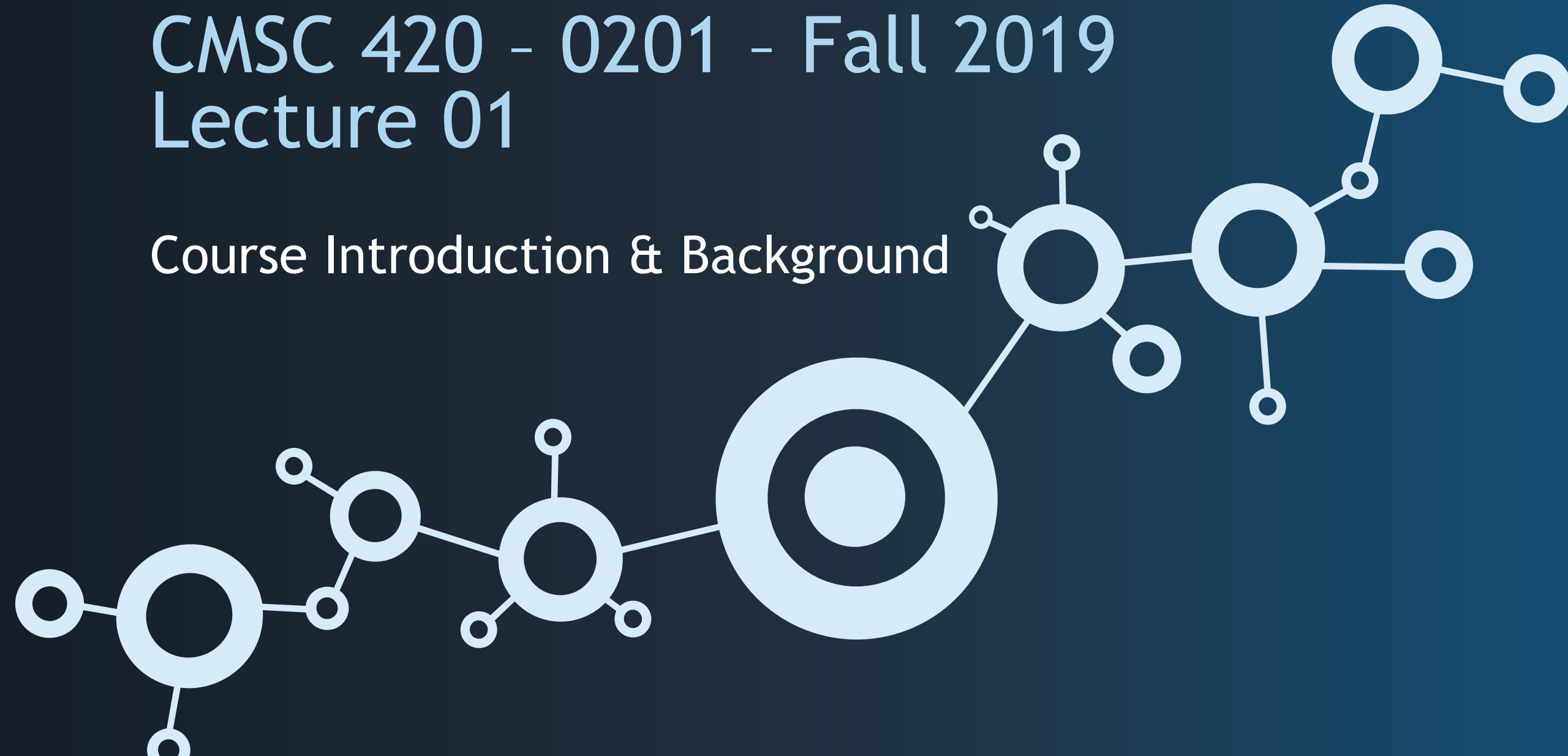


# CMSC 420 - 0201 - Fall 2019

## Lecture 01

Course Introduction & Background



# CMSC 420 - Course Overview

- Fundamental data structures and the algorithms for building and maintaining them
- Mathematical methods for analyzing their efficiencies
- Applications and implementation
- See the course syllabus:

<http://www.cs.umd.edu/class/fall2019/cmssc420-0201/syllabus.html>

# Elements of Data Structures

- How to store, access, and manipulate data
- Fundamental to Computer Science

## Examples:

**Information Retrieval:** Web search, forensic search, image search

**Geographic Information Systems:** Proximity searching (How many people live within 25 miles of the Mississippi river?)

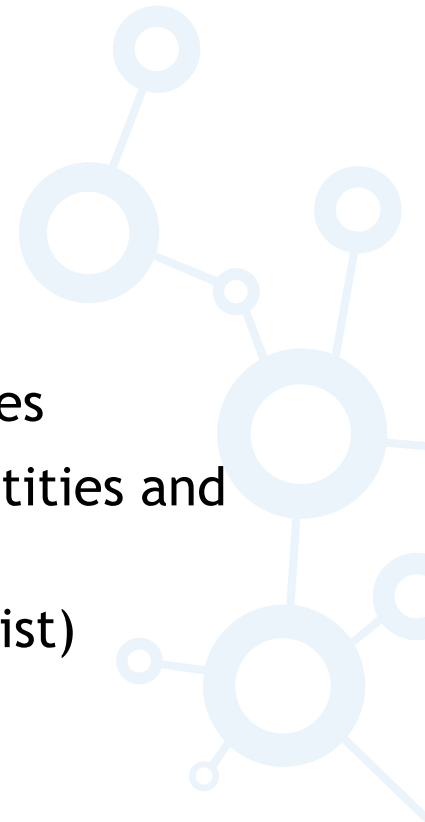
**Text/String Search:** Symbol tables in a compiler, document search

**Networking:** Shortest paths/distances in a graph

**Computer Graphics/VR:** Visibility culling - What can be seen from a given vantage point?

# Elements of Data Structures

- Basic aspects of a data structure:
  - **Modeling**: How real-world objects presented as abstract mathematical entities
  - **Operations**: The allowed functions to store, access, and manipulate these entities and their formal meanings
  - **Representation**: Concrete representations in memory (e.g., array vs. linked list)
  - **Algorithms**: Computational approaches to execute these operations
- What we will teach:
  - **Tools**: Common techniques in the design of data structures
  - **Design principles**: Methods that can be widely applied in data structure design
  - **Overall**: How to design data structures, how to implement these designs, and how to evaluate them



# Two Angles Towards a Common Goal

- **Theoretical:**
  - Mathematical description of the resources needed by a data structure
  - Criteria:
    - Query time
    - Construction time
    - Update time
    - Memory requirements (and possible tradeoffs with query time)
- **Practical/Empirical:**
  - Ease of implementation
  - Efficiency on actual data sets



# Review: Asymptotic Analysis

## ■ Principles of Asymptotic Analysis:

- Running time as function of **basic parameters** (input size, number of vertices/edge in a graph, basic properties of the input set, dimension of the space)
- Worst-case or average-case?
  - **Worst-case**: Maximum running time for a given input size
  - **Average-case**: Expected running time over a given probability distribution of inputs
  - **Randomized**: When algorithms use randomization, it is common to express running time in the worst-case over inputs, but in the expected-case over randomized choices
  - **Amortized**: Running time averaged over a long sequence of operations

## ■ Asymptotic notation (“big-O”) to focus on growth rate:

$$T(n) = 22n^2 + 7n \log n + 5n^3 = O(n^3)$$

# Review: Asymptotic Analysis

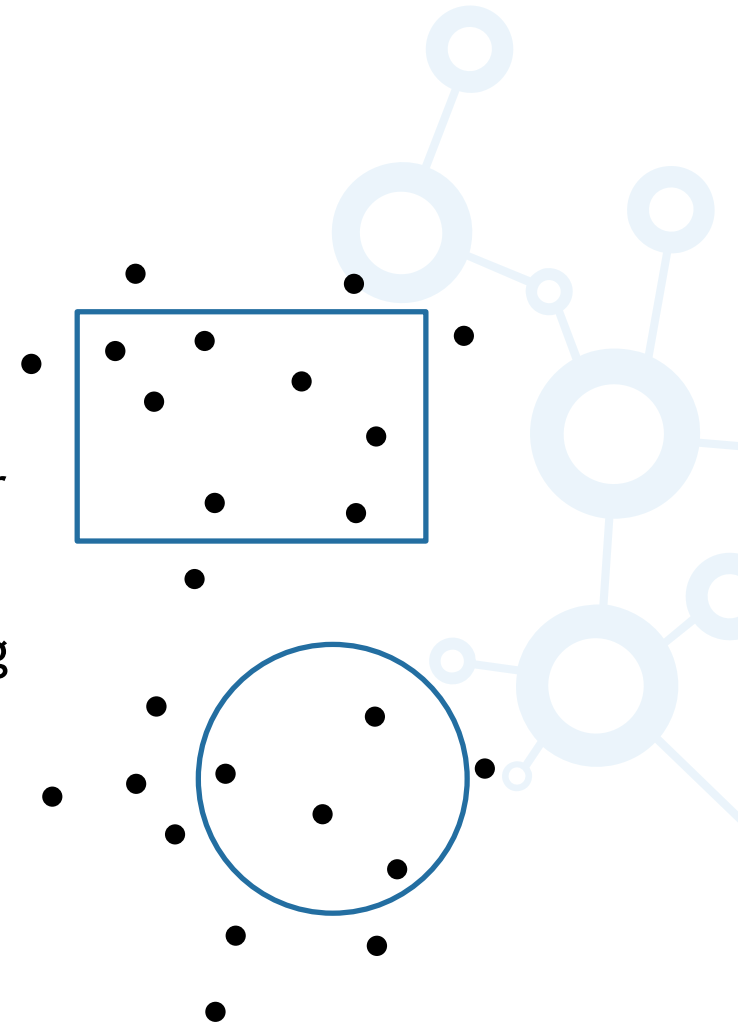
## ■ Examples:

- $O(1)$  (**Constant time**): Can't beat this!
  - Examples: Stack push/pop, hashing (expected case)
- $O(\alpha(n))$  (**Inverse of Ackermann's function**):
  - Insanely slow-growing, but not constant.  $\alpha(n) \leq 5$  if  $n \leq$  number of atoms in universe
  - Example: Disjoint-set union/find (used in Kruskal's MST algorithm)
- $O(\log \log n)$ 
  - Example: Van Emde Boas Trees - For storing small integers
- $O(\log n)$  (**Logarithmic time**): “Gold standard” for comparison-based structures
  - Examples: AVL tree, Red-Black tree, 2-3 tree, ...

# Review: Asymptotic Analysis

## ■ More Examples:

- $O((\log n)^c)$  (Polylogarithmic time):
  - Example: Orthogonal range searching in 3-dimensions and higher
- $O(n^p)$ , where  $0 < p < 1$  (Sublinear polynomial time):
  - Examples: Nearest neighbor searching, spherical range searching
- $O(n)$  (Linear time)
- $O(n \log n)$ 
  - Standard for any algorithm based on comparison-based sorting
- $O(n^p)$ , where  $p$  is any constant (Polynomial time)
- $O(p^n)$ , where  $p > 1$  is any constant (Exponential time)





# Summary

- Course Overview
- Basic elements of Data Structure Design
- Review of Asymptotic Analysis

