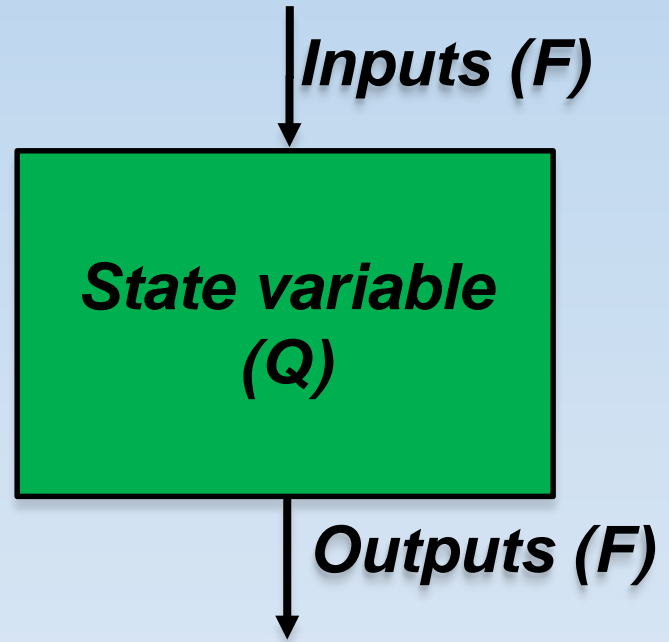


# Introduction and model construction



Timothy J. Hackmann<sup>1</sup>, Mark D. Hanigan<sup>2</sup>, Veridiana L. Daley<sup>3</sup>

<sup>1</sup>University of Florida

<sup>2</sup>Virginia Tech

<sup>3</sup>University of Kentucky

# Learning objectives

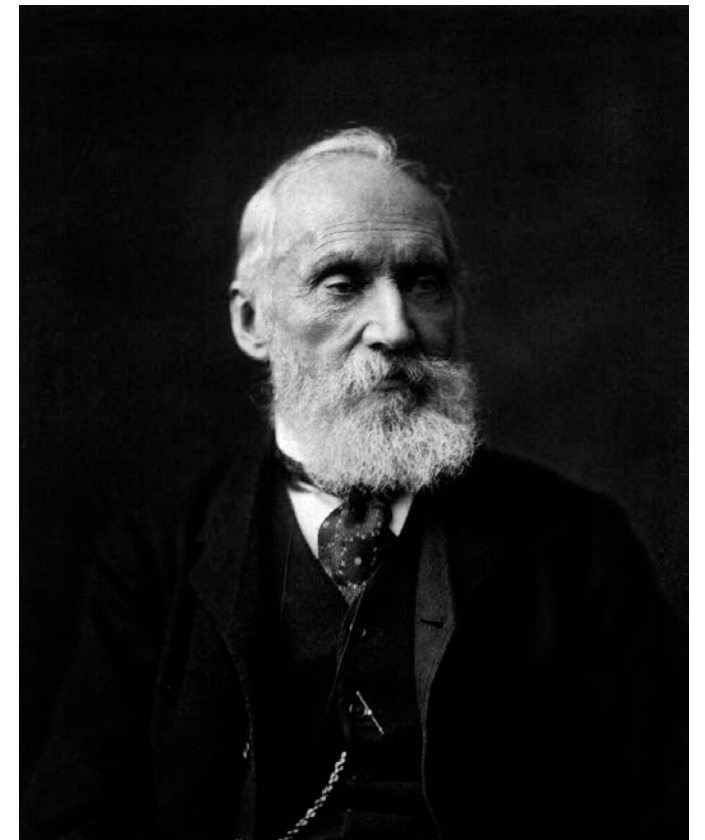
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- Explain the motivation for modeling
- Compare and contrast different types of models
- Outline the steps of constructing and evaluating a model

# Why modeling?

*“When you cannot express [something] in numbers, your knowledge is of a meager and unsatisfactory kind . . . you have scarcely, in your thoughts, progressed the level of science.”*

Lord Kelvin



# Why modeling?

## WEAK HYPOTHESIS

*“NDF is a major factor determining the level of feed intake by cattle.”*



## STRONG HYPOTHESIS

$$\text{Intake} = 1.2 \times \text{BW} / \% \text{ NDF}$$



Equation: Mertens. 1985. In Georgia Nutr Conf. pp. 1-18.

Images: [app.hedgeye.com/insights/37422-cartoon-of-the-day-muscle-vs-russell?type=video](http://app.hedgeye.com/insights/37422-cartoon-of-the-day-muscle-vs-russell?type=video)

# Goal of modeling

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- Take a hypothesis
- Convert it a system of equations
- Determine how well the equations describe reality

# Types of models

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- Empirical vs. mechanistic
- Deterministic vs. stochastic
- Static vs. dynamic

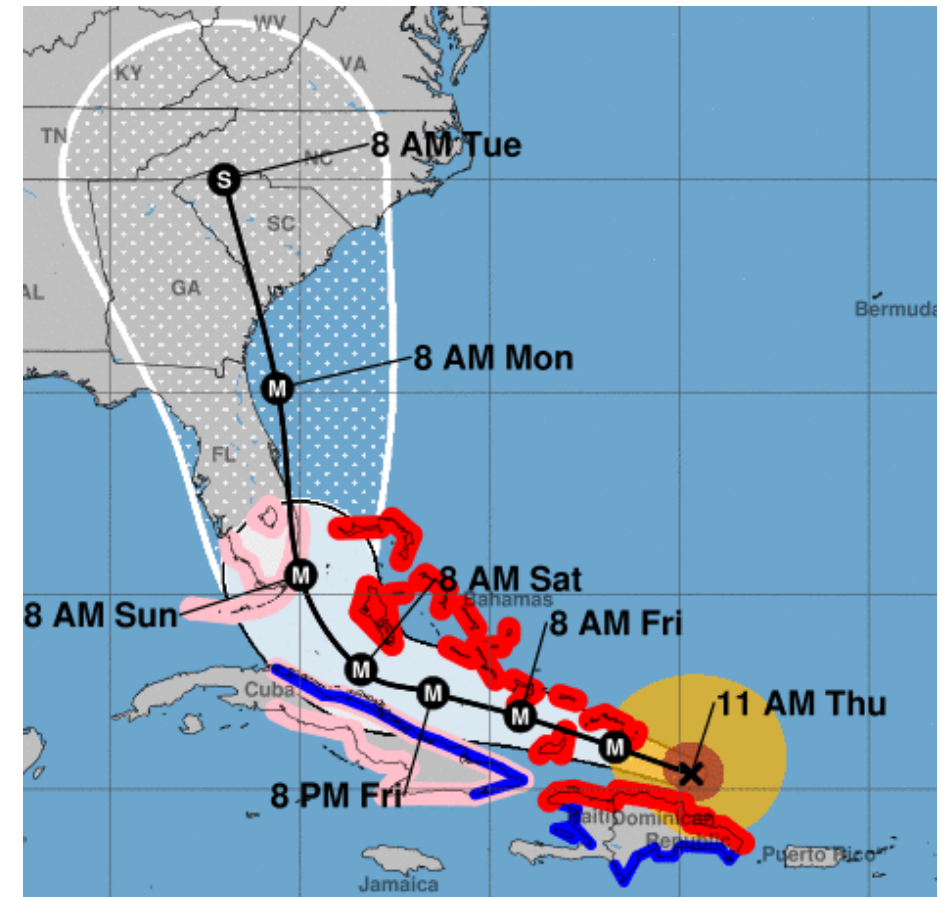


# Empirical vs. mechanistic

## □ Mechanistic

- Represents underlying biological or physical causes
- Example: weather models

Hurricane trajectory simulation





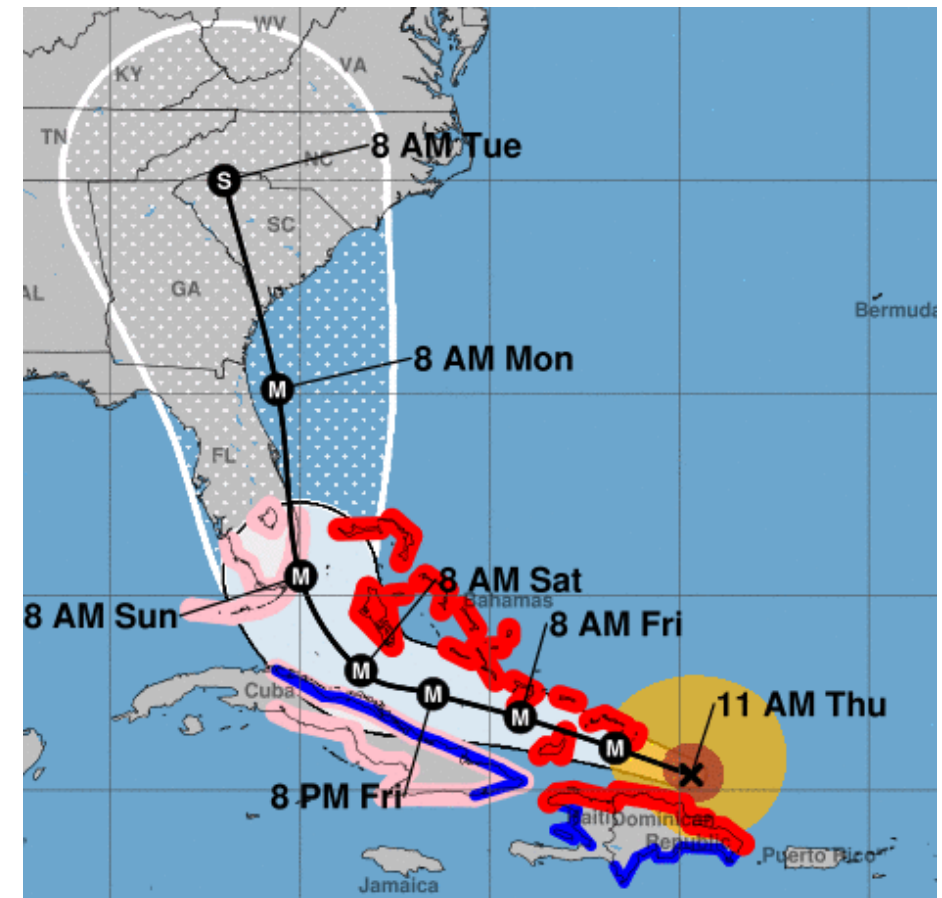


# Deterministic vs. stochastic

## □ Stochastic

- ▣ Assumes variables and thus predictions have uncertainty
- ▣ Example: weather models

Hurricane trajectory simulation



# Static vs. dynamic

## □ Dynamic

### □ Variables are

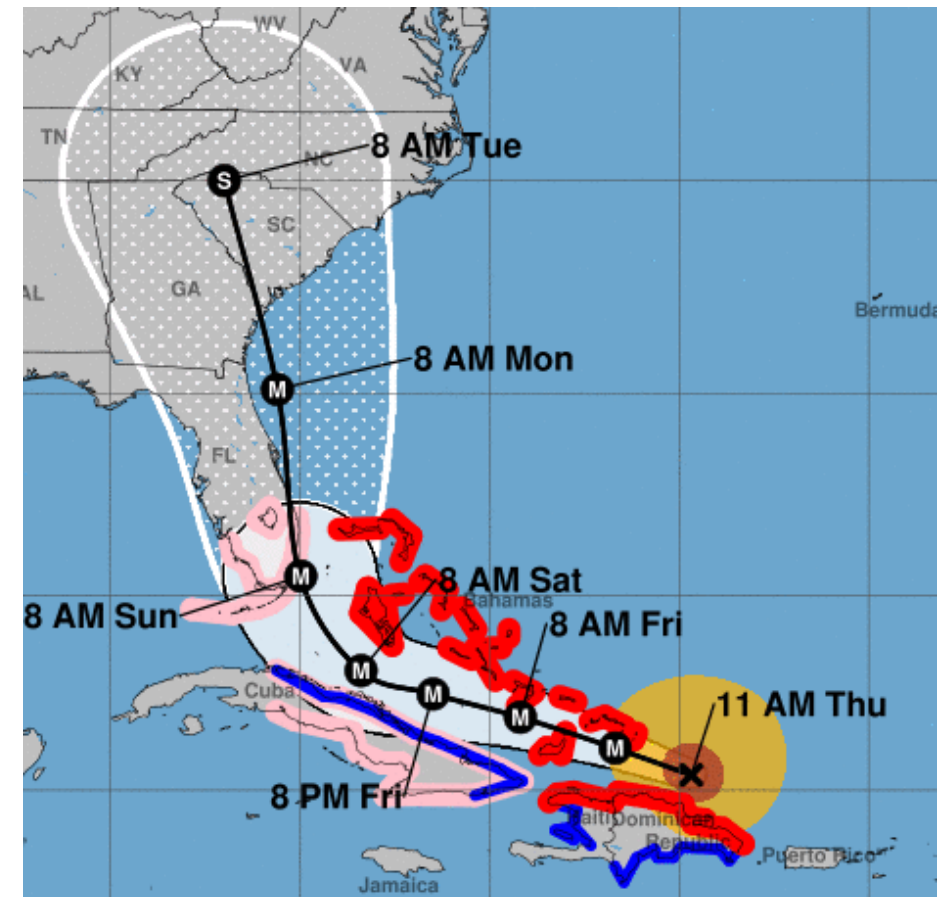
- a function of time
- represented in differential equations (Baldwin, 1995)

### □ Model has a “runtime”

- States change over time
- State at  $t$  is starting point for predicting state at  $t + 1$

### □ Example: weather models

## Hurricane trajectory simulation



Reference: Baldwin. 1995. Modeling ruminant digestion and metabolism. Chapman & Hall.

Image: [https://www.nhc.noaa.gov/archive/2017/IRMA\\_graphics.php?product=5day\\_cone\\_with\\_line\\_and\\_wind](https://www.nhc.noaa.gov/archive/2017/IRMA_graphics.php?product=5day_cone_with_line_and_wind)



# Examples

- Dairy NRC
  - Empirical, deterministic, static
- CNCPS (v. 6.5)
  - Semi-mechanistic, deterministic, static
- Molly
  - Mechanistic, deterministic, dynamic

# Organizational levels

<u>Description</u>	<u>Level</u>
Herd	$i + 1$
Animal	$i$
Organ	$i - 1$
Tissue	$i - 2$
Cell	$i - 3$

} Data for empirical model

} Data for mechanistic model

# Steps

---



Identify objective



Draw block diagram



Write equations



Define parameters values



Solve model and generate predictions



Evaluate predictions

# Objective

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- Defines
  - Goal
  - Model type
  - Organizational levels



# Objective

- *Defines*

- *Goal*

- *Model type*

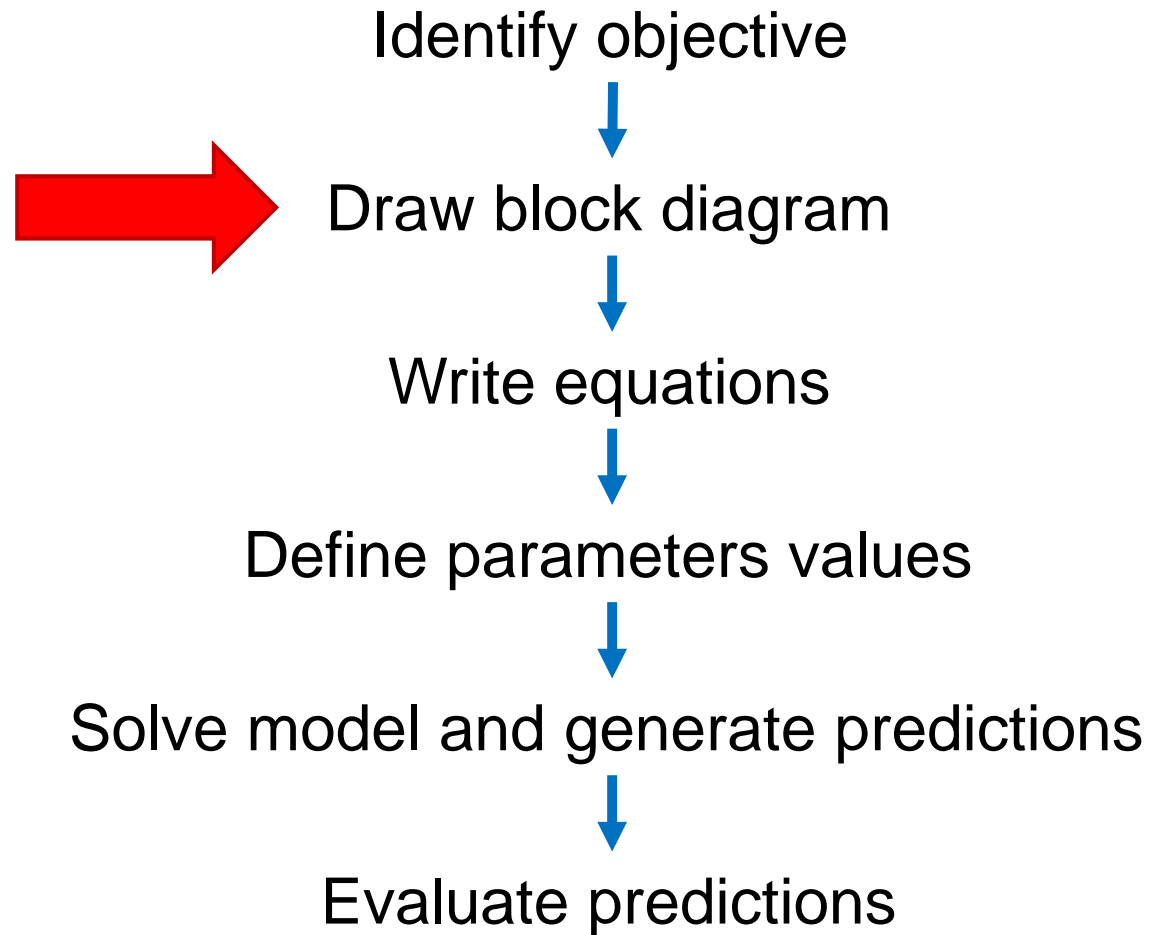
- *Organizational levels*

- **Example**

- ***“Develop a mechanistic, dynamic, deterministic model to predict digestion of feed carbohydrate in the rumen”***

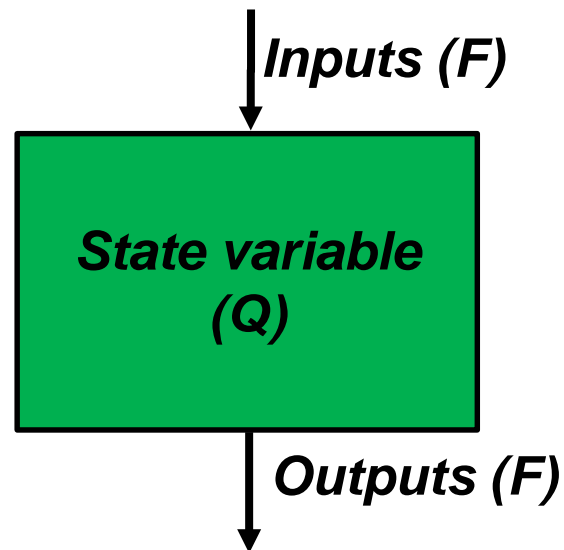
# Steps

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# Block diagram

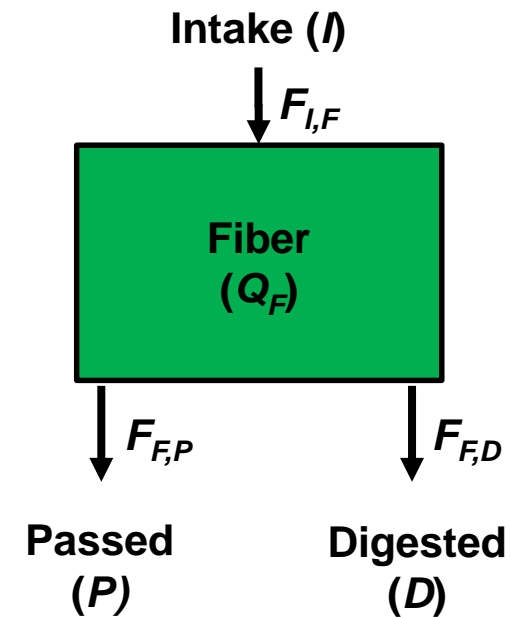
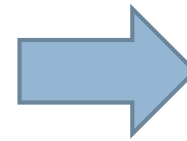
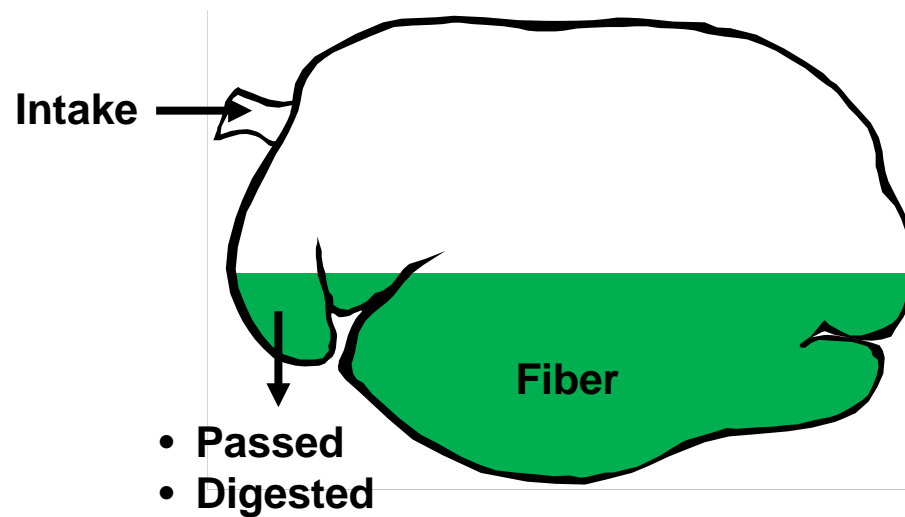
- Organizes model conceptually
  - ▣ Rectangle = state variable (pool)
  - ▣ Arrows = inputs and outputs (fluxes)



# Block diagram

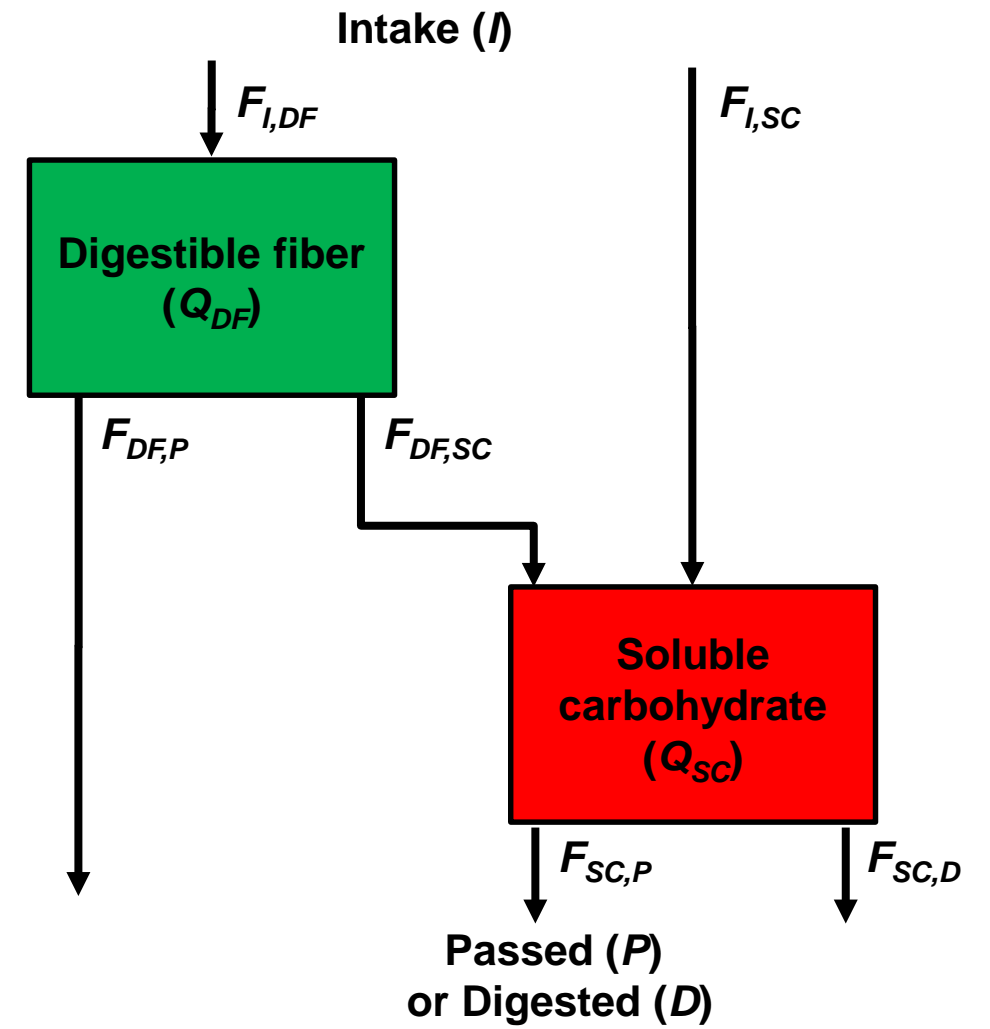
## □ Example

### ▣ Fiber digestion in rumen



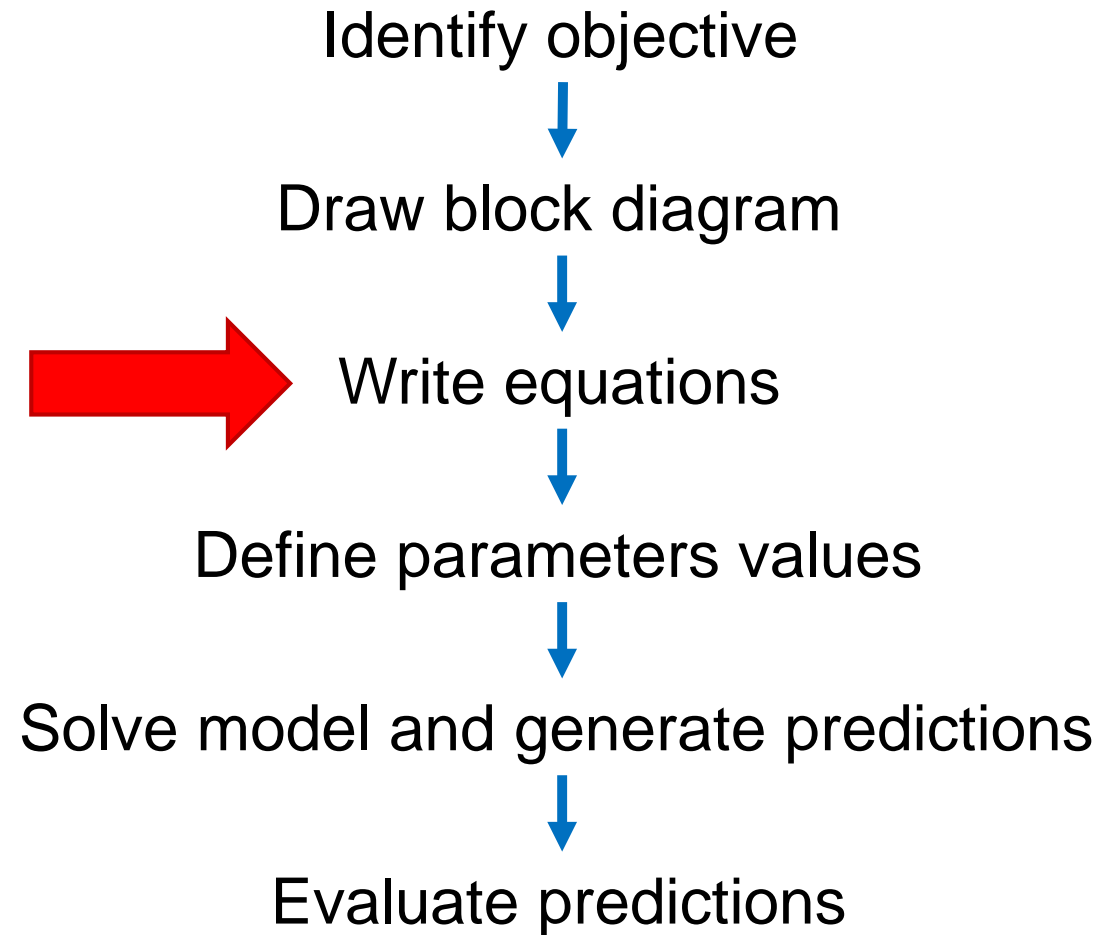
# Block diagram

- Multiple pools connected (usually)



# Steps

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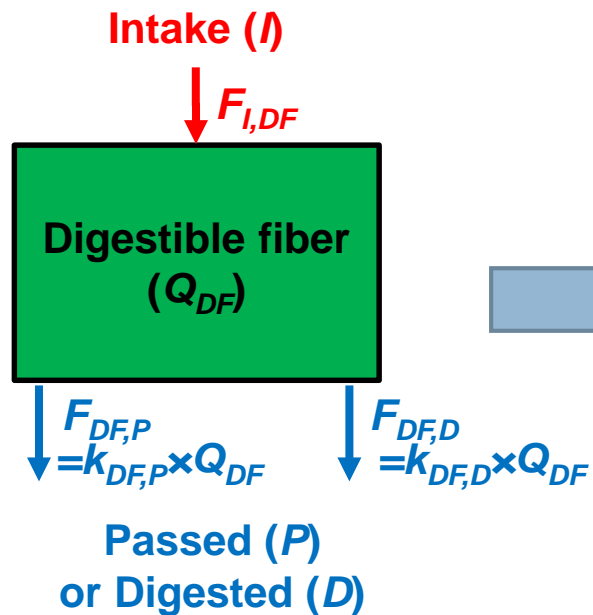
# Differential equations

- Written from block diagram
- Dynamic models
  - ▣ Define change in state variables (pools) over time

$$\frac{d(\textit{State variable})}{dt} = \textit{Inputs} - \textit{Outputs}$$

# Differential equations

## □ Example

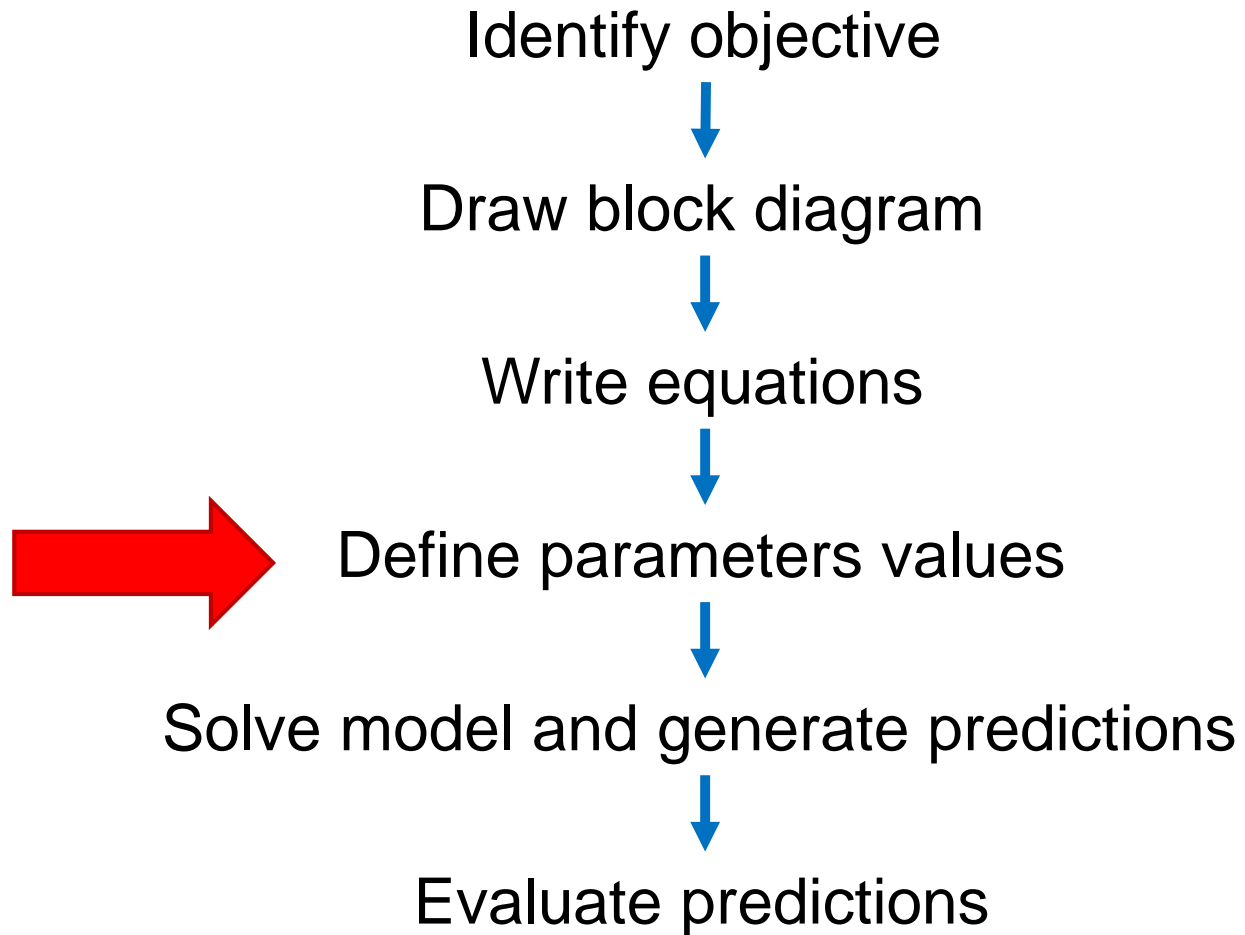


$$\begin{aligned} \frac{d(\text{Fiber})}{dt} &= \text{Intake} - (\text{Passed} + \text{Digested}) \\ \frac{d(Q_{DF})}{dt} &= F_{I,DF} - (F_{FD,P} + F_{DF,D}) \\ &= F_{I,DF} - (k_{DF,P} + k_{DF,D}) \times Q_{DF} \end{aligned}$$



# Steps

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# Define parameter values

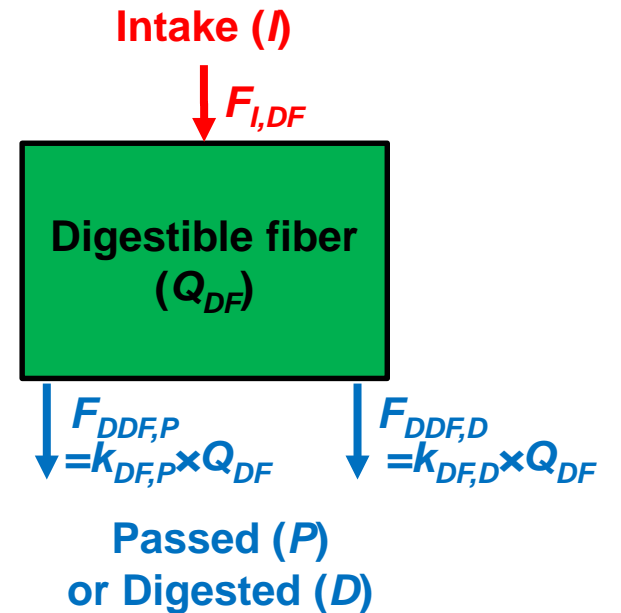
- Measure using experiments

- Typical values

- $F_{I,DF} = 0.30 \text{ kg h}^{-1}$

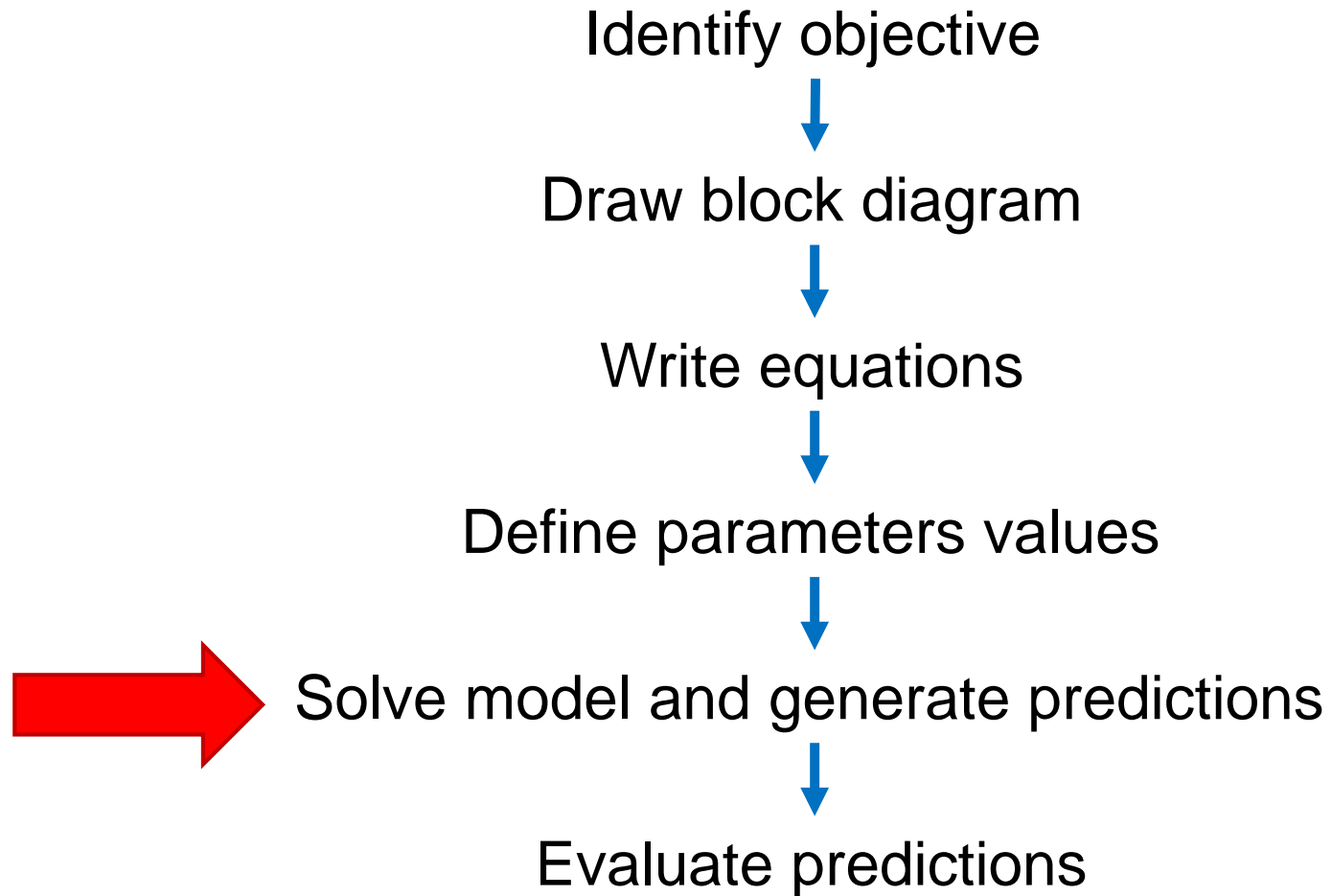
- $k_{DF,P} = 0.05 \text{ h}^{-1}$

- $k_{DF,D} = 0.05 \text{ h}^{-1}$



# Steps

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# Solution

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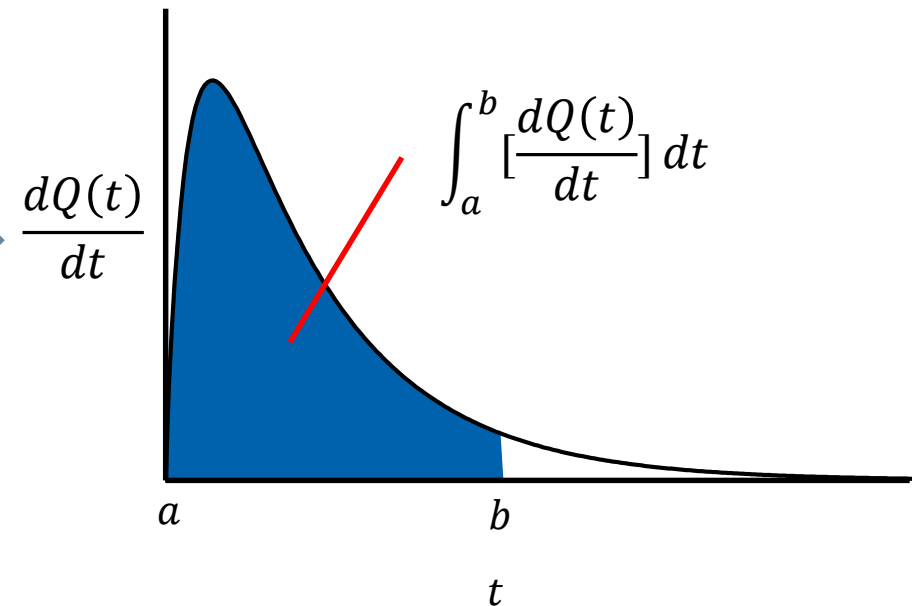
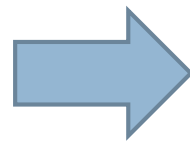
- Equations need to be solved to generate predictions
- Simple models have **analytical solutions**
- Complex models have **numerical solutions only**

# Solution

## □ Analytical solution

- Integrate using rules taught in calculus courses

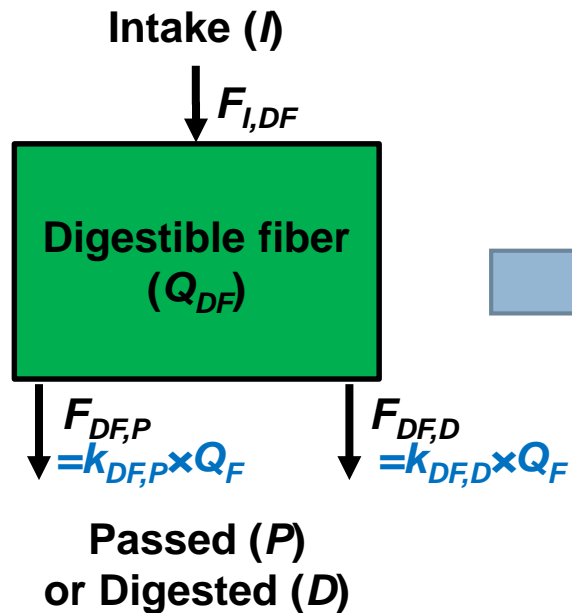
$$Q(t) = \int \left[ \frac{dQ(t)}{dt} \right] dt$$



# Solution

## □ Analytical solution

### □ Example



$$Q_{DF}(t) = \int \left[ \frac{dQ_{DF}(t)}{dt} \right] dt$$
$$= \frac{F_{I,DF}}{k_{DF,P} + k_{DF,D}} + \left[ Q_{DF}(0) - \frac{F_{I,DF}}{k_{DF,P} + k_{DF,D}} \right] \times \exp \left[ - (k_{DF,P} + k_{DF,D}) \times t \right]$$

***Make predictions by evaluating this expression at any time t***

# Solution

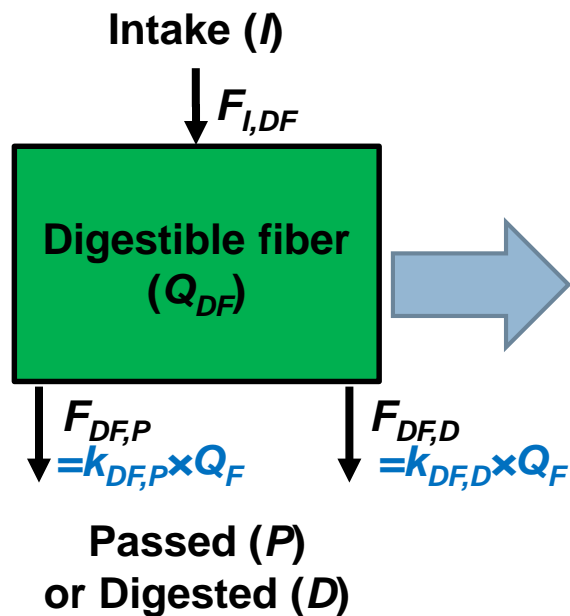
- Numerical solution
  - ▣ Integrate by calculating value numerically over short time intervals ( $\Delta t$ )
  - ▣ Done with **difference equations**

$$Q(t + \Delta t) \approx Q(t) + \frac{dQ(t)}{dt} \times \Delta t$$

# Solution

## □ Numerical solution

### □ Example



$$Q_{DF}(t + \Delta t) \approx Q_{DF}(t) + \frac{dQ_{DF}(t)}{dt} \times \Delta t$$
$$\approx Q_{DF}(t) + [F_{I,DF} - k_{DF,P} \times Q_{DF}(t) - k_{DF,D} \times Q_{DF}(t)] \times \Delta t$$

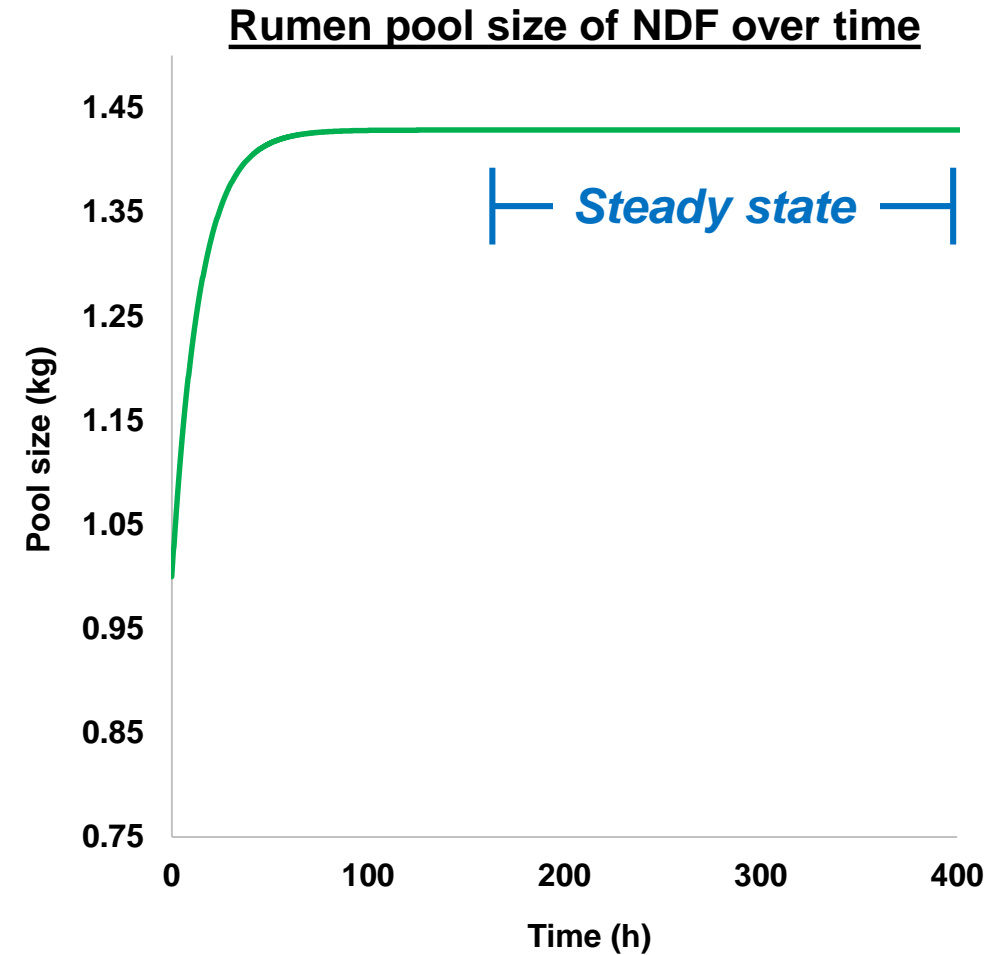
*Make predictions by evaluating this expression iteratively (over time)*



# Solution

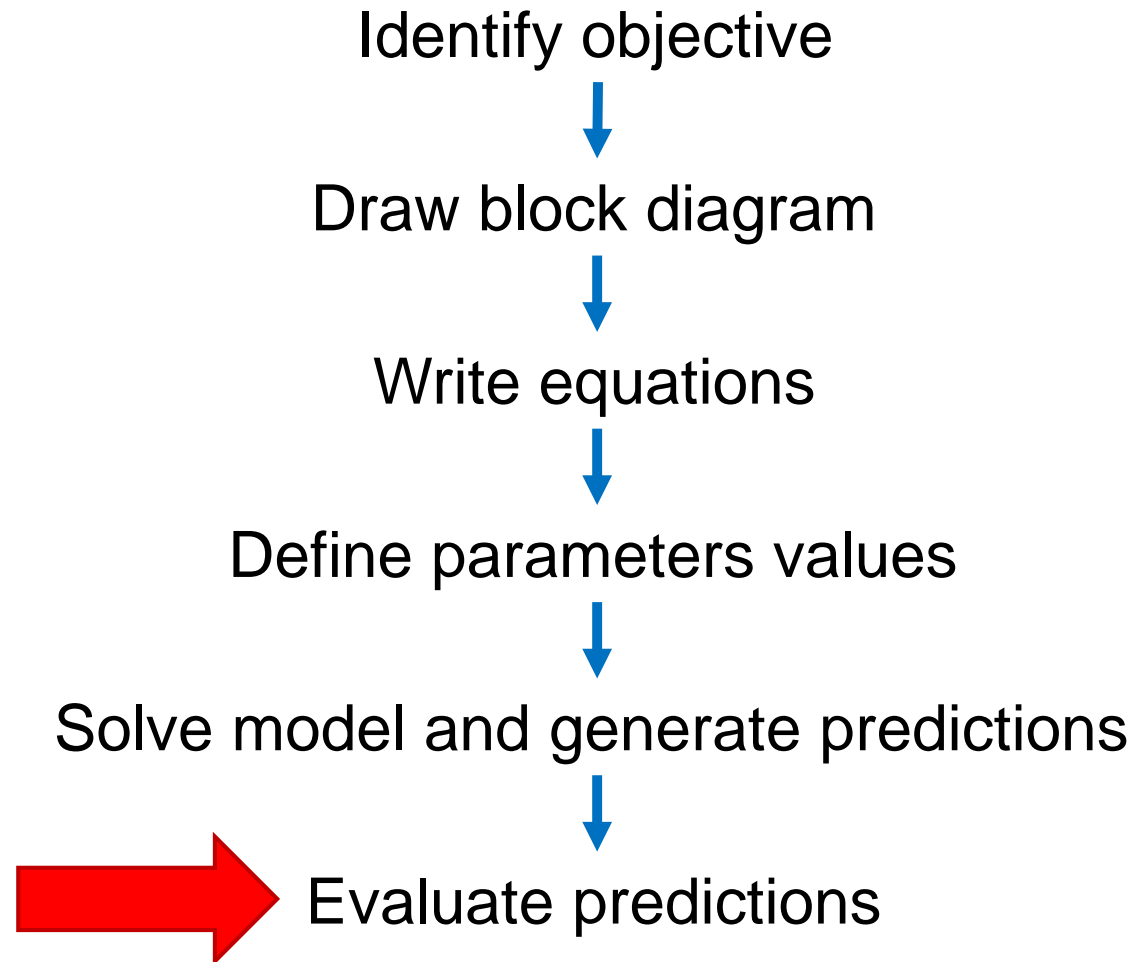
## □ Steady state

- ▣ Reached when value of state variables no longer change
- ▣ Predictions reported for many models are at steady state



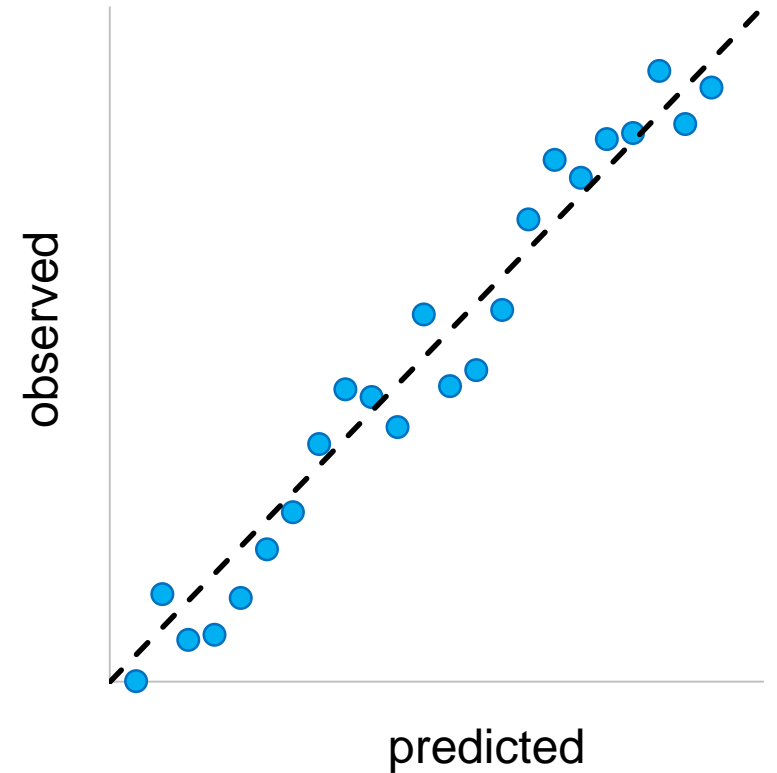
# Steps

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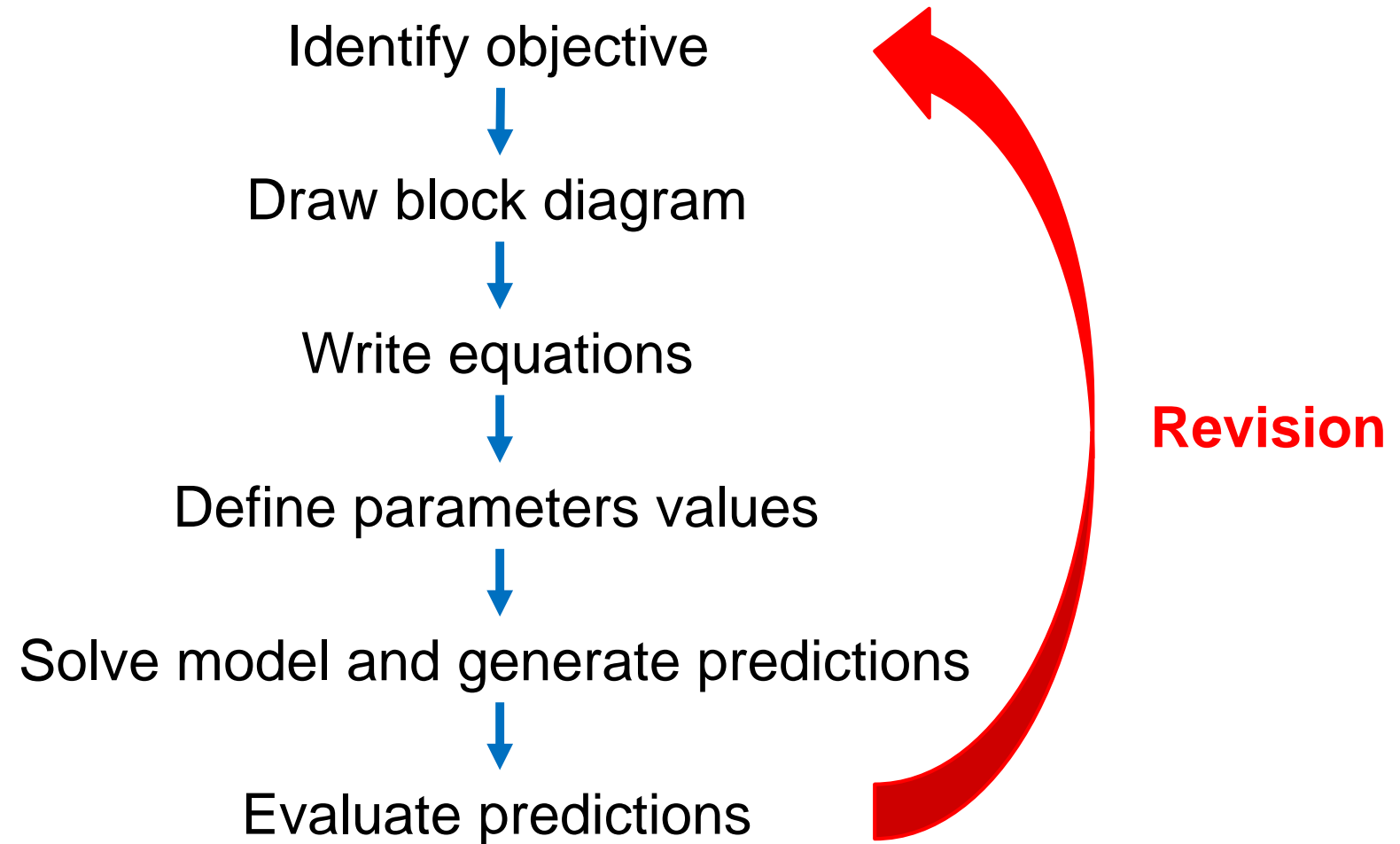


# Evaluation

- Compare how well predictions match reality
  - ▣ Topic of Lesson 2 (Ermias Kebreab)
- Revise model as needed



# Steps



# Revision

*Other things being equal, simpler explanations are generally better than more complex ones.*

William of Ockham

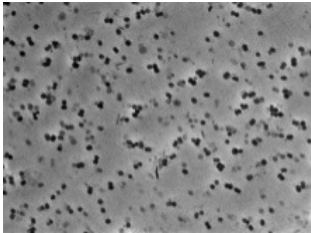


# Revision

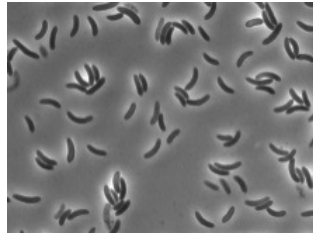
## Baldwin et al. (1970)

Rumen model with three microbial groups

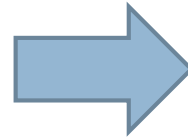
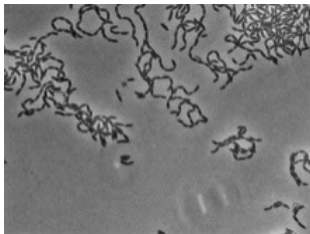
Cellulolytics (Group A)



Amylolytics (Group B)



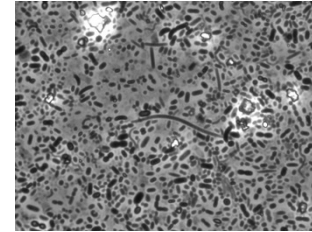
Generalists (Group C)



## Baldwin et al. (1977)

Revision using “one bug” approach

Mixed microbes



# Take home messages

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- Modeling is more than just skill with numbers!
- Modeling starts with an objective
- There are many different types of models, but all are constructed using same steps
- Modeling is a cyclical process, and evaluation of predictions directs revision and further experimentation