## COT 6936: Topics in Algorithms

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https://moodle.cis.fiu.edu/v2.1/course/view.php?id=612
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## Presentation Outline

## Applications

- Graphics


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## Convex Hull Problem

- This is a fundamental problem in Computational Geometry.
- There are many algorithms for solving this problem ...


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Also shows proof of correctness;
$O(n \log n)$ time, mainly for sorting


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$$

