Duality, thermodynamics, and the linear programming problem in constraint-based models of metabolism

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It is shown that the dual to the linear programming problem that arises in constraint-based models of metabolism can be given a thermodynamic interpretation in which the shadow prices are chemical potential analogues, and the objective is to minimise free energy consumption given a free energy drain corresponding to growth. The interpretation is distinct from conventional non-equilibrium thermodynamics, although it does satisfy a minimum entropy production principle. It can be used to motivate extensions of constraint-based modelling, for example to microbial ecosystems.

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In biology, the metabolism of an organism provides energy and raw materials for maintenance and growth. As such, an interesting and important question concerns the application of thermodynamics to metabolic reaction networks [1, 2, 3, 4]. For example, Prigogine and Wiame suggested a long time ago that an organism’s metabolism might be governed by a minimum entropy production (MEP) principle [5]. From the physical point of view, a metabolic reaction network is an excellent example of a system in a non-equilibrium steady state, since one can usually assume that the metabolite concentrations are unchanging after a short transient relaxation period. The appropriate generalisation of thermodynamics and statistical mechanics to non-equilibrium steady states is a large field [6], which continues to attract attention to this present day [7]. In this Letter, we show that a novel thermodynamic interpretation can be given to the dual linear programming problem which arises in constraint-based models of metabolism. The resulting interpretation is rigorously defined, and uniquely determined by the mathematics. It is closely analogous to, but distinctly different from, conventional non-equilibrium thermodynamics. We also show that it satisfies an MEP principle similar to that proposed by Prigogine and Wiame.

Constraint-based modelling (CBM) of metabolic networks has been pioneered by Palsson and co-workers [8]. In a typical application, described in more detail below, the steady-state assumption is combined with a target function to make a linear optimisation or linear programming (LP) problem. The LP variables are the fluxes through the various reactions that comprise the network, and the LP constraints arise from basic considerations of stoichiometry and from the reversibility or otherwise of the reactions. The LP objective function is biologically motivated, for example a ‘growth’ reaction is commonly inserted, and the target is to maximise flux through this reaction to correspond to maximal growth rate. CBM has been applied to microorganisms from all three domains of life [9, 10, 11], and has been remarkably successful in predicting phenotypic behaviour [12, 13, 14].

Mathematically, every LP problem has a unique dual

\[ \sum_{i=1}^{m} \pi_i b_i \]

subject to \( \sum_{i=1}^{m} \pi_i A_{i\alpha} \geq a_{\alpha} \), with no restriction on the sign of dual variables \( \pi_i \). The LP strong duality theorem guarantees that max \( z = \min w \) provided both problems have optimal solutions. In addition, at optimality, ‘complementary slackness’ (CS) conditions hold. To formulate these, first define the ‘slack’ in the inequalities in the dual problem to be \( y_{\alpha} = \sum_{i=1}^{m} \pi_i A_{i\alpha} - a_{\alpha} \). The CS conditions state that the inequalities \( x_{\alpha} y_{\alpha} \geq 0 \) are saturated (i.e. = 0) at optimality, and only at optimality.

In many applications of LP, the dual problem can be given an economic interpretation, which has led to the dual variables being generically known as ‘shadow prices’. We note that shadow prices can be obtained directly from

FIG. 1: Schematic metabolic network for a prokaryote like E. coli, showing intracellular metabolites (open circles), extracellular metabolites (hatched circles), internal and exchange reactions (arrows), and a growth reaction (dashed arrow).
the solution to the primal problem \[15\], so the dual problem need never be explicitly formulated. This may be the reason why the remarkably simple structure of the dual problem in CBM has not been described before. The use of shadow prices in CBM was pioneered by Varma and Palsson to assess efficiencies in a model of the central metabolism of \textit{E. coli} [16].

Now let us turn to the LP problem in CBM. We start with the set of chemical rate equations that describe the metabolic reaction network, $dc_i/dt = \sum_\alpha S_{i\alpha} v_\alpha$, where the $c_i$ are metabolite concentrations, the $v_\alpha$ are reaction velocities or fluxes, and $S_{i\alpha}$ is a stoichiometry matrix giving the number of moles of the $i$th metabolite involved in the $\alpha$th reaction. Making the steady-state assumption, the chemical rate equations reduce to a set of flux-balance conditions $\sum_\alpha S_{i\alpha} v_\alpha = 0$. At this point, in CBM, attention shifts from the metabolite concentrations to the reaction fluxes. From this point of view, the flux-balance conditions become a set of linear constraints on the $v_\alpha$. In addition, one usually imposes the ‘thermodynamic’ constraint that $v_\alpha \geq 0$ if a reaction is irreversible.

In modern approaches [8], the reactions in the network are elementally and charge-balanced. To make the network ‘do’ something, two kinds of imbalanced reactions are typically added. The first, as mentioned already, is a growth reaction. This reaction drives the endpoints of metabolism in the appropriate ratios and represents the combined effect of the biochemistry subsequent to metabolism. The flux through the growth reaction (the growth rate) will be labelled $v_{gr}$. The second type of imbalanced reaction is an ‘exchange’ reaction, which represents the exchange of an extracellular metabolite with the environment (the model additionally includes transporter reactions which allow extracellular metabolites to enter and leave the intracellular environment). The exchange reactions enable the uptake of food substrates, trace minerals, dissolved gases, and vitamins; and the discharge of metabolic waste products. By convention, a positive (negative) flux through an exchange reaction represents the discharge (uptake) of the corresponding metabolite. Exchange reactions may be reversible, or irreversible if discharge only is possible. A special case arises when one wishes to represent limited availability, for example of a food substrate. In this case the exchange flux is allowed to become negative to a limited extent, thus $v_\alpha \geq -v_\alpha^{\text{min}}$ with $v_\alpha^{\text{min}} > 0$ representing a ‘cap’ on the (negative) reaction flux. The value of $v_\alpha^{\text{min}}$ is typically empirically determined to agree with experimentally measured uptake rates. Fig.1 shows schematically how the exchange reactions and the growth reaction are connected into the rest of the metabolic network. For high accuracy work, an ATP $\rightarrow$ ADP maintenance reaction with a specified flux is sometimes included in the model [11]. We omit this here although it can easily be accommodated with a small extension to the formalism.

The LP problem in CBM is then to find values for the fluxes $v_\alpha$ which maximise $v_{gr}$ subject to the above constraints. This is summarised in Table II. Usually it is the limited availability of substrates through the exchange reactions that prevents the problem being unbounded. Technically the LP problem is not quite in the standard form but it does not take much to make it so.

We formulate the dual to the above (primal) LP problem following the textbook approach described above. After some straightforward simplifications, the following picture emerges. Each metabolite has an associated shadow price $\pi_i$ which is unrestricted in sign. Each reaction has an associated constraint, of the form $\sum_i \pi_i S_{i\alpha} = 0$ for reversible and unlimited exchange reactions, $\sum_i \pi_i S_{i\alpha} \geq 0$ for irreversible and limited exchange reactions, and $\sum_i \pi_i S_{i\alpha} \geq 1$ for the growth reaction. The objective function is $w = \sum_\alpha v_\alpha^{\text{min}} |B_\alpha|/B^*$ where the sum is over the limited exchange reactions only. The LP problem is to find values for the shadow prices $\pi_i$ which minimise this objective function subject to the constraints. Note that, although the $v_\alpha^{\text{min}}$ appear in the dual objective function $w$, these are numerical constants common to both the primal and dual problems. The actual fluxes

| Variables | Primal | Dual | CS inequalities |
|-----------|--------|------|-----------------|
| flux balance / thermodynamics | $\sum_\alpha S_{i\alpha} v_\alpha = 0$ | $B_\alpha = \sum_\alpha \mu_i S_{i\alpha}$ | $v_\alpha B_\alpha \leq 0$ |
| reversible reactions | $v_\alpha$ unlimited | $B_\alpha = 0$ | |
| irreversible reactions | $v_\alpha \geq 0$ | $B_\alpha \leq 0$ | $v_\alpha B_\alpha \leq 0$ |
| growth reaction | $v_{gr} \geq 0$ | $B_\alpha \leq -B^* < 0$ | $(v_\alpha + v_\alpha^{\text{min}}) B_\alpha \leq 0$ |
| limited exchange reactions | $v_\alpha \geq -v_\alpha^{\text{min}}$ | $B_\alpha \leq 0$ | |
| objective function | $z = v_{gr}$ | $w = \sum_\alpha v_\alpha^{\text{min}} |B_\alpha|/B^*$ | |

TABLE I: The primal and dual linear programming problems in constraint-based models of metabolism. In the dual objective function, the sum is over the limited exchange reactions only. The complementary slackness (CS) inequalities are saturated (i.e. $= 0$) at optimality, where also $\max z = \min w$. The shadow prices for the primal problem are given by $-\mu_i/B^*$. |
We now show that the dual problem admits a striking thermodynamic interpretation. The motivation is the standard expression for the free energy change in a chemical reaction, or reaction affinity, $A_\alpha = \sum \mu_i S_{i\alpha}$, where the $\mu_i$ are chemical potentials [6]. The similarity between this, and the rules for formulating the dual LP problem above, make it natural to interpret the shadow prices as chemical potential analogues. To aid the interpretation, we rescale the dual problem by a factor $-B^* < 0$, set $\mu_i = \pi_i B^*$, and write $B_\alpha = \sum \mu_i S_{i\alpha}$ as the analogue of reaction affinity. The resulting thermodynamic formulation of this rescaled dual LP problem is summarised in Table I. We have introduced $B_\alpha$ to distinguish our interpretation from conventional non-equilibrium thermodynamics; in general $B_\alpha \neq A_\alpha$, as explained in more detail below.

Let us discuss the thermodynamic interpretation in a bit more depth. We see that the constraints assert that $B_\alpha = 0$ for a reversible reaction, and $B_\alpha \leq 0$ for an irreversible reaction. These are precisely in accord with equilibrium chemical thermodynamics. In addition we interpret the fact that $B_\alpha \leq -B^* < 0$ for the growth reaction to mean that a minimum free energy drain equal to $B^*$ is required for growth. The magnitude of $B^*$ sets the overall energy scale, and can be arbitrarily chosen. Finally, in the rescaled dual LP problem the objective is to minimise $wB^* = \sum v_{\alpha}^\mathrm{min}|B_\alpha|$, in other words a weighted sum of the free energy consumption associated with the limited exchange reactions. At optimality, one has $\max z = \min w$, hence the growth rate is easily calculated from the solution to the thermodynamic LP problem using $v_{gr} = \sum v_{\alpha}^\mathrm{min}|B_\alpha|/B^*$.

We now show that the formalism satisfies an MEP principle. To derive this, we consider the internal entropy production due to the chemical transformations, $T\dot{S} = -\sum v_{\alpha}B_\alpha$. The sum excludes the exchange reactions since the flux balance condition implies the total entropy production $\sum v_{\alpha}B_\alpha = 0$ when the sum is over all reactions. It is straightforward to show that $z \leq T\dot{S}/B^* \leq w$. Thus, at optimality, the entropy production is ‘pinched’ between the two objective functions. Alternatively, for a fixed growth rate $v_{gr}$, one has $v_{gr}B^* \leq T\dot{S}$. Since the minimum value is attained at the combined solution of the primal and dual problems, this gives the desired MEP principle.

What does the dual solution look like for a constraint-based model of metabolism? To give an example, we computed the chemical potential analogues for a genome-scale model of the metabolism of *E. coli* [17]. Lack of space precludes a detailed discussion of the results, but we find that the vast majority of chemical potentials are positive and there is a broad distribution over several decades of magnitude. An interesting observation is that the chemical potentials increase with increasing molecular complexity. This is shown in Fig. 2 using molecular weight (discounting metal ions) as a stand-in for molecular complexity. This correlation arises because the chemical potential of a complex molecule is given approximately by the sum of the chemical potentials of its constituent parts. This in turn follows from the CS conditions which imply $B_\alpha = \sum \mu_i S_{i\alpha} = 0$ for all reactions with a flux $v_{\alpha} \neq 0$ (see further discussion below).

Let us discuss our findings in a wider context. Our results show that the dual to the CBM LP problem has a thermodynamic interpretation in which the dual variables are analogous to chemical potentials. The rules to formulate this thermodynamic LP problem are summarised in Table I. In principle we could strip away the CBM ‘scaffolding’ and let the thermodynamic LP problem stand on its own, since LP duality guarantees this is equivalent to solving the original (primal) LP problem. Such a viewpoint motivates a number of interesting questions.

Firstly, a technical point arises since the primal LP problem is often degenerate, in the sense that alternative optimal flux distributions exist [13]. This reflects the fact that multiple pathways may exist in the metabolism. But this is not a serious problem, for the dual problem will be similarly degenerate but the strong duality theorem and the CS conditions still hold, allowing one to move from a solution of the dual problem to a solution of the primal problem, and vice versa.

A more serious discussion point concerns the relationship to conventional non-equilibrium thermodynamics. For the reversible reactions, $B_\alpha = 0$ is a constraint. For the irreversible reactions, the CS conditions (Table I) show that $B_\alpha = 0$ if there is a flux ($v_\alpha > 0$) through a reaction, and $B_\alpha < 0$ only if there is no flux ($v_\alpha = 0$).
through the exchange reactions of the extracellular metabolites. Of course LP duality means there is a corresponding primal model (in this case the primal objective function becomes a weighted sum of growth rates \( \alpha \)).

Further exploration of this we leave to future work. It would, for example, be an interesting exercise to compare the \( E. coli \) shadow prices with what is known about the thermodynamic metabolic state of this organism \([2]\). We should emphasise that our MEP principle is couched in terms of \( B_\alpha \) and not \( A_\alpha \), and therefore our results do not constitute a proof of the original proposition of Prigogine and Wiame \([3]\).

Another interesting remark is that the choice in the primal LP problem to maximise the flux through a growth reaction seems to be ‘pure biology’. Experiments demonstrate that this works well under controlled conditions \([12, 13]\), and it can be supported by examining population dynamics for continuous culture growth in a chemostat \([19]\). Other choices could and perhaps should be made in different circumstances \([20, 21, 22]\). In terms of the thermodynamic LP problem, this biologically-motivated component is translated into the existence of a growth reaction with a minimal free energy drain \( B^* \). We could turn this observation to our advantage, to suggest extensions to the CBM approach which are perhaps unobvious in the primal LP problem. Consider for example metabolism in a microbial ecosystem, comprising multiple species which share a pool of common extracellular metabolites. The obvious generalisation of the thermodynamic LP problem is to include a growth reaction with a minimal free energy drain \( B^\ast \) for each organism, and seek to minimise the free energy consumption of the ecosystem through the exchange reactions of the extracellular metabolites. Of course LP duality means there is a corresponding primal model (in this case the primal objective function becomes a weighted sum of growth rates \( \alpha \)).

Further exploration of this we leave to future work.

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