# CS 112: Modeling Uncertainty in Information Systems

Prof. Jenn Wortman Vaughan April 2, 2012

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Everything Images Maps Videos News Shopping	Cs 112 Introduction to CS II www.cs.bu.edu/fac/snyder/cs112/ Instructor: Wayne Snyder Email: snyder@cs.bu.edu. Office: MCS 147. Office Hours: Thurs 2-5 in MCS 147; Math Tutoring in Rich Hall Tues 7-10pm. Cell Phone: CAS CS 112 B1 - Spring 2012 www.cs.bu.edu/fac/byers/cs112.html CAS CS 112 B1 - Spring 2012 - Introduction to Computer Science II. Class meeting time: Tues/Thurs 12:30-2:00, CAS 324. Lab meeting times: Mon 1-2, Mon 2-3	Ads	White Tiger Costume Claw Legwarmers Cs112 \$49.99 - eBay Buy It Now from eBay Top-rated Sellers!
More Los Angeles, CA Change location	CS112 Introduction to Programming (Spring 2012) — Home flint.cs.yale.edu/cs112/ CS112 Introduction to Programming (Spring 2012) : Home Use Piazza to ask any cs112- related questions. Email specific instructor, TA, or peer tutor if you		
Show search tools	CS112: Computer System Modeling Fundamentals Spring 2011 www.cs.ucla.edu/~jenn/courses/S11.html CS112: Computer System Modeling Fundamentals. Lectures: Tuesdays & Thursdays, 4:00- 5:50pm, Kinsey Pavilion 1200B Discussions: Fridays, 4:00-5: 50pm,		







To: Jenn Wortman Vaughan From: Jeff Vaughan Subject: Plans for tonight

To: Jenn Wortman Vaughan From: Jens Palsberg Subject: Meeting



To: Jenn Wortman Vaughan From: Bob Smith Subject: V14GR4 4 U



To: Jenn Wortman Vaughan From: NIPS Committee Subject: Paper decision







Many areas of computer science involve uncertainty.

In this course, you will develop foundational, mathematical reasoning skills that will help you deal with it.

Course Overview & Logistics

# Part I: Formalizing Uncertainty

- Sample spaces and events
- Probability laws
- Conditional probability, Bayes' theorem, and independence
- Counting problems (combinations, permutations)
- Random variables
- Expectation, mean, and variance
- Information and entropy
- Covariance and correlation
- Limit theorems

# Part II: Reasoning Under Uncertainty

**Inference:** How can we estimate quantities or properties of interest from observed data?

• Example: Inferring if a new email is spam or not

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**Inference:** How can we estimate quantities or properties of interest from observed data?

- Example: Inferring if a new email is spam or not
- **Markov Chains:** How can we reason about random processes that evolve over time?
  - Example: Random surfer used in Google's PageRank

### Course Staff

Instructor: Prof. Jenn Wortman Vaughan (jenn@cs.ucla.edu) Office Hours: Thursdays, 11–12pm, Boelter 4532H and by appointment

TA: Jacob Mathew (jacobgmathew@gmail.com)Office Hours: Tuesdays, 11am–1pm, Boelter 2432Discussions: Fridays, 2-3:50pm and 4–5:50pm

Graders: Ding Zhao (zhaoding@ucla.edu) and Jake Stothard (stothardj@gmail.com)

#### Resources

Required textbook:

Introduction to Probability (2nd Edition) by Dimitri P. Bertsekas and John N. Tsitsiklis

Other material, including reading assignments, problem sets, and lecture slides, will be posted at:

http://www.cs.ucla.edu/~jenn/courses/S12.html

## Breakdown of Grades

Five Homework Assignments (25%)

• Mostly pencil-and-paper, a little C++ programming

In-class Exercises (15%)

• Given frequently during both lecture and discussions

Midterm (30%)

• Wednesday, May 2, sheet of hand-written notes allowed

Final Exam (30%)

• Tuesday, June 12, cumulative, same rules as midterm

**Grading:** On each assignment, a subset of problems will be graded. Your grade will be based on these problems only.

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**Regrades:** Requests for regrades must be submitted in writing. Requests will not be accepted until 48 hours after assignments are returned, and all requests must be submitted within one week.

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Academic Honesty: Collaboration is encouraged, but you must follow the academic honesty policy!

## Academic Honesty

- Each student must write down his or her own solutions independently in his or her own words.
- Each student must submit a list of anyone with whom the assignment was discussed.
- All sources (internet included) must be properly credited.
- Solution sets from this course or any other course cannot be used under any circumstances.

## Piazza Discussion Board

Use Piazza to...

- Discuss or ask for clarifications about material covered in class or in your reading
- Share useful references you've found
- Find other students to form study groups
- Answer questions to solidify your understanding

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The course academic honesty policy must be followed on Piazza at all times!

## In general, to succeed in this class...

- Complete the required reading before class
- Attend lecture and discussion sections regularly
- Start your assignments early
- Check the course website often:

http://www.cs.ucla.edu/~jenn/courses/S12.html

• Ask the course staff if you're unsure about any policies

#### A Two-Minute Review of Sets

• A set S is a collection of objects

$$S = \{1, 3, 5\}$$

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- In this example, 1 is an element of S, while 2 is not  $1 \in S$ ,  $2 \notin S$
- {1, 3} is a subset of S, but {1, 2} is not  $\{1, 3\} \subset S, \quad \{1, 2\} \not\subset S$
- S is a subset of itself, but not a strict subset  $\{1, 3, 5\} \subseteq S, \qquad \{1, 3, 5\} \not\subseteq S$

• The universal set  $\Omega$  contains all possible elements of interest in some context

for die rolls,  $\Omega = \{1, 2, 3, 4, 5, 6\}$ 

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- The complement S<sup>c</sup> of a set S (with respect to Ω) is the set containing all elements of Ω not contained in S
  S = {1, 2, 3}
  S<sup>c</sup> = {4, 5, 6}
- The empty set  $\emptyset$  is the set containing no elements

• The union of sets S and T is the set of elements contained in at least one of S and T

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S U T =  $\{1, 2, 3, 5\}$ 

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- S and T are disjoint if  $S \cap T = \emptyset$
- S and T partition  $\Omega$  if they are disjoint and S U T =  $\Omega$

#### Algebra of Sets

If you don't feel completely comfortable with these properties, spend some time reviewing Chapter 1.1 soon...

SUT = TUS $(S^c)^c = S$  $S \cap S^c = \emptyset$  $S \cup \Omega = \Omega$  $S \cap \Omega = S$  $S \cap (T \cup U) = (S \cap T) \cup (S \cap U)$  $S \cup (T \cap U) = (S \cup T) \cap (S \cup U)$  $S \cup (T \cup U) = (S \cup T) \cup U$ 

# Venn Diagrams



## What is probability?

## Interpretation: Relative Frequency

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What is the probability that a spaceship will remain operational during a mission to Mars? Hmm....

## Interpretation: Subjective Belief

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Can be useful when we need a principled, systematic way to make choices in the presence of uncertainty

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  - Flipping two coins in a row. (Multiple alternatives..)
  - Pulling a random card from a deck.
  - Running a computer program.

### Samples Spaces

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Always has to satisfy two properties:

- Elements must be mutually exclusive (only one can be the real outcome)
- Elements must be exhaustive

(we always obtain an outcome from the sample space)

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- If the experiment is a toss of a die, events could include: "the number is odd"
  "the number is less than 3"
  "the number is 2"
- Each of these can be represented as a set of atomic events the basic elements of the sample space.

Since events are sets, we can apply set operations to them...

 $A = \{1,3,5\}$ "the number is odd" $B = \{1,2,3\}$ "the number is  $\leq 3$ "

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We can encode our knowledge or beliefs about the likelihood of particular sets of outcomes using a probability law, which assigns a probability P(A) to each event A

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Together, a sample space and a probability law define a probabilistic model

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First Exercise!! Use the probability axioms and set theory to derive answers to the following questions.

- Suppose A, B, and C are disjoint. What is  $P(A \cup B \cup C)$ ?
- What is  $P(\emptyset)$ ?
- What is P(A<sup>c</sup>)?

#### Other Properties of Probability Laws

For any events A and B,

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- $P(A \cup B) \le P(A) + P(B)$

These can all be verified informally using Venn diagrams.

## For next time...

- Sign up for Piazza.
- Get the textbook.
- Check the website for assigned reading.

Homework 1 will be posted later this week!