

Welcome to STAT 302, *Intro to Probability*

Instructor: Alexandre Bouchard
Fall 2014

Plan for today:

- Logistics.
- Why you should care about probability.

To get this information & more

- Main website :
<http://www.stat.ubc.ca/~bouchard/courses/stat302-fa2014-15/>
- Piazza (contact link on webpage)
- Click on 'Files' to get lecture slides, assignments, etc.

Administrative details

- Prerequisites: Math 200 or 226 (which may be taken concurrently)
- Exclusions: Stat 241/251, Math 302
- Textbook: *A first course in probability* Sheldon Ross
 - 9th edition recommended
 - Suggested, optional exercises posted on website

Assessment

- Assignments (4), 20%
- We encourage you to discuss your work with other students...
- However, you must write up your own solutions independently.
- No extension possible!

Assessment

- Midterm, 25%
- Date: Wed Oct 22 (check this week that this date is not a problem)
- No make-up exam!

Assessment

- Webwork/clickers, 5%
- Get your clicker this week! (I will start using clicker questions by Monday Sep 8)
- More on this soon
- Final, 50%
- After lectures are over
- Date announced centrally by University
- You must pass the final to pass the course

To get help

- Great TAs:

Sean Jewell

sean.jewell@stat.ubc.ca

Vincent Huang

finial79@hotmail.com

- My email: bouchard@stat.ubc.ca
- But remember: use piazza (unless it is a personal matter)
- Office hours: TBA, will create a Doodle on main website shortly

What STAT 302 is about:

- *Probability spaces*: arguably the best quantitative tool to model reality
- Properties of probability spaces
- Lots of examples

Why this topic is important

- Fundamental tool in statistics, computer science, physics, econometrics, ... and increasingly, biology, linguistics, sociology, ...
- Creating models
- Inverting them (Bayesian statistics/conditioning)
- Computational power of randomness
- Also a branch of pure math in its own right
- Replacing logic as the philosophical foundations of science and cognition?
(‘Dawning of the age of stochasticity’, D. Mumford)

Probability in action: Diverse examples

Engineering, technology, logistics

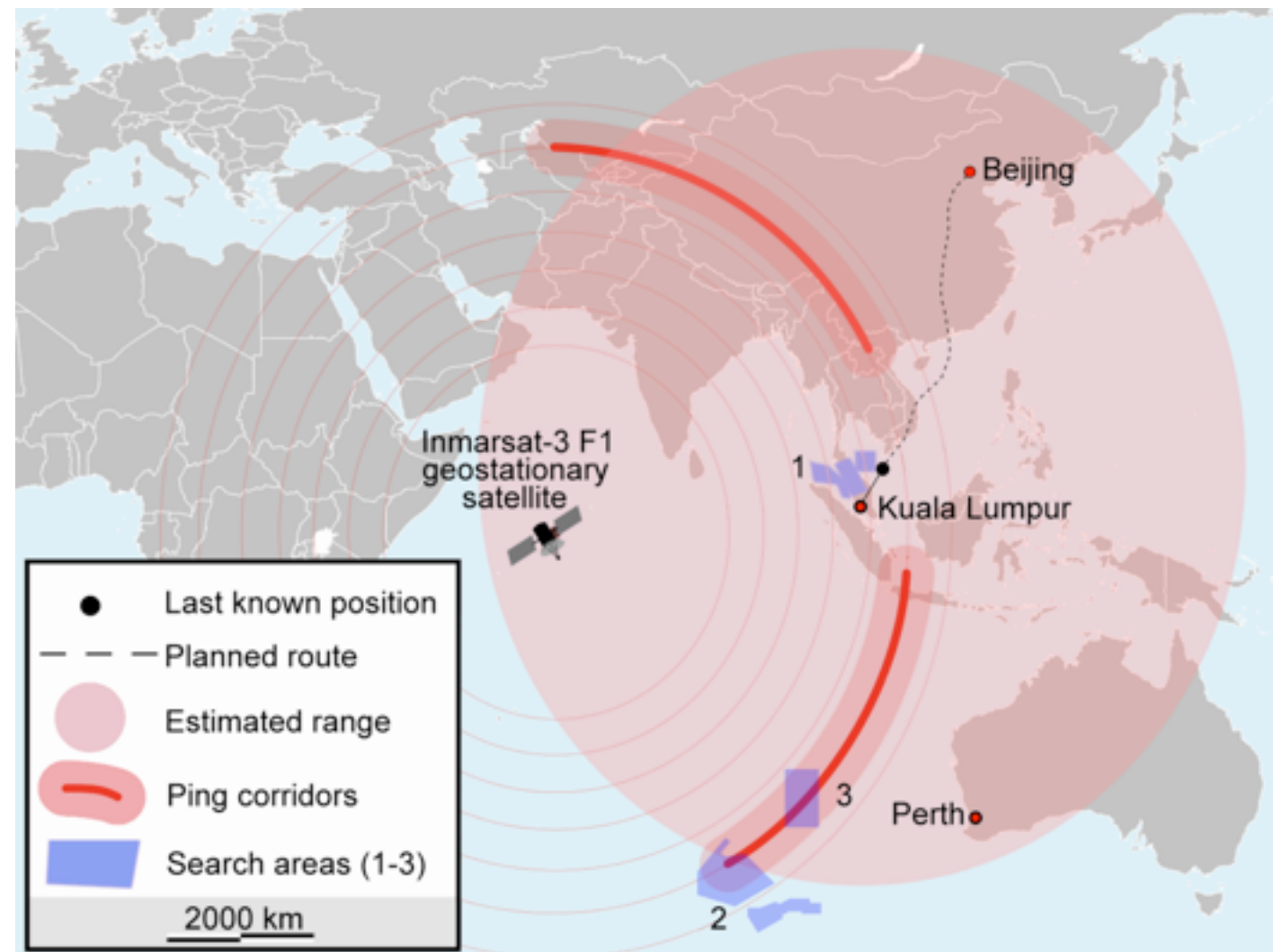
Ex. I The Search for Malaysia Airlines Flight 370

Goal: finding the location of the crash

Question: how to prioritize search

How to reconcile several sources of partial info:

- Last known position
- Fuel range
- Last satellite ping



<http://tinyurl.com/lhzrufa>

Ex. I

Bayesian Search



1966: Palomares B-52 crash

Photo # NH 97221-KN Stern section of sunken USS Scorpion, 1986



1968: USS Scorpion
disappearance

Conditioning

- Say you search in the square of highest success probability
- You find nothing
- What should you do next?
 - Note: even if the submarine is there, you might have missed it!
- Probability as a calculus of belief and uncertainty

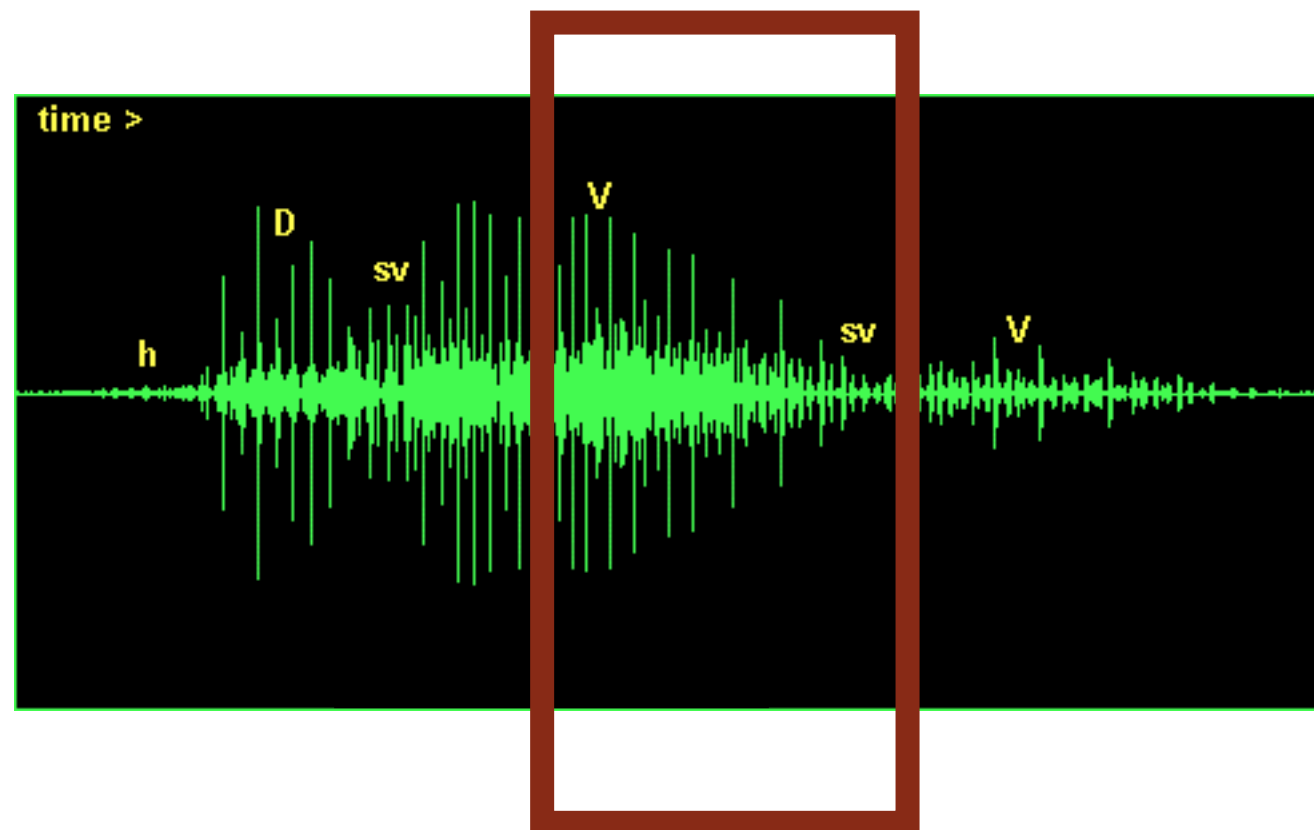


Bayes theorem
(Thomas Bayes),
1763

<http://tinyurl.com/pcznhtml>

Ex. 2

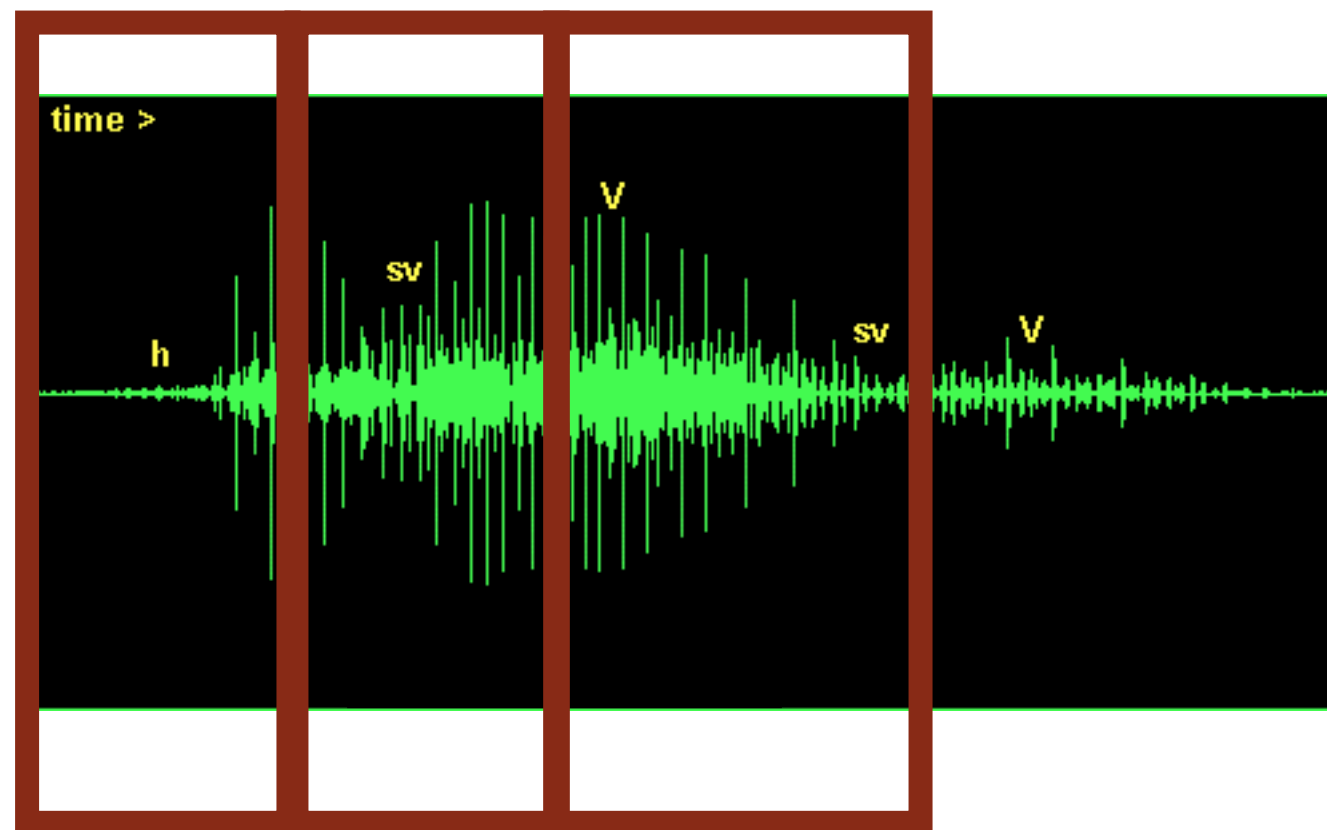
AI and machine learning: Speech recognition



???

Ex. 2

AI and machine learning: Speech recognition



How are ???

Ex. 3

Rational behavior and uncertainty

General question: how to **act** when

- we are facing uncertainty
- errors have different costs

Examples:

- fraud detection
- medical diagnosis
- spam classifiers

Key tool: *expected value*

Sciences

Ex. 4

Ecology: Estimating animal population sizes

Example: finding
the number of
Sockeye salmon in
the Pacific Ocean (!)

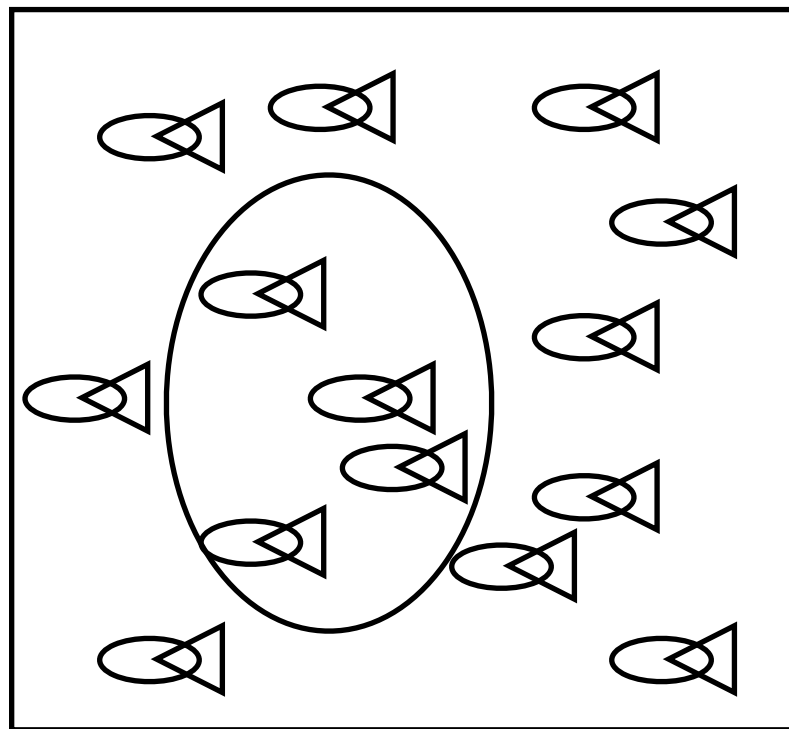
Very important
problem for
conservation, setting
fishing quotas, etc.



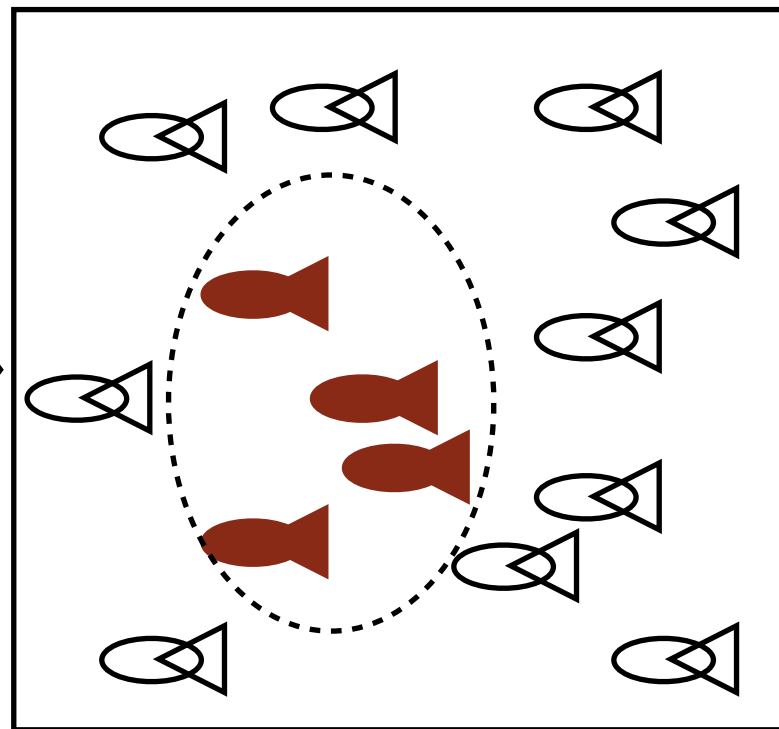
Ex. 4

Insight: the capture-recapture trick

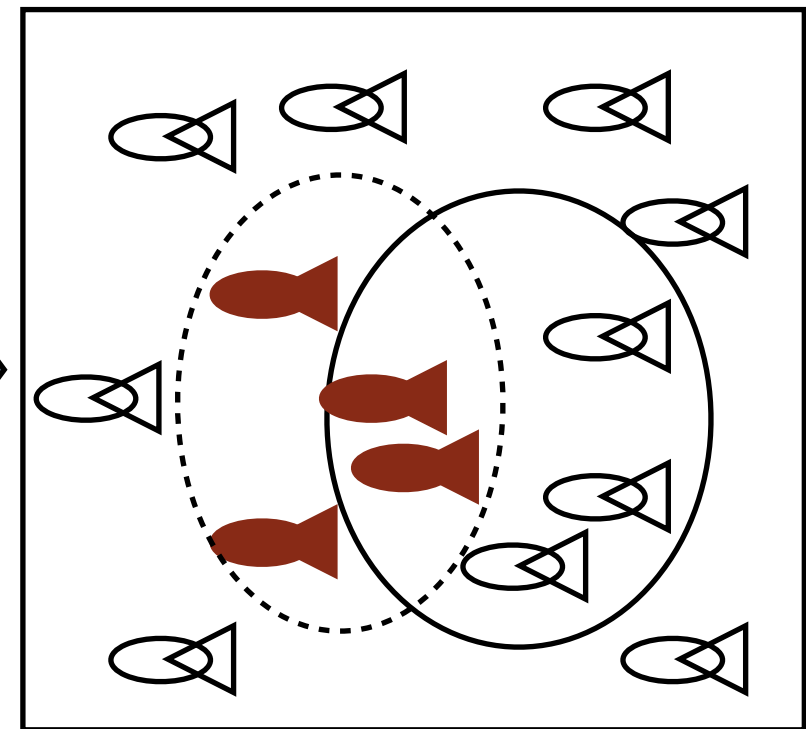
Population



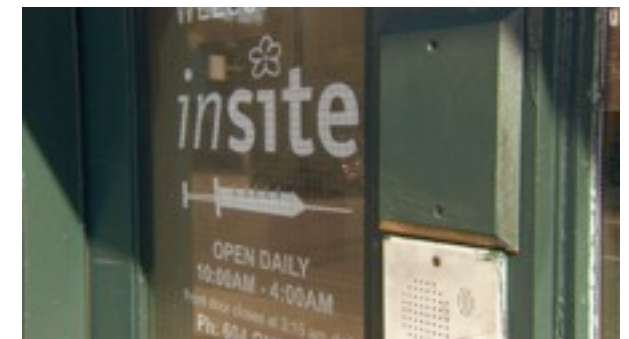
Capture and tag



Recapture and count

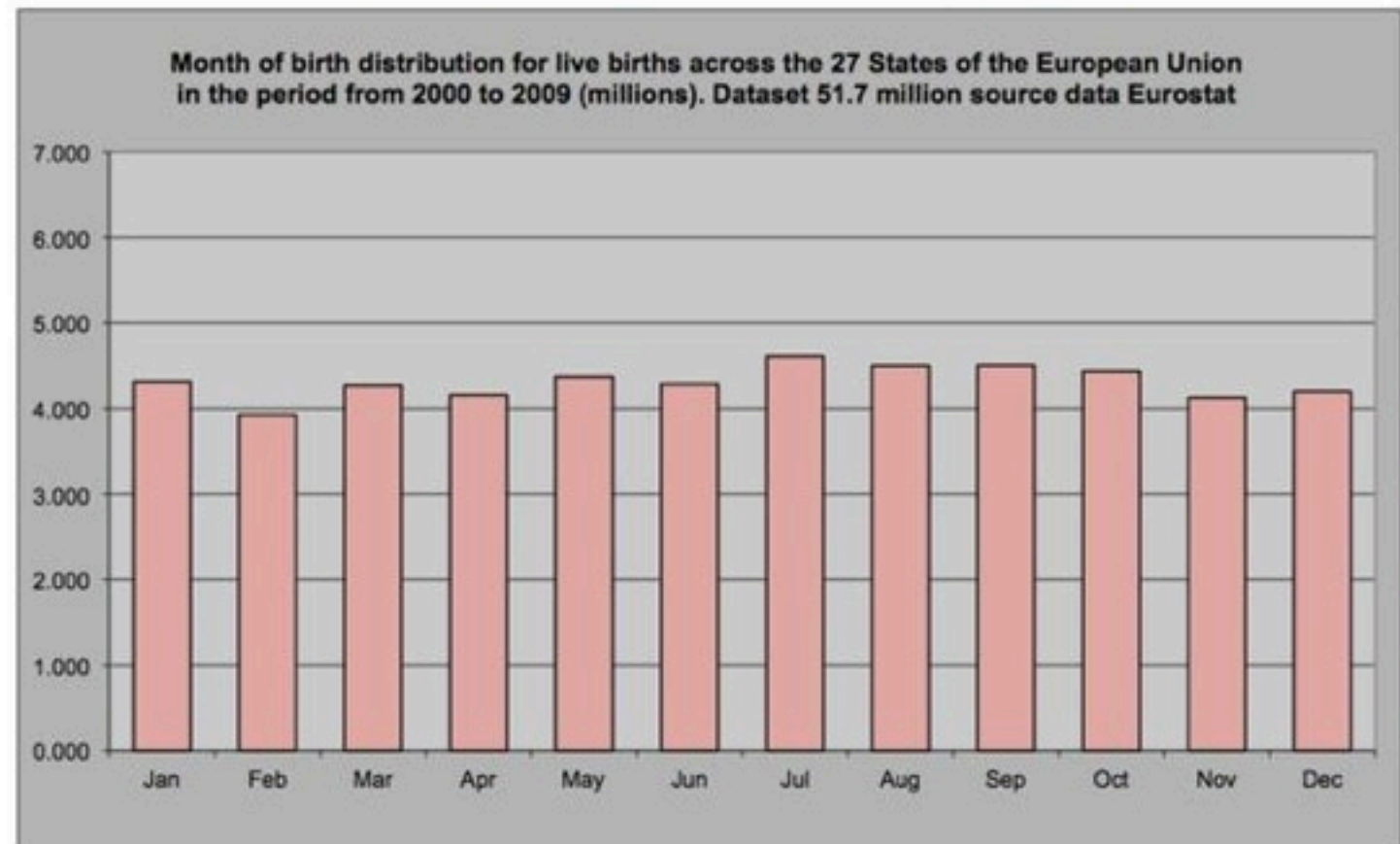


Examples:



Assessing *significance*

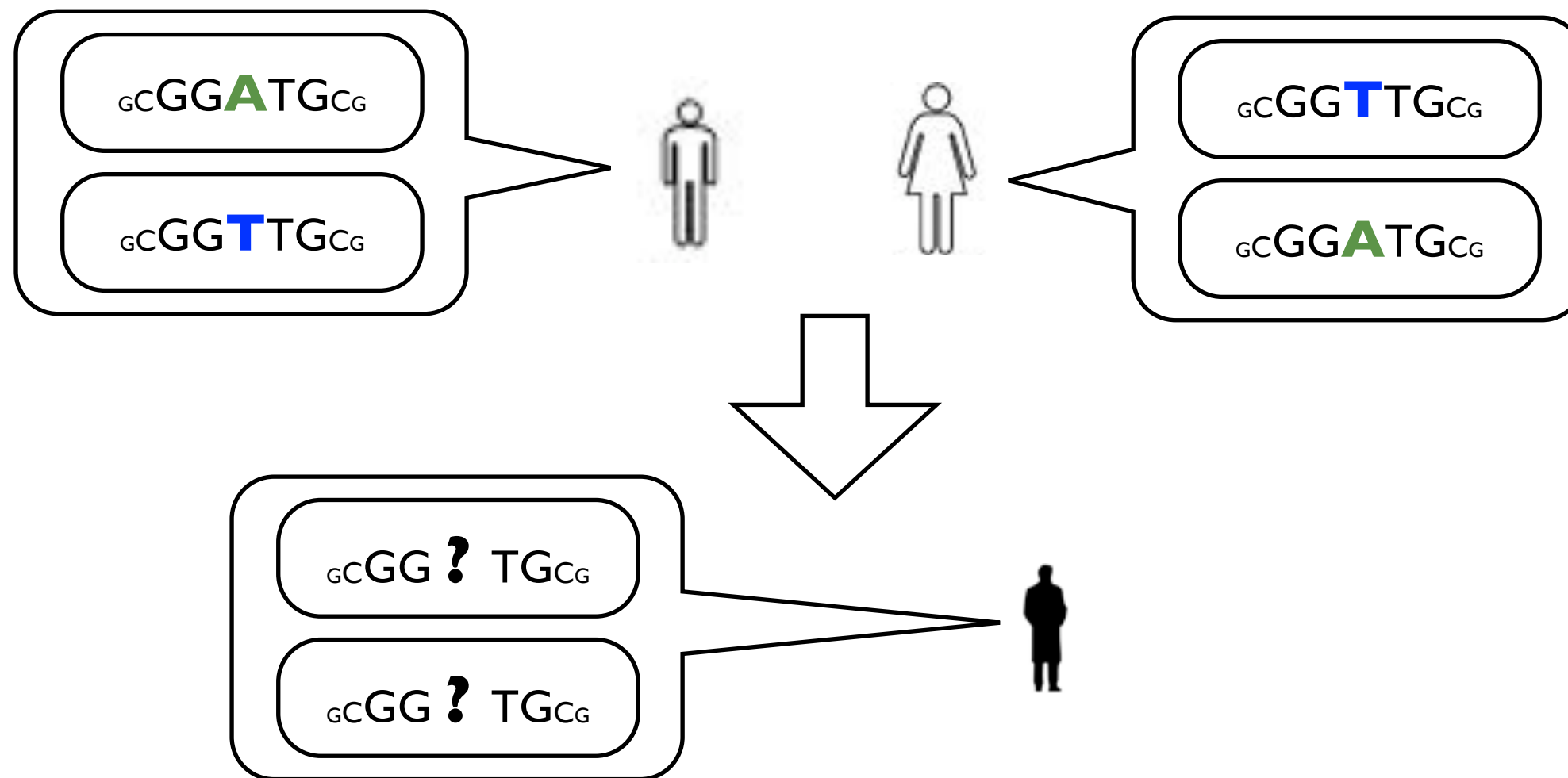
- *Histogram* of births organized by month:
- Question: in *general*, are there months where there are more births than others? (are births *uniform* across months?)
- Note: even if the answer is no, we would expect small differences across months.
 - How small?



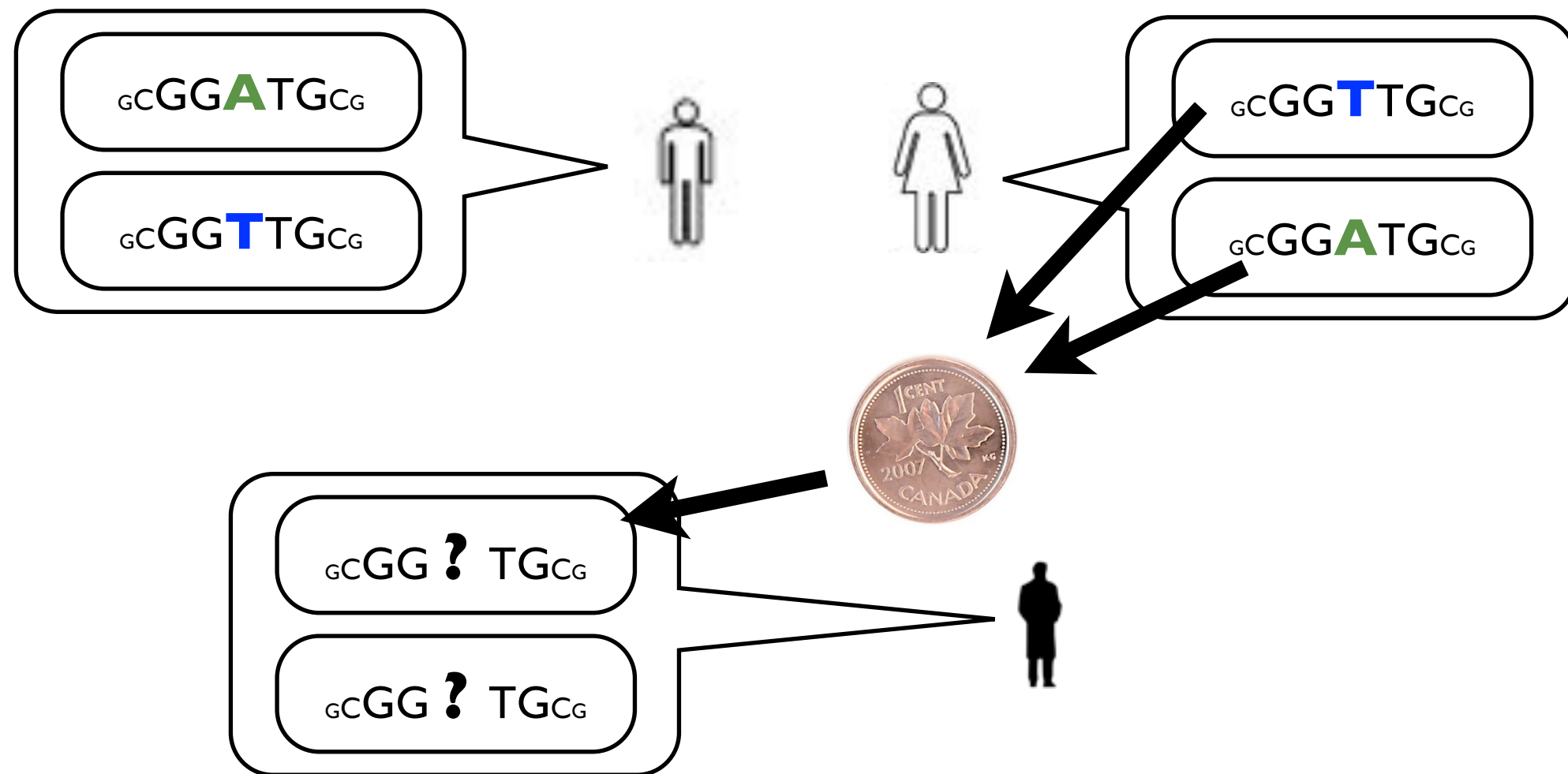
Ex. 6

Genetics: inheritable diseases

Randomness in inheritance :

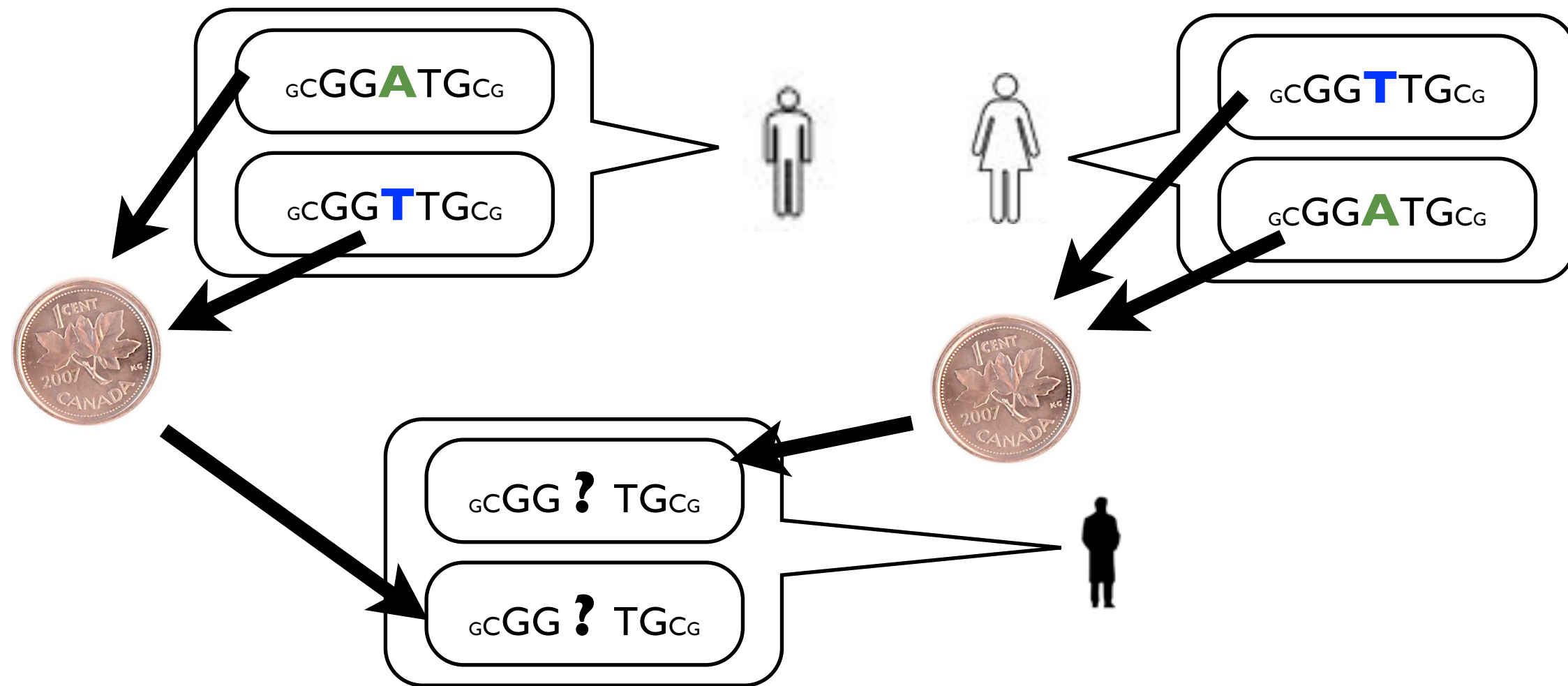


Genotype inheritance



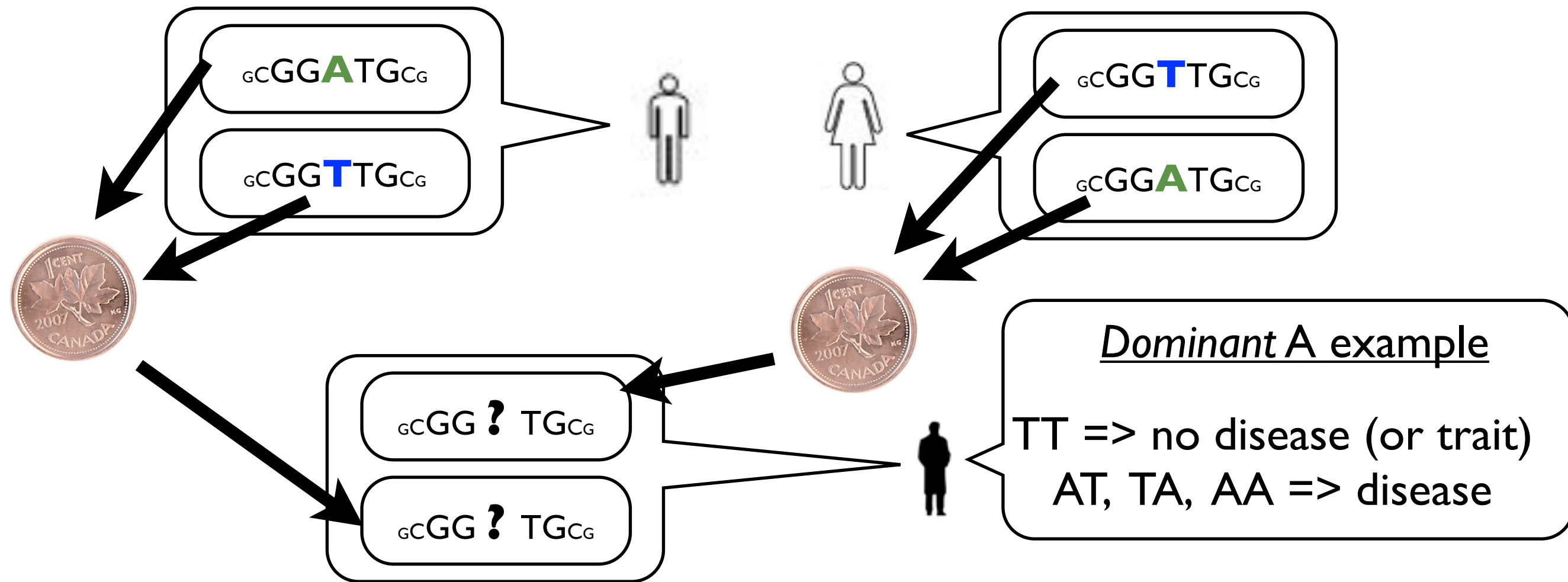
I) Flip a fair coin to decide if you inherit mom's T or A

Genotype inheritance



- 1) Flip a fair coin to decide if you inherit mom's T or A
- 2) Flip another fair coin to decide if you inherit dad's T or A

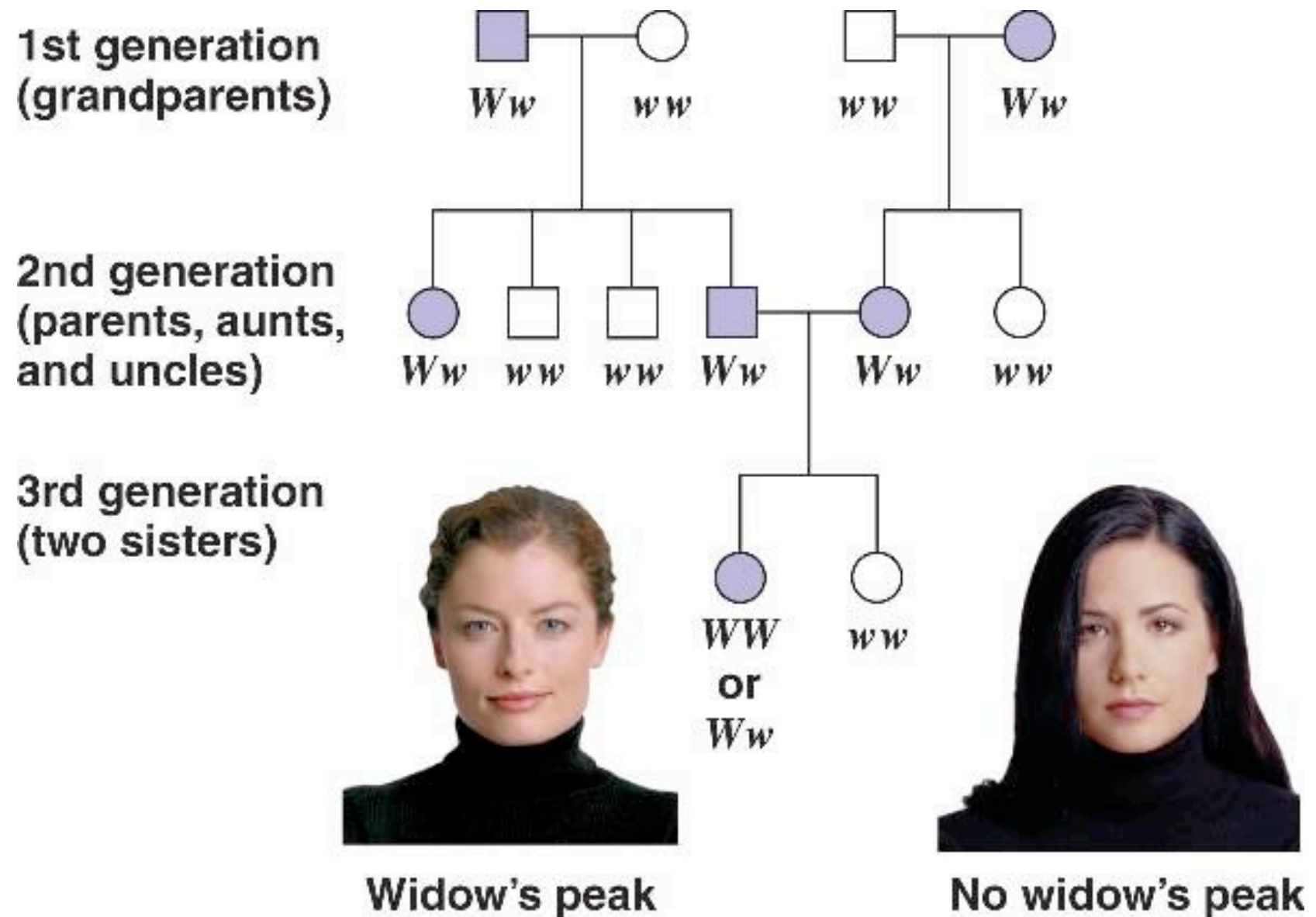
Genotype inheritance



- 1) Flip a fair coin to decide if you inherit mom's T or A
- 2) Flip another fair coin to decide if you inherit dad's T or A

Larger family trees

- A larger example where W is dominant over w
- Goals:
 - genetic counseling
 - finding genetic factor of diseases / traits
- Complication factor
 - incomplete data

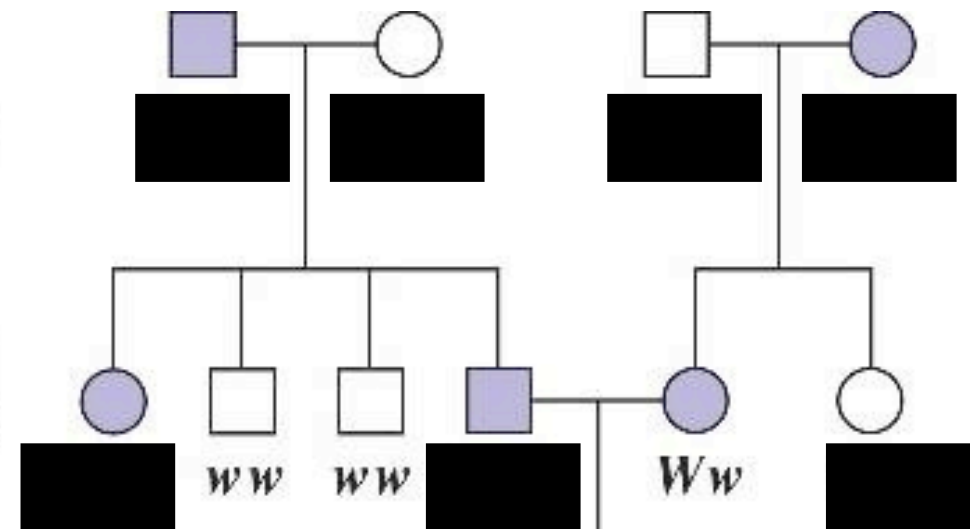


Ex. 6

Larger family trees

- A larger example where W is dominant over w
- Goals:
 - genetic counseling
 - finding genetic factor of diseases / traits
- Complication factor
 - incomplete data

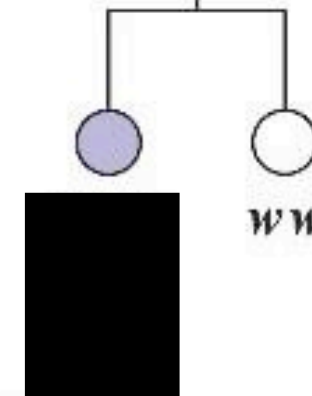
1st generation
(grandparents)



3rd generation
(two sisters)



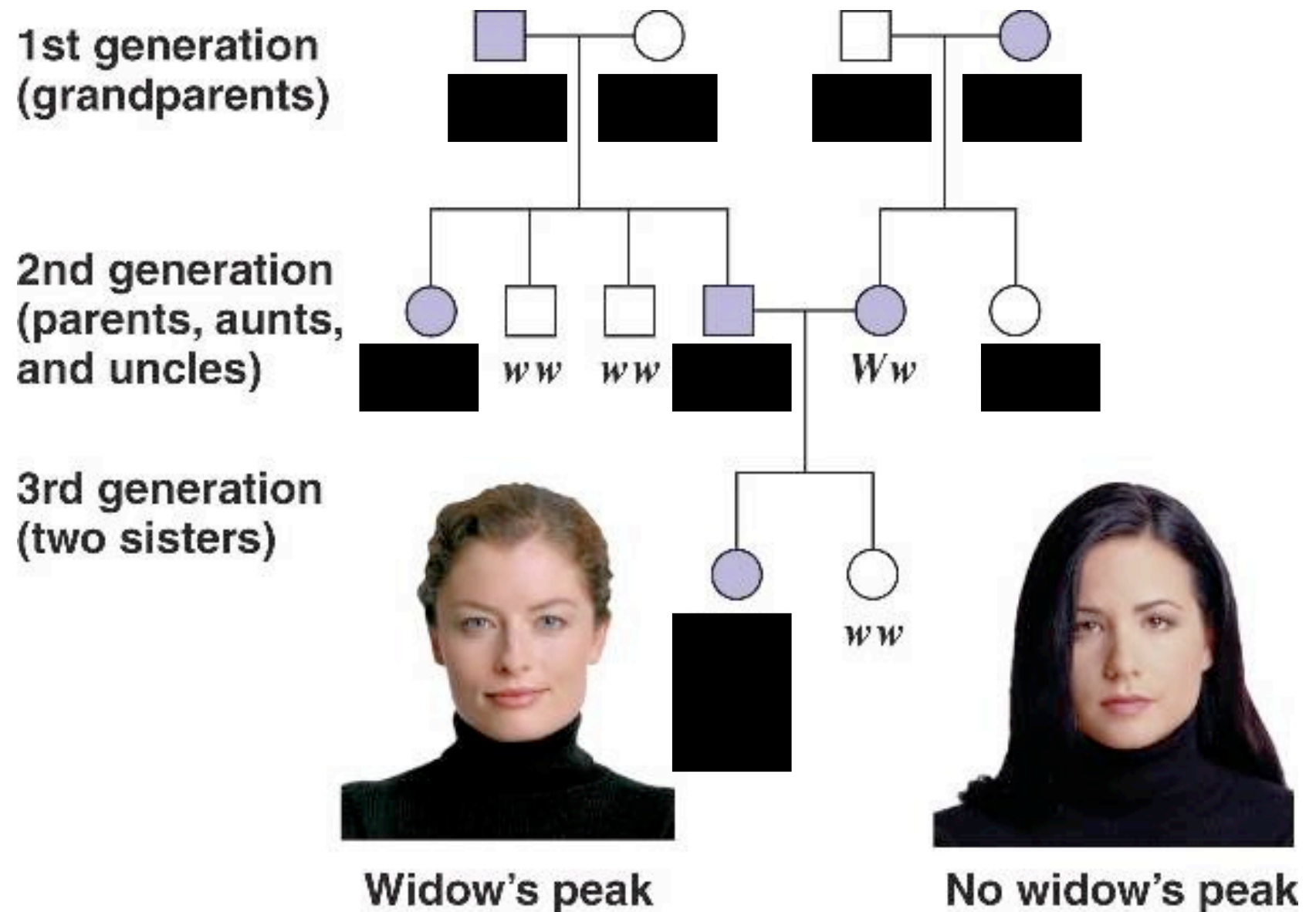
Widow's peak



No widow's peak

Larger family trees

- A larger example where W is dominant over w
- Goals:
 - genetic counseling
 - finding genetic factor of diseases / traits
- Complication factor
 - incomplete data

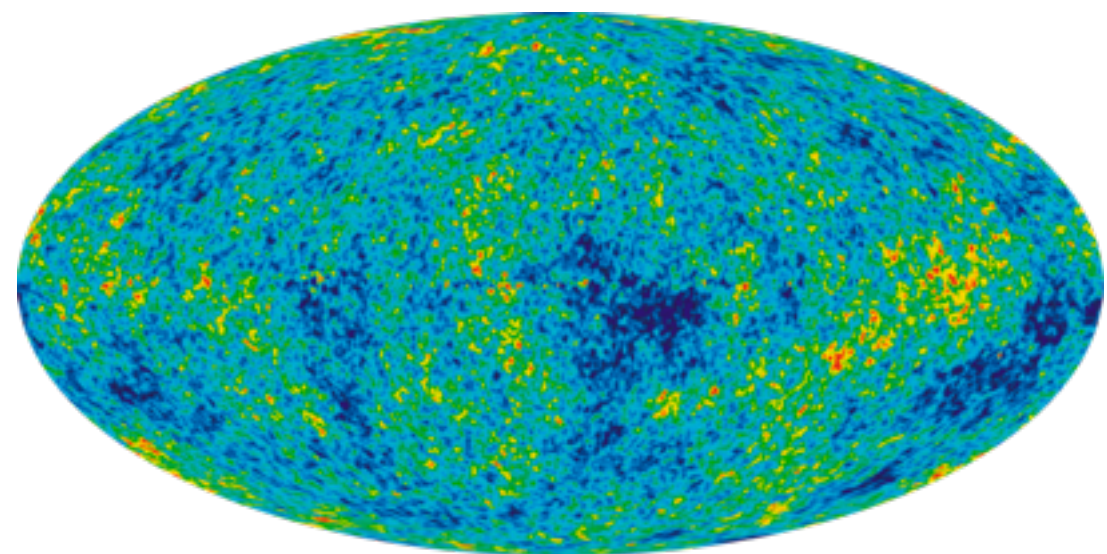
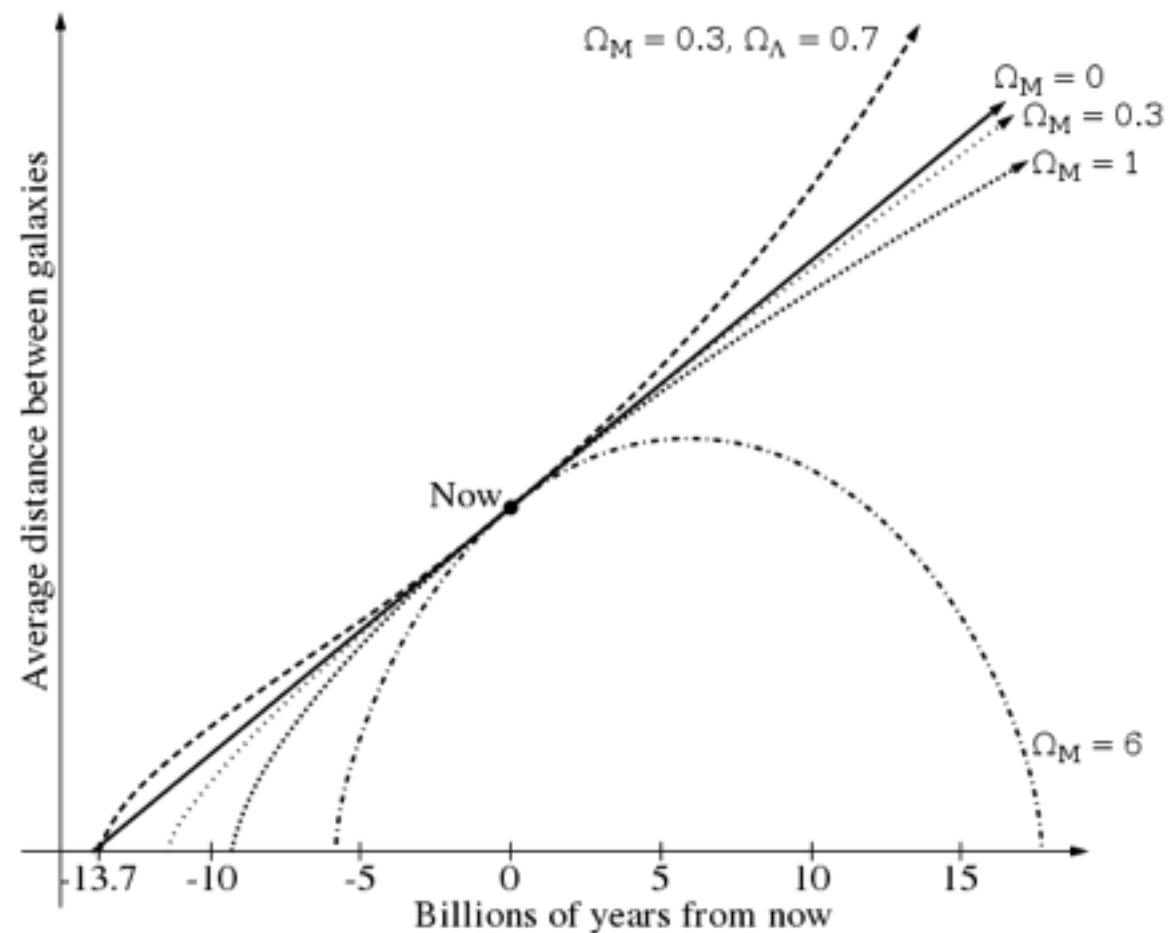


Technique: marginalization

Ex. 7

Astrophysics: Estimating the age and faith of the Universe

- **Goals:** finding the Universe's
 - age
 - density (\Rightarrow faith)
- **Data:** Cosmic Microwave Background (CMB): remnants of Big Bang
 - Detailed map from the Planck satellite
- Age, Physical constants \Rightarrow known *distribution* on CMP
- Invert using Bayes' rule



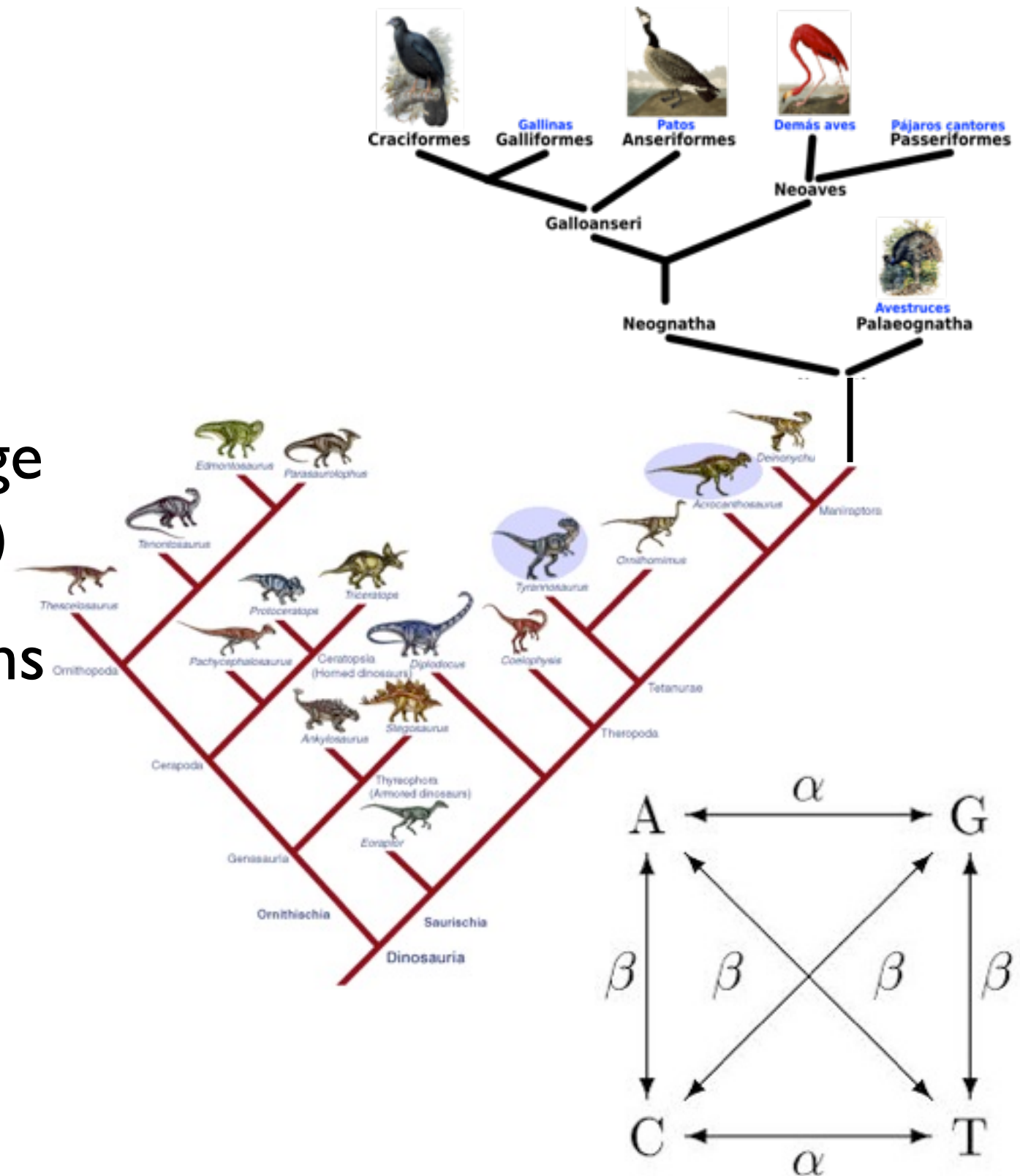
Phylogenetics: Reconstruction of ancient species

- **Goals:**
 - better understand ancient species
 - revive them?
- **Data:** fossil DNA
- Limitation: degrades after few 1000s years
- Are dinosaurs' genomes completely lost?



Phylogenetic tree

- **Idea:** use the genomes from the descendants of dinosaurs (modern birds)
- We know how DNA change over time (probabilistically)
- Marginalization of unknowns (as in family tree example)
- Additional challenge: structure of tree is unknown



Outline of the course

- Discrete probability models
- Conditioning and Bayes
- Expectation
- Continuous probability models
- Asymptotics

Random variables

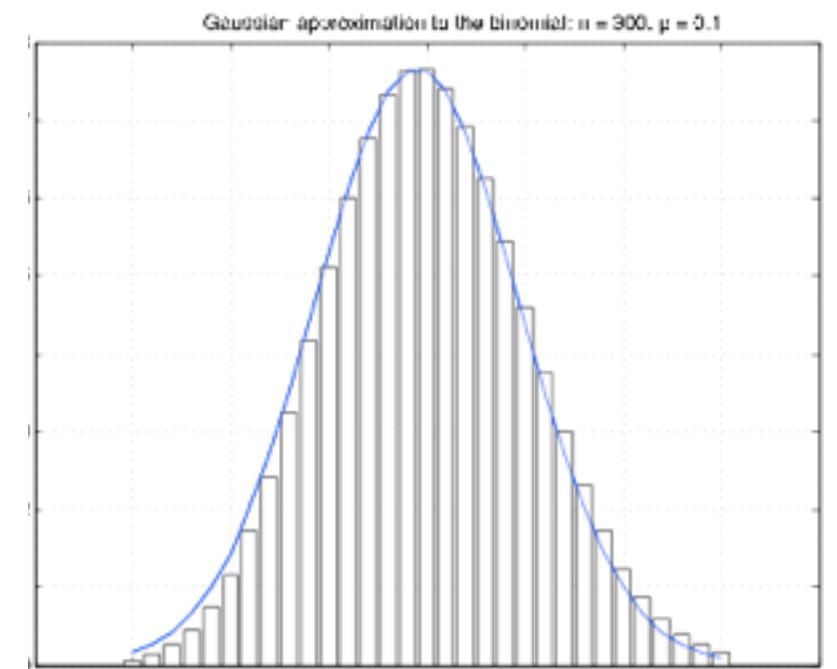
- Fundamental object of study
- Examples of a random variable X
 - The height of a UBC student picked at random
 - Gambling example ('Rademacher coin')

Surprising challenge

- Sums of random variables
 - Omnipresent
 - Taking the sum of variables is easy, so taking the sum of *random* variables should also be easy, right?
 - Not quite... consider for example the problem of computing the probability that the sum of 100 coins is greater than 50.
 - Would have been hard in the pre-computer era
 - Generalized versions of this problem still hard with computer

Asymptotics to the rescue

- Another surprise: sums of random variables can be approximated by something simple when large number of terms involved
- No matter what each X is!!! (almost)
- Also explains why we will spend disproportionate amount of time on some specific types of random variables (normal/gaussian, Poisson, ...)



300 coins