the nfkB2 gene, which encodes the protein p100 (56-58). Total loss of p100 in both knockout cell lines and multiple myeloma
cells prolongs the NF-κB response to stimuli, providing a pro-survival response (30, 59). Mutations in the gene encoding p100 often result in the deletion of its C-terminal, inhibitory ankyrin repeats (60), causing constitutive generation of a p52-like protein that can bind to Rel monomers and translocate to the nucleus (20, 61, 62). Whereas constitutive processing of p100 results in oncogenesis, the mechanisms remain unclear (63). Constitutive generation of p52 due to rearrangements in *nfkb2* results in both the increased abundance of p52 and loss of the higher molecular weight inhibitory complexes formed by p100 (IκBδ) (62). Both mechanisms may result in increased and oncogenic NF-κB activity. We can disentangle these two effects by comparing the effect of p100 truncation mutations (resulting in increased amounts of p52 and reduced amounts of IκBδ) with *nfkb2* deletion (resulting in loss of both p52 and IκBδ). *Nfkb2* knockout mice do not form tumors, suggesting that the generation of increased amounts of p52 resulting from C-terminal mutations is responsible for oncogenesis (64). Because p52 transgenic mice expressing wild-type p100 do not develop tumors (65), these studies suggest that both the loss of IκBδ function and the gain of transcriptional p52 function contribute to the oncogenesis of cells that contain rearrangements in *nfkb2*. Reports have suggested that increased nuclear RelA:p52 activity results from these genetic events, whereas other reports have suggested the formation of a p52 homodimer that is capable of binding to κB sites in the DNA (62, 66).

The identities of the specific NF-κB dimers that are activated in these tumors because of *nfkb2* rearrangement are still unclear. The mathematical model enables exploration of the signaling effect of the tumor-associated *nfkb2* mutations in the context of the conserved NF-κB signaling network. The *nfkb2* rearrangement is simulated by eliminating IκBδ activity and by removing the NIK dependence of p100 processing to p52, so that the processing is purely constitutive. These
Simulations predict minimal difference in basal canonical dimers in unstimulated DCs harboring \textit{nfb}2 truncation mutants (Fig. 6B). The model also predicts that RelB:p52 is the primary NF-κB dimer that is increased in abundance in unstimulated DCs that have the \textit{nfb}2 mutation (Fig. 6B). The absence of IκBδ is predicted to only mildly affect basal amounts of canonical dimers due to compensation by other members of the IκB family. However, in response to a canonical, IKK-activating stimulus, the increased inhibition of NF-κB dimers by IKK-responsive IκBs results in a hyper-responsive induction of the activities of RelA:p50, cRel:p50, and, to a lesser extent, RelB:p50 (Fig. 6B). Encoding available biochemical knowledge into the mathematical model demonstrates that the \textit{nfb}2 truncation mutants can result in increases in RelB:p52 nuclear abundance. Simulations of \textit{Nfb}2 truncation mutants also predict that this mutation may sensitize the canonical pathway to activation, indicating another crosstalk mechanism by which mutations affecting the noncanonical pathway may lead to oncogenic signaling in the canonical NF-κB pathway.

\textit{MYD88}-mutated diffuse large B cell lymphoma and Waldenström Macroglobulinemia

In diffuse large B cell lymphoma (DLBCL), nuclear translocation of one or more of the five NF-κB monomers (cRel, RelA, RelB, p50, and p52) has been identified, with each subunit being activated with varying frequency in patient samples (67-72). RelB is activated in about 60% of patient samples, whereas RelA is almost always active (72). Gain-of-function mutations in \textit{MYD88} are found in many of these cancers (73-77). The gain-of-function mutation of \textit{MYD88} in DLBCLs results in high constitutive NF-κB activity (78). \textit{MYD88}^{L265P} also occurs in ~90% of Waldenström Macroglobulinemia (WM) cases. \textit{MYD88}^{L265P} results in activation of the canonical NF-κB pathway and RelA-dependent gene expression (79). However, introducing \textit{MYD88}^{L265P} into
splenic murine B cells does not increase RelA activation \((80)\). Therefore, it is unclear to what extent inducible NF-κB inhibitors may overcome the activating effect of \(MYD88^{\text{L265P}}\) and which of the specific NF-κB dimers are activated by this mutation.

The scope of the consensus NF-κB system model does not include receptor-proximal interactions that converge on IKK. To explore the effect of \(MYD88\) mutations, we leveraged a previously constructed mathematical model that represents a knowledge base of TLR-proximal signaling mechanisms (Fig. 7A) \((81)\). \(MYD88^{\text{L265P}}\) results in spontaneous activation of downstream signaling \((82)\), and analogous model simulations show increased TRAF6 activity, as was seen experimentally (Fig. 7A) \((82)\). By taking the predicted increase in IKK activity with \(MYD88^{\text{L265P}}\) as an input to the NF-κB signaling networks, the change in the basal and inducible NF-κB dimer repertoire is predicted. This increased IKK activity is combined with the different extents of NIK activity that may be the result of the tumor microenvironment, such as interactions with T cells or secreted cytokines (Fig. 7B).

Simulations of WT vs. \(MYD88^{\text{L265P}}\) B cells predict a large increase in the abundances of complexes of p100 (IκBδ) in unstimulated \(MYD88^{\text{L265P}}\) B cells (Fig. 7C). Indeed, p100-mediated signaling has been reported in ABC-DLBCL patient samples, particularly in tumors harboring \(MYD88^{\text{L265P}}\) mutations \((83, 84)\). Increases in RelA:p50 and cRel:p50 abundance are also predicted under unstimulated conditions but they remain largely inactive because they are bound by inhibitory IκBδ (Fig. 7D). In response to increasing NIK activity, this inhibitory IκBδ is degraded (Fig. 7C) resulting in substantial induction of nuclear RelA:p50 and cRel:p50 activities in \(MYD88^{\text{L265P}}\) B cells (Fig. 7D). In contrast, RelB:p50 is predicted to be increased in \(MYD88^{\text{L265P}}\) B cells under
basal conditions, but the abundance of RelB:p50 in MYD88\textsuperscript{1,265P} B cells is not substantially increased by NIK activity. This is a result of the NIK-mediated induction of p52 activity and hence RelB:p52 dimer formation, thereby limiting the amount of RelB available for RelB:p50 formation. Furthermore, due to the abundance of IκBδ in MYD88\textsuperscript{1,265P} mutant simulations, RelB:p52 is slightly reduced due to “substrate complex competition” (40). Together, these simulations predict that Myd88\textsuperscript{1,265P} may unlock crosstalk from noncanonical pathway activation to canonical cRel- and RelA-containing NF-κB dimers. The result is that when Myd88\textsuperscript{1,265P} is present in B cells, developmental signals such as those from CD40 may lead to the oncogenic activation of cRel.

Incorporating knowledge of cancer-associated signaling mutations into the regulatory knowledge encoded in a mathematical model illustrates how microenvironmental signals may have unexpected, nonlinear effects in tumor cells.

**Conclusions**

A key goal of systems biology is to represent our understanding of a biological regulatory system in the language of mathematics, whose precision and quantitative claims drive iterative experimental studies and thus research progress. Such mathematical models may then function as knowledge bases that can be leveraged for studying diverse biological phenomena. The best characterized molecular network may be the one responsible for the metabolism of sugars and carbohydrates (2, 3). This knowledge base has been exploited to drive research in numerous prokaryotic and eukaryotic organisms and in the analysis of numerous human cell types and disease contexts. Here, we showed that the same goals may be achieved with regulatory or signaling network models. The NF-κB signaling network model may be applied to different cell types, considering different stimulation contexts, which govern pathway crosstalk, and in how
disease-causing mutations affect stimulus-response behavior. The mathematical model enables the use of cell type–specific gene expression data alone to generate insights about cell type–specific signaling behavior.

The present knowledge base is restricted to the NF-κB signaling systems themselves; however, a similar approach may enable extension of the model to receptor-associated signaling modules that activate this pathway (15, 81, 85) or downstream gene regulatory modules (81, 86, 87) Other efforts in the field have focused on single-cell macrophage responses to proinflammatory stimuli, requiring models to account for the observed cell-to-cell heterogeneity (81, 88). In all cases, the goal of representing all observations by a single model whose expression parameters are cell type–specifically adjusted and potentially distributed to account for cell-to-cell heterogeneity is in principle achievable.

Given the increasing availability of single-cell RNA-seq (scRNA-seq) data and descriptions of cell differentiation trajectories within a Waddington landscape, mathematical models that are reliable knowledge bases of quantitative interaction networks may be used to extrapolate from scRNA-seq datasets to predict the trajectory of how the functional responses of an individual cell to stimuli alter during development. Further studies focused on diseased tissues could extrapolate functional characteristics of cells based merely on gene expression data. Therefore, leveraging the knowledge base of a reliable mathematical model may enable unprecedented insight for research about diverse physiological and pathological questions. Whereas broad NF-κB inhibition often results in on-target toxicity, efforts to develop therapeutic compounds that target specific NF-κB dimers in
specific pathological settings may also benefit from iterative systems biology studies with the described consensus NF-κB multi-dimer system model.

Supplementary Materials
Figs. S1 to S3.
Supplementary text.

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**Fig. 1. Two signaling pathways are mediated by a single NF-κB network.** Schematic illustration of the canonical and noncanonical NF-κB activation pathways that regulate the indicated cellular functions. These functions are broadly characterized as transient and reversible immune responses and longer lasting immune development, respectively. Whereas the regulatory architecture of the signaling networks is conserved between cell types, cell types differ in which
NF-κB dimers they generate and activate in response to a given stimulus. The cell type–specific dimer repertoire is, at least in part, a function of the cell type–specific abundances of NF-κB monomers. Dysregulation of the network in any of the many cells it operates in may result in diverse pathologies, including cancer, inflammatory disease, and autoimmune disease.

Fig. 2. Cell type–specific NF-κB repertoires generated by cell type–specific simulations of the NF-κB network model. (A) Schematic of the process used to generate cell type–specific models through varying only the abundances of the indicated NF-κB monomers while conserving all reactions and kinetic parameters. M, mouse embryonic fibroblasts; B, B cell; T, T cell; DC, dendritic cell. (B) Bar graph showing the predicted abundances of RelA:p50 (A:50), cRel:p50 (C:50), RelB:p50 (B:50), and RelB:p52 (B:52) at steady state in cell type–specific simulations. Colors indicate the inhibitory (IκB) binding partners of each NF-κB dimer. (C) Bar graph showing the NF-κB-dimer composition of unbound NF-κB [dark blue in (B)].

Fig 3. NF-κB dimer activation in response to canonical and noncanonical stimuli in the cell type–specific computational models. (A) Schematic showing the cell-specific NF-κB monomer abundances that were used to generate cell-specific models that all received the same canonical (IKK-dependent) or noncanonical (NIK-dependent) activity profiles representing activation of canonical and noncanonical pathways. (B) Line graphs of the abundances of the indicated NF-κB dimers in response to the canonical (IKK-dependent) stimuli indicated in (A) in each cell type–specific model. (C) Line graphs of the indicated NF-κB dimer abundances in response to the noncanonical (NIK-dependent) stimuli indicated in (A).
Fig. 4. Simulation of canonical (IKK-mediated) pathway activation at different basal levels of NIK activity. (A) The physiological IKK activity curve used as input for all heatmaps. (B) Surface plot of RelA:p50 (A:50) abundance (A) in simulations with a physiological IKK activity curve and the indicated NIK abundances. (C) Heatmaps of the nuclear concentrations of the indicated NF-κB dimers in simulations with a physiological IKK activity curve and the indicated NIK abundances.

Fig. 5. Simulation of noncanonical (NIK-mediated) pathway activation at different basal levels of IKK activity. (A) The physiological NIK activity curve used as input for all subsequent heatmaps. (B) Surface plot of RelB:p52 (B:52) abundance in simulations with the physiological NIK activity curve and the indicated basal IKK abundances. (C) Heatmaps of the nuclear concentrations of the indicated NF-κB dimers in simulations with a physiological NIK activity curve with the indicated basal IKK abundances.

Fig. 6. NF-κB dimer repertoire in simulations of IκBα loss and nfk2 truncation cancers. (A) Line graphs of the abundances of the indicated NF-κB dimers in simulations of WT B cells (dashed lines) and IκBα-deficient B cells (solid lines). (B) Line graphs of the abundances of the indicated NF-κB dimers in simulations of WT DCs (dashed lines) and DCs expressing mutated truncated nfk2 (p100ΔC, solid lines). For (A) and (B), the IKK input curve used is from Fig. 4A and the abundance of NIK is kept constant.

Fig. 7. NF-κB dimer repertoire in simulations of DLBCL. (A) Schematic of the model of Cheng et al. (2015) that was used to simulate signaling from a TLR to active IKK (IKK*). The effect of
MYD88^{L265P} is indicated in purple (see Supplementary text). Constitutive receptor shuttling, independent of ligand binding, is omitted from the diagram. Bar graphs indicate the abundances of the indicated molecular species in WT (green) and MYD88^{L265P} (purple) simulations. (B) Line graphs of IKK (left) and NIK (right) activity curves for WT B cells (green) and MYD88^{L265P} B cells (purple). (C) Line graphs of IκBδ responses in WT B cells (green) and MYD88^{L265P} B cells (purple) in response to the indicated kinase activities in panel B (D) Line graphs of the abundances of the indicated NF-κB dimers in simulations of WT B cells (green) and MYD88^{L265P} B cells (purple) in response to the indicated kinase activities in (B).
Figures

Figure

Cell-type-independent biochemical properties
- binding kinetics
- enzyme kinetics

Cell-type-specific epigenetic landscape
- expression rates

<table>
<thead>
<tr>
<th>Disease</th>
<th>Health</th>
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</thead>
<tbody>
<tr>
<td>Inflammatory disease</td>
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<tr>
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<td>Cell survival</td>
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<tr>
<td>Inflammatory disease</td>
<td>Cell survival</td>
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<td>Lymphoid organogenesis</td>
<td>Autoimmune disease</td>
</tr>
</tbody>
</table>
Figure 3

A

Cell-type specific Monomer Expression

Transcription rates

Measurement

Model

B

MEFs

B cells

T cells

DCs

Nuclear Abundance (nM)

Time (h)

C

MEFs

B cells

T cells

DCs

Nuclear Abundance (nM)

Time (h)
Figure 4

(A) IKK Activity % over time (h).

(B) 3D graph showing NIK (nM) over time (h).

(C) Heat maps for MEFs and B cells showing NIK (nM) over time (h).
Figure 6

A

B Cell

WT
IkBα

B:50

B:52

$nM$

Time (h)

0 10 20

10 20 30

50

0 10 20

Time (h)

0 10 20

100 150

250

B

Dendritic Cell

WT
$p100ΔC$

P52:p52

A:50

C:50

B:50

B:52

$nM$

Time (h)