

Mathematical Modelling and Scientific Computing in the Biosciences

| 5 June 2007

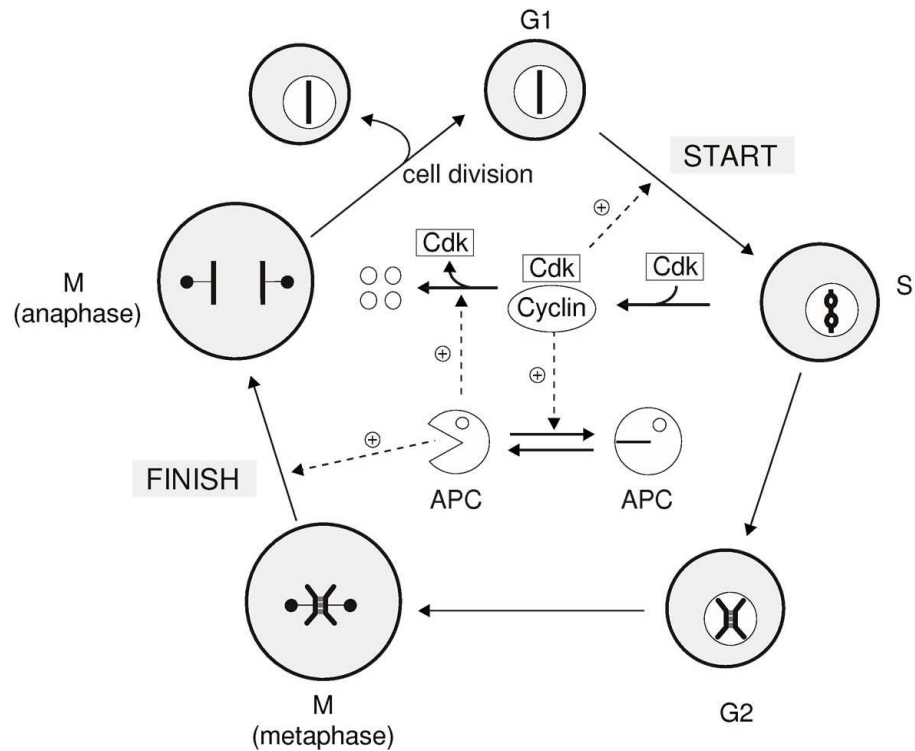
Lecture 8: Overview

- **Cell Cycle Controls**

- **Biological description**
- **Simple mathematical model and analysis**

Cell Cycle

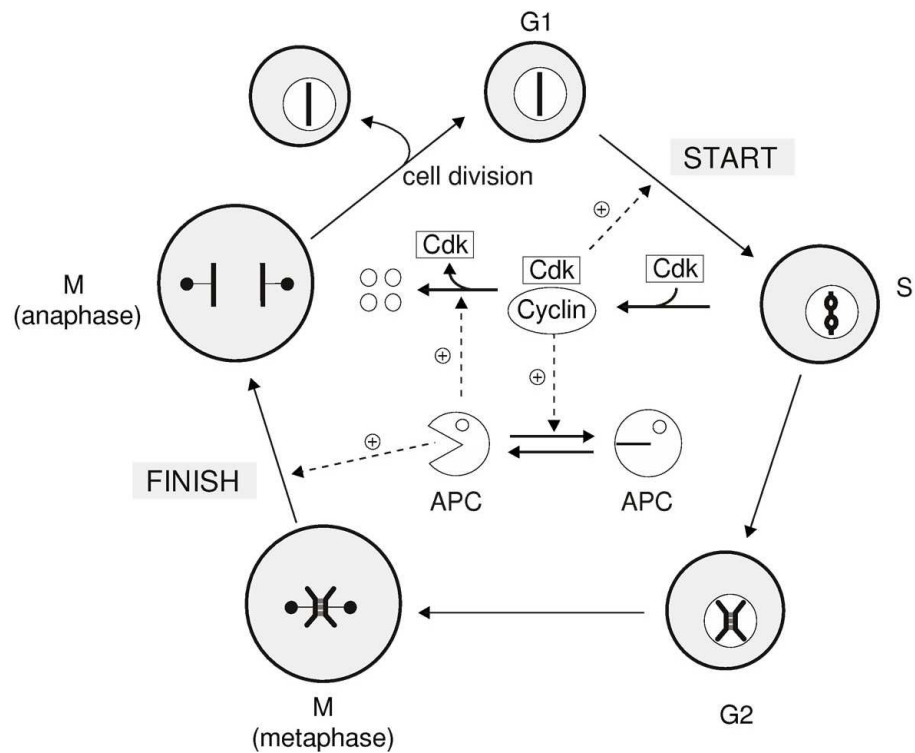
- **Cell cycle**: a sequence of events whereby a cell duplicates its components (including **chromosomes**) and divides into 2 daughter cells
- Complex set of regulation mechanisms underlie cell cycle; cannot be understood just by verbal arguments. Need mathematical modelling and analysis



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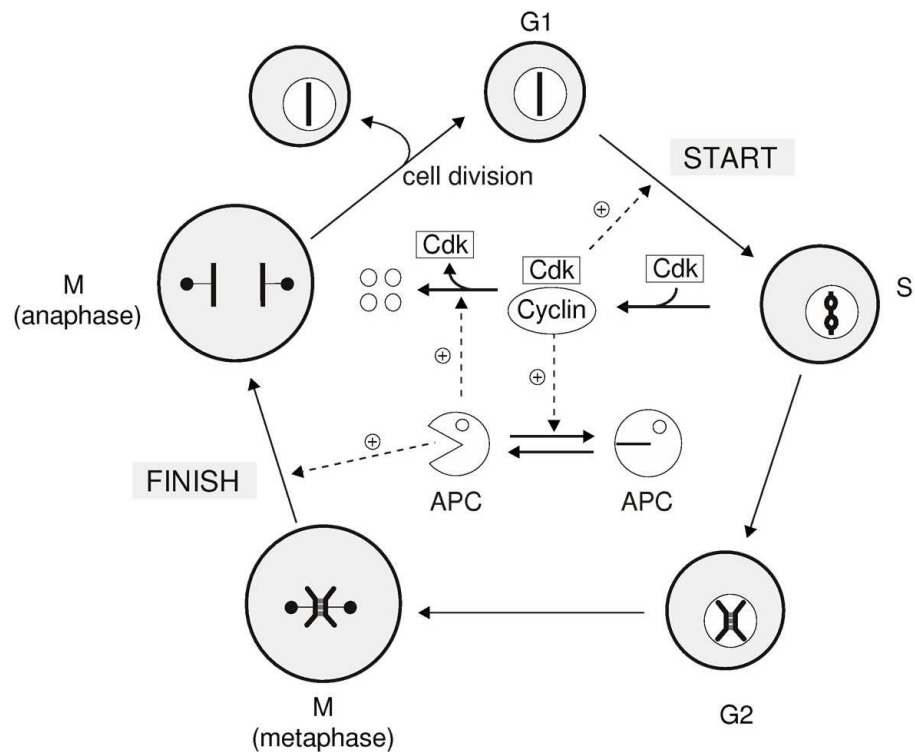
Cell Cycle: Chromosome Cycle

- **Phases of cell cycle:** Start (S), Gap (G1, G2), Mitosis (M)



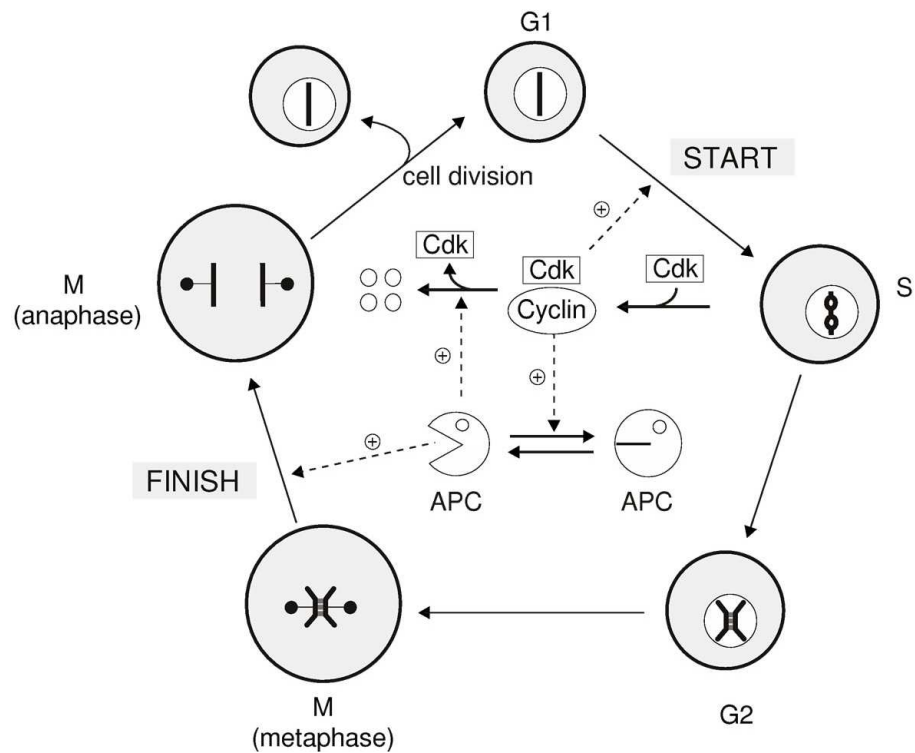
Cell Cycle: Chromosome Cycle

- **G1-S transition:** DNA is replicated. At end of phase, **sister chromatids** are held together by tethering proteins



Cell Cycle: Chromosome Cycle

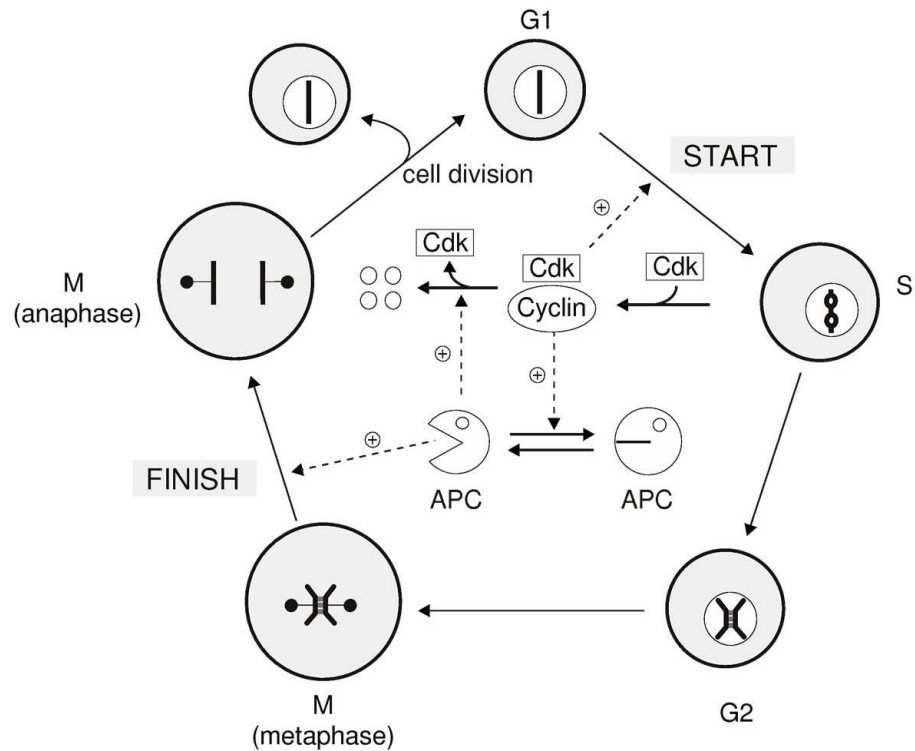
- **Mitosis**: after the **G2** (gap) phase, replicated chromosomes are aligned on the **metaphase spindle**, sister chromatids are attached to opposite poles of the spindle by **microtubules**



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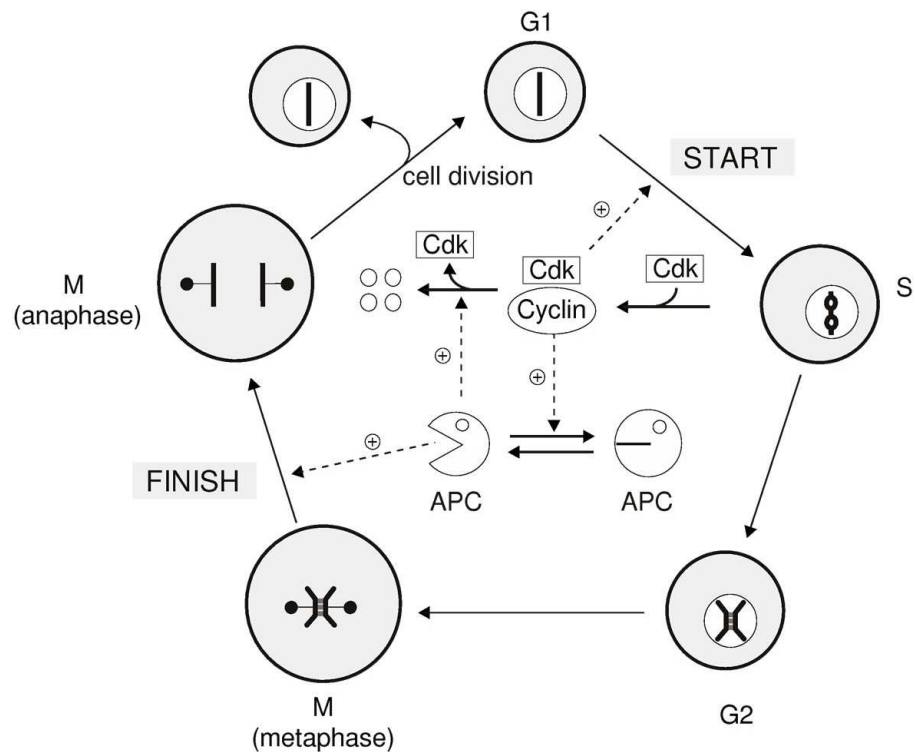
Cell Cycle: Chromosome Cycle

- **Anaphase:** tether proteins are removed so that sister chromatids can be segregated to opposite sides of the cell
- Cell then divides and the process repeats



Cell Cycle: Molecular Machinery

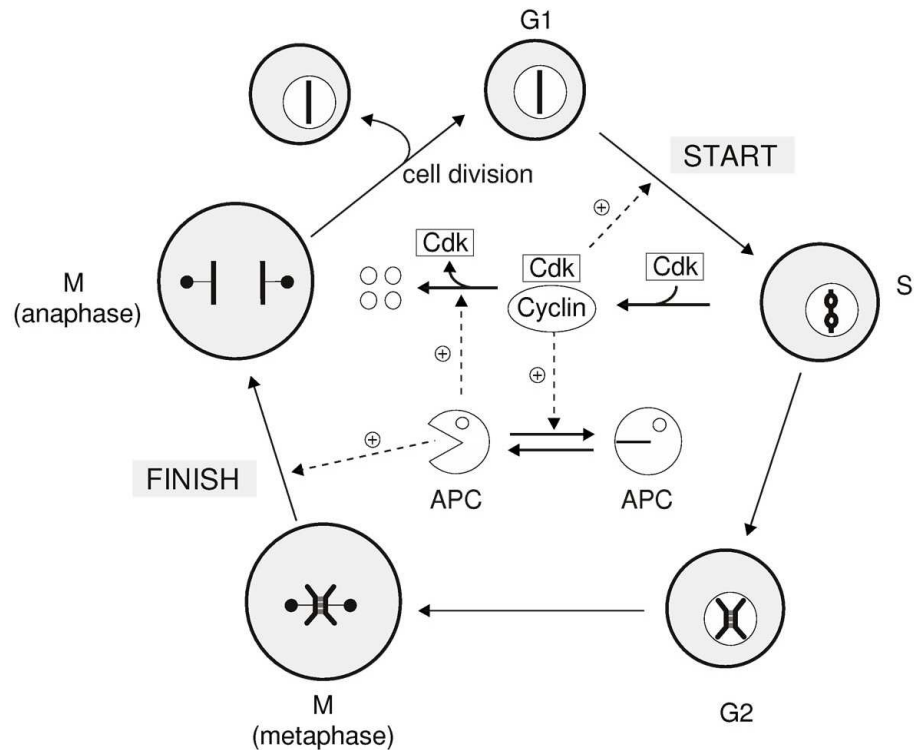
- **Start phase** triggered by the protein, cyclin–dependent kinase (**Cdk**)
- **Cdk** drives the cell through **S**, **G2** phases, up to **metaphase**



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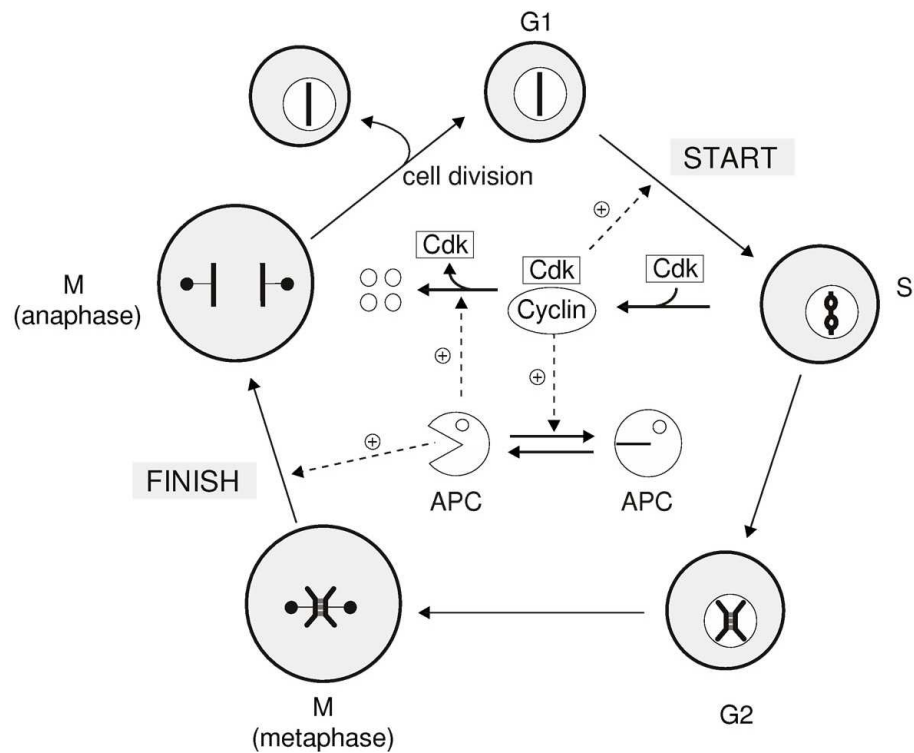
Cell Cycle: Molecular Machinery

- **Finish** is accomplished by the proteolytic machinery, anaphase–promoting complex, (**APC**)
- **APC** destroys tethers and **cyclin** molecules



Cell Cycle: Molecular Machinery

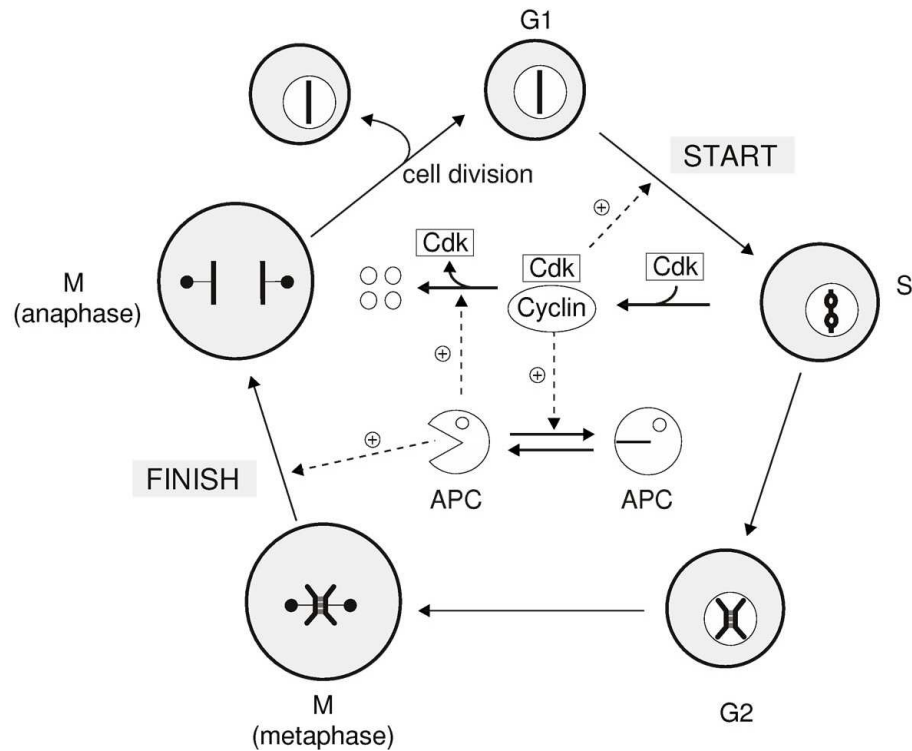
- In the **G1** phase that follows anaphase, **Cdk** is inactive due to the presence of APC which reduces the abundance of **cyclin** molecules



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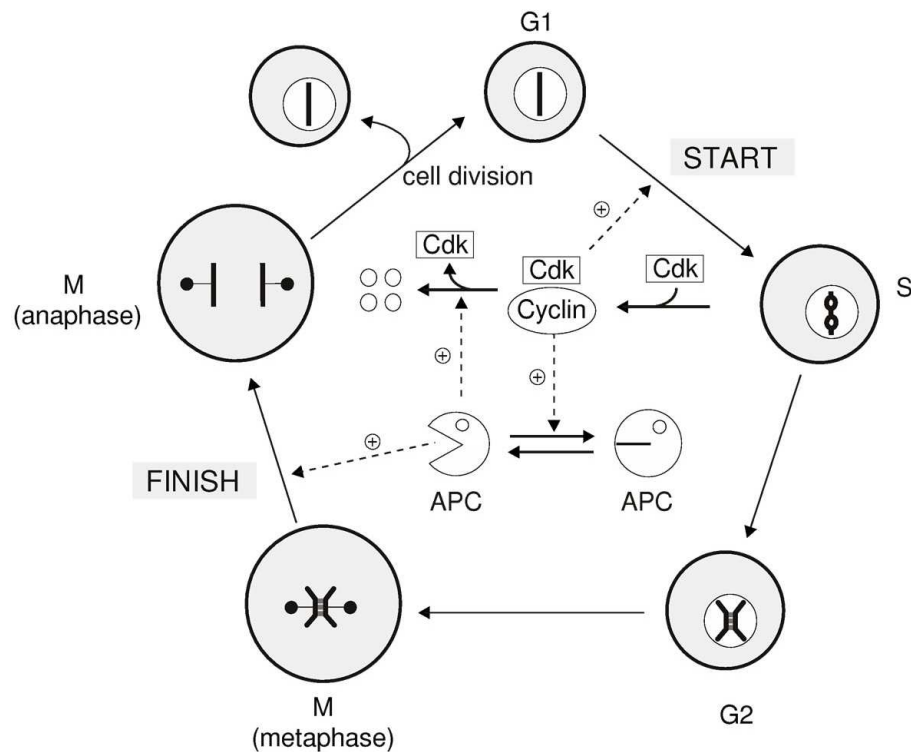
Cell Cycle: Molecular Machinery

- At the subsequent **Start** phase, **APC** must be turned off so that **Cdk** can accumulate
- Antagonistic effects: **APC** destroys cyclins needed for **Cdk**, **Cdk** inactivate **APC** by **phosphorylating** its subunits



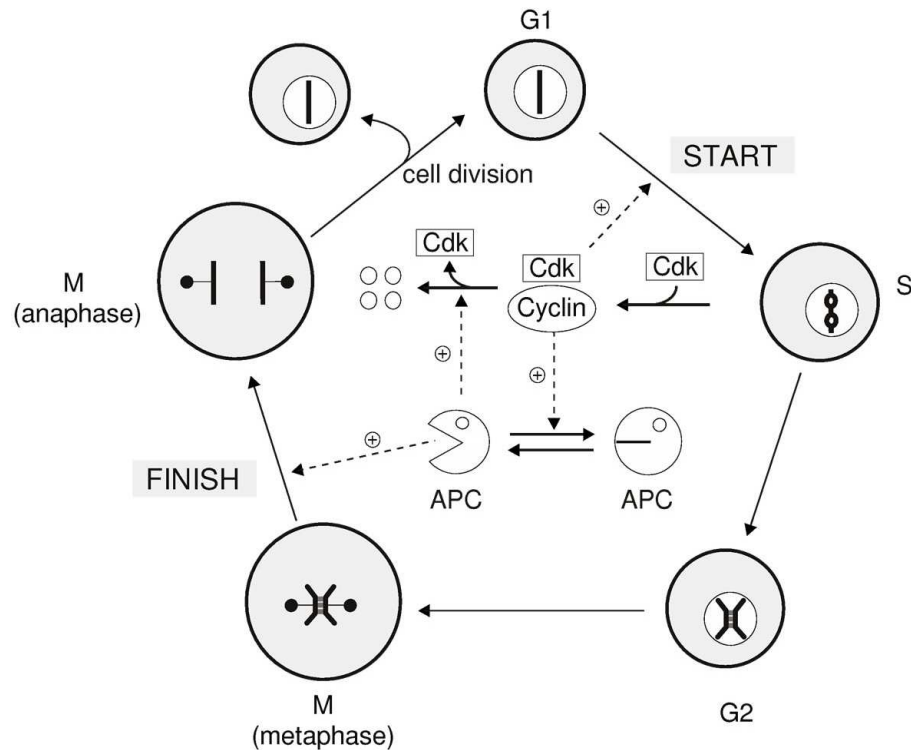
Cell Cycle: Regulation Constraints

- Major transitions of cell cycle must be tightly regulated; e.g.:
 - cell must grow to a critical size before chromosome replication and division
 - if there are problems with DNA replication, the **Finish** transition must be delayed, else it would be a fatal error



Cell Cycle: Regulation Constraints

- Roughly, 2 alternative states: **G1** and **S-G2-M**
- Transition from **G1** to **S-G2-M** is **irreversible**: once DNA synthesis starts, it cannot stop but go to completion
- With respect to molecular machinery, **G1**: **high APC, low Cdk**
S-G2-M: **high Cdk, low APC**

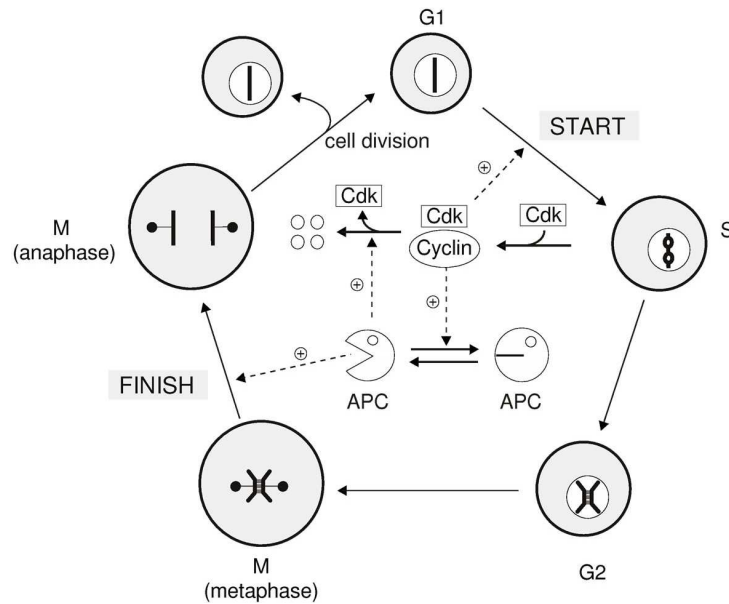


Cell Cycle: Simple Model

- The core biochemical reactions can be described by a pair of nonlinear ODEs:

$$\begin{aligned}
 X'(t) &= k_1 - X(t) (k_2 + k_2p Y(t)) \\
 Y'(t) &= \frac{(k_3 + A k_3p) (1 - Y(t))}{J_3 - Y(t) + 1} - \frac{k_4 m X(t) Y(t)}{J_4 + Y(t)}
 \end{aligned}$$

where X and Y are concentrations of Cdk, APC; m is cell mass, A the concentration of a protein activating APC complex at Finish



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where X and Y are concentrations of **Cdk**, **APC**; m is cell mass, A the concentration of a protein **activating APC** complex at Finish

- Assumptions:

- cyclins are synthesized at a constant rate k_1 in the cytoplasm
- cyclins are degraded at a constant rate, as well as additionally destroyed by **APC** at a rate proportional to **APC** concentration
- **inactive** (phosphorylated) **APC** ($1 - Y(t)$) is turned on to active **APC**, described by Michaelis–Menten kinetics, with Michaelis constant J_3
- **active APC** is phosphorylated (**deactivated**) at a rate proportional to the

mass of cell, m , and depends on the concentration of Cdk in the Michaelis–Menten manner

Cell Cycle: Simple Model

- Dimensional, nonlinear ODEs:

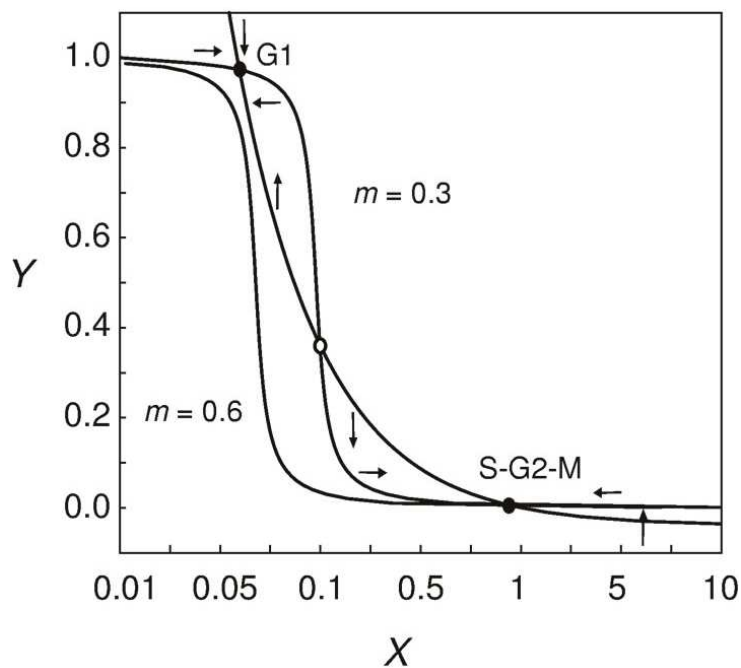
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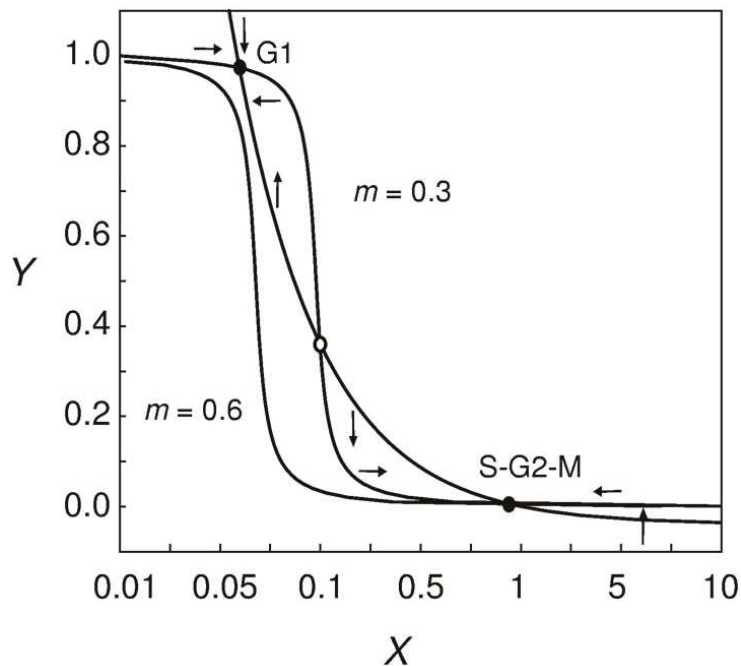
- After non-dimensionalization, obtain null-cline equations:

$$\begin{aligned} X \text{ null-cline : } & X = \frac{\beta}{J_2 + Y} \\ Y \text{ null-cline : } & X = \frac{p(1 - Y)(J_4 + Y)}{(J_3 - Y + 1)Y} \end{aligned}$$

where $\beta = k_1/k_2p$, $J_2 = k_2/k_2p$, $p = (k_3 + k_3p A)/(k_4 m)$



Cell Cycle: Simple Model



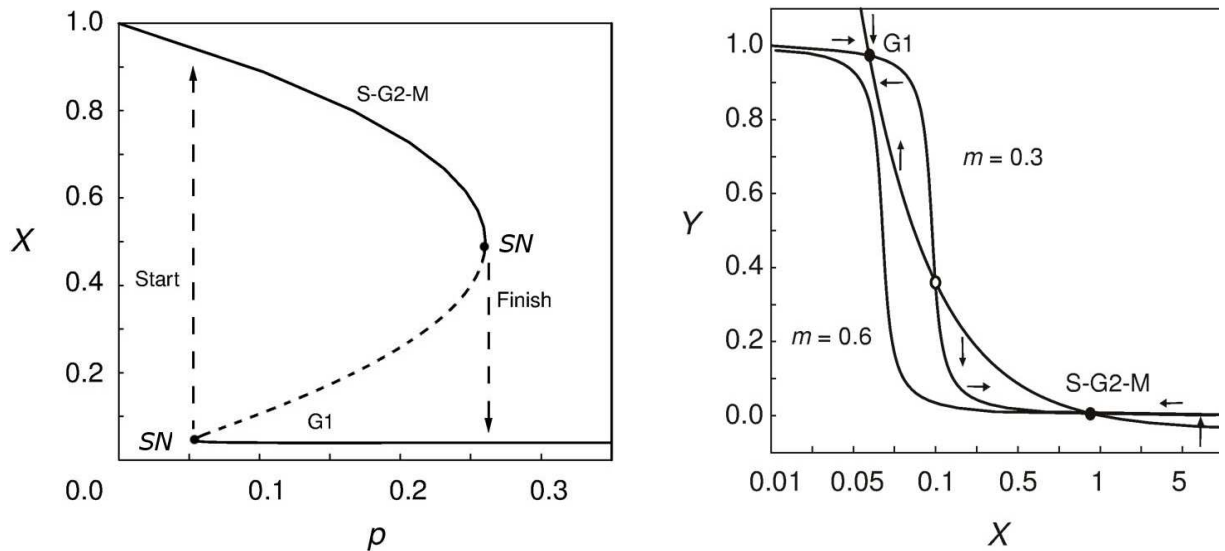
- Interpretations:

- for $m=0.3$, the system has 3 steady states: 2 **stable nodes** (dots), 1 unstable **saddle-node** (circle)
- for $m=0.6$, the system has only 1 steady state: **S-G2-M** point
- hence, as the cell grows, the **G1** point is lost through a **saddle-node bifurcation**, and the system is forced to go to the **S-G2-M** point



Cell Cycle: Simple Model

- How does the solution change as the parameter p changes?
Perform **bifurcation analysis**

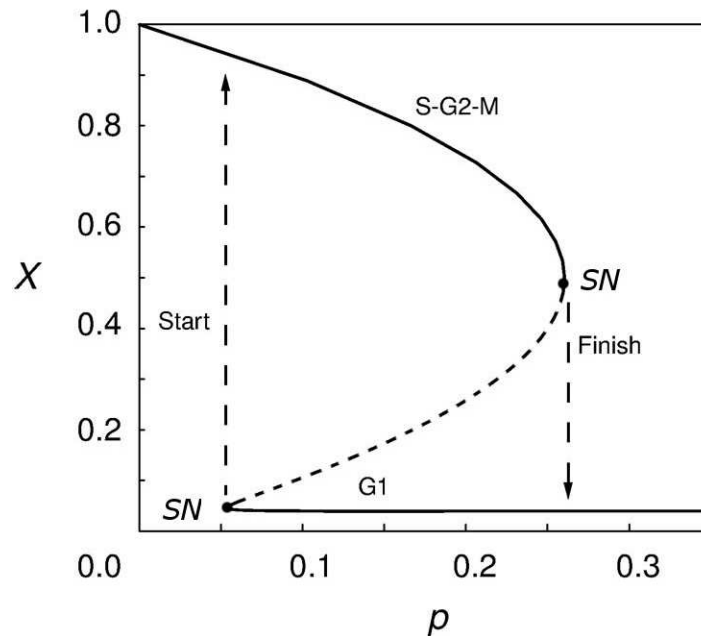


- **Stable**, steady-state solutions coalesce/collide with **unstable** solutions at **saddle-node bifurcation (SN)** points



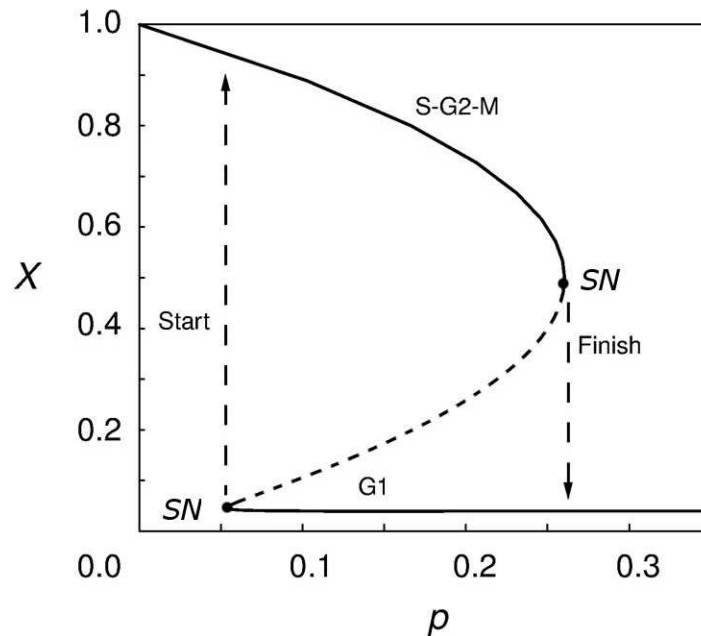
Cell Cycle: Simple Model

- The bifurcation diagram shows that the system exhibits **hysteresis**: i.e., the state depends on the history of the trajectory
- E.g., after the system goes from S-G2-M to G1 by increasing **bifurcation parameter p** , a subsequent decrease in p does not switch the system back to S-G2-M
- Hysteresis effect makes the system less sensitive to fluctuations in signal strengths



Cell Cycle: Simple Model

- Interpretation of bifurcation diagram in biological setting:
 - progress through cell cycle can be seen as a tour through the hysteresis loop
- Small new born cell: $A \approx 0$, hence $p = (k_3 + k_3 p A) / (k_4 - m) \approx k_3 / (k_4 - m)$; so p large, and cell is in the G1 state
- Growth: m increases, so p decreases, until the left saddle-node is reached and X jumps up to the high, S-G2-M state



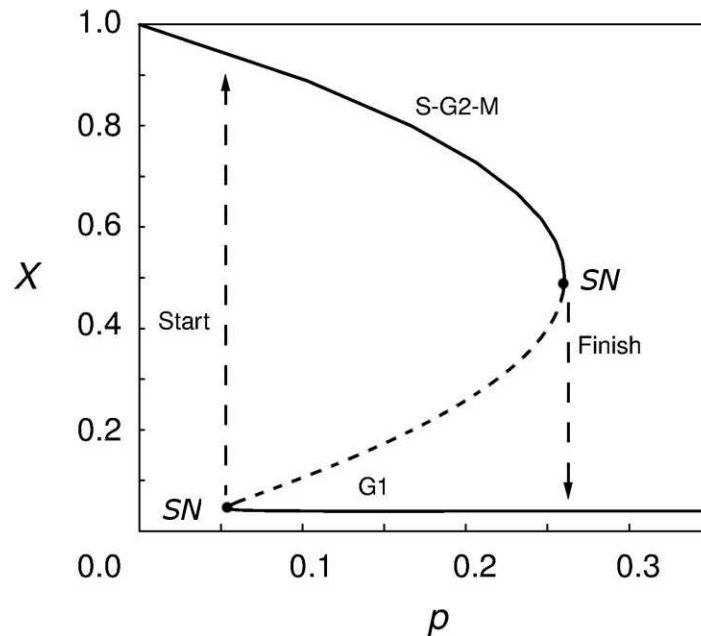
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Cell Cycle: Simple Model

- High X (cyclin) activates process of DNA synthesis and mitosis.
- When DNA replication is complete, the parameter A increases;

hence $p = (k_3 + k_3 p A) / (k_4 m)$ increases, until S–G2–M state is lost via a saddle–node bifurcation

- After the cell jumps irreversibly to G1, the cell divides; $m \rightarrow m/2$, $A \rightarrow 0$ and the whole process starts again



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Cell Cycle: First Extensions of Simple Model

- In the previous model, no mechanism has been given for:
 - increase in parameter A at metaphase to anaphase transition
 - decrease in parameter A in G1 phase
- Simple model for activator A : production is promoted by cyclin X

$$A'(t) = \frac{k5p (m X(t))^n}{J5^n + (m X(t))^n} + k5 - k6 A(t)$$

- Hence, cyclin X , activator A and Y are involved in a long, negative feedback loop:
 - X turns on A , which indirectly activates Y , which destroys X
- Negative feedback: possible to obtain oscillations?

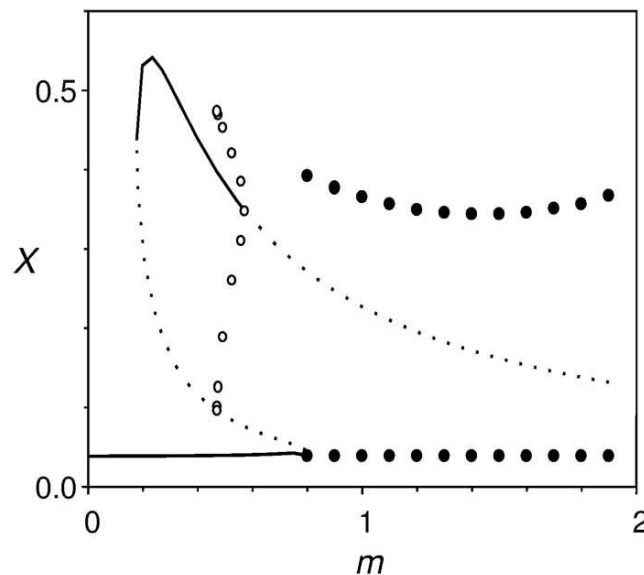
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Cell Cycle: First Extensions of Simple Model

$$X'(t) = k_1 - X(t) (k_2 + k_2 p Y(t))$$

$$Y'(t) = \frac{(k_3 + k_3 p A(t))(1 - Y(t))}{J_3 - Y(t) + 1} - \frac{k_4 m X(t) Y(t)}{J_4 + Y(t)}$$

$$A'(t) = \frac{k_5 p (m X(t))^n}{J_5^n + (m X(t))^n} + k_5 - k_6 A(t)$$



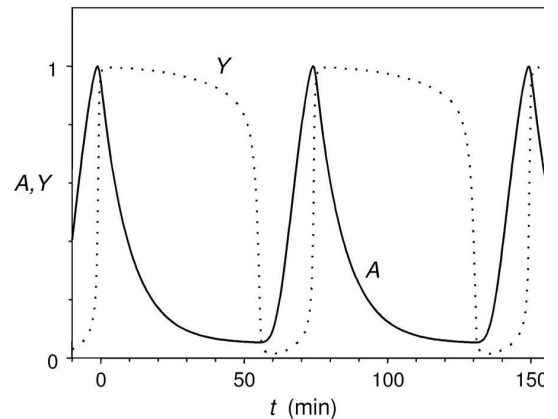
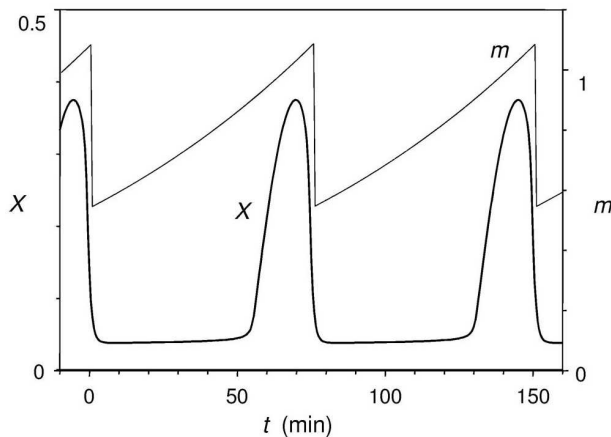
- Analysis shows bifurcation points:
 - saddle–node
 - Hopf point: loss of stable steady–state, appearance of (limit cycle) oscillations



Cell Cycle: First Extensions of Simple Model

- The cell mass was assumed to be an independent parameter; of course, it is time dependent
- Denoting μ to be the growth rate, ms the maximum size of the cell, then the ODE for cell growth is taken to be:

$$m'(t) = \mu m(t) \left(1 - \frac{m(t)}{m_s} \right)$$



- The cell mass grows almost linearly, with halves periodically when cell divides
- **X** initially at low value, until mass m becomes sufficiently large to cross the lower saddle–node bifurcation point
- Subsequently, **X** increases rapidly, causing:
 - **Y** to be degraded rapidly
 - **A** to be produced rapidly, thereby raising p
- When p passes the upper saddle–node, level of **X** drops back to low value

Conclusions

- We have looked at some of the important species participating in controlling cell cycle
- Simple ODE model was constructed, computed bifurcation diagrams, interpreted in biological setting
- Bifurcation diagram captures the qualitative behavior of solutions obtained by doing the full ODE simulation
- Next lectures: bifurcation analysis
 - What are the generic bifurcations that can occur?
 - Matlab software, [MATCONT](#)
 - Algorithms for [solution continuation](#) and [bifurcation detection](#)