



# CS356 Hardware Accelerator Architectures

## Research Patterns

Suhaib Fahmy

[suhaib.fahmy@kaust.edu.sa](mailto:suhaib.fahmy@kaust.edu.sa)



Slides based on “Research Patterns” by Nick Feamster and Alex Gray



# General Approach

1. Find a problem
2. Understand the problem
3. Somehow make a plan for a solution, carry it out
4. Review the solution



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# Finding Problems

- ▶ Hop on a trend
- ▶ Find a nail that fits your hammer
- ▶ Revisit old problems (with new perspective)
- ▶ Making life easier
  - ▶ Pain points
  - ▶ Wish lists
- ▶ “\*-ations”
  - ▶ Generalization
  - ▶ Specialization
  - ▶ Automation



# Hop on a trend

- ▶ Need places to discover trends
- ▶ Funding agencies
  - ▶ Funded proposals
  - ▶ Calls for proposals
- ▶ Conference calls for papers
- ▶ Industry/technology trends: trade publications



# Hop on a trend - examples

- ▶ Presently:
  - ▶ Large Language Models specifically and Machine Learning generally
  - ▶ Quantum computing
  - ▶ Self-driving cars/robotics
  - ▶ Distributed ledgers/cryptocurrency
- ▶ Varying stages, longevity, applicability



# Finding a nail for your hammer

- ▶ Become an expert at something
  - ▶ You'll become valuable to a lot of people
- ▶ Develop a system that sets you ahead of the pack
- ▶ Apply your "secret weapon" to one or more problem areas
  - ▶ Algorithm
  - ▶ System
  - ▶ Expertise
- ▶ "Turn the crank"





# Revisiting problems

- ▶ Previous solutions may have assumed certain problem constraints
- ▶ What has changed since the problem was “solved”?
  - ▶ Processing power
  - ▶ Cost of memory
  - ▶ New protocols
  - ▶ New applications
  - ▶ ...



# Revisiting problems - examples

- ▶ Machine learning accelerators based on systolic arrays from the 1960s!
- ▶ Networking to software-defined networks to in-network computing
- ▶ Cryptographic methods with scaling compute capability/quantum



# Pain points

- ▶ Look to industry, other researchers, etc. for problems that recur
- ▶ In programming, if you have to do something more than a few times, script!
- ▶ In research, if the same problem is recurring and solved the same silly way, there may be a better way...



# Automation

- ▶ Some existing problems, tasks, etc. are manual and painful
  - ▶ Automation could make a huge difference
  - ▶ It's also often very difficult because it requires complex reasoning
- ▶ Related to pain points





# Automation – examples

- ▶ Manual code optimisations for specific architectures to maximize performance
- ▶ Generators to automate use of intrinsics for auto-vectorisation
- ▶ Automated passes and backends for compiler flows to solve the general case



# Wish lists

- ▶ What systems do you wish you had that would make your life easier?
  - ▶ Lighter VR headset?
  - ▶ Faster response time from an LLM?
  - ▶ ...
- ▶ What questions would you like to know the answer to?
  - ▶ Chances are there is data out there to help you find the answer...



# Generalize from specific problems

- ▶ Previous work may outline many points in the design space
- ▶ There may be a general algorithm, system, framework, etc., that solves a large class of problems instead of going after “point solutions”
- ▶ E.g. highly application-specific accelerators, vs. more generalised architectures optimized for a domain of applications



# Specialise a general problem

- ▶ Finding general problems
  - ▶ Look for general “problem areas”
  - ▶ Look for taxonomies and surveys that lay out a problem space
- ▶ Applying constraints to the problem in different ways may yield a new class of problems
  - ▶ Example: Routing (in wireless, sensor networks, wired, delay-tolerant networks, etc.)





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# Exhaustive search

- ▶ Collect data
  - ▶ Often this can enhance your expertise as a side effect
- ▶ Model the problem
  - ▶ List all of the constraints to a problem space
  - ▶ Consider all of the different angles within your model that you might be able to attack the problem (example: phishing attacks, routing configuration errors)
- ▶ Consider many other examples
  - ▶ May suggest general framework or approach
  - ▶ You may also see a completely different approach



# Formalization

- ▶ Define metrics
  - ▶ Consider ways to measure the quality of various solutions
  - ▶ What constitutes a “good solution”
  - ▶ Objective functions can be optimized
- ▶ Formalization/modeling can lead to simplifying assumptions (hopefully not over-simplifying)
  - ▶ Can also suggest ways to attack the problem
  - ▶ ...or an algorithm itself



# Formalization

- ▶ Think about what works best for your domain
  - ▶ What metrics are of interest at the component and system level?
  - ▶ How are the components composed and how do they interact?
  - ▶ What methodology can you use to model at the component level and higher?
    - ▶ Event-driven, networks of queues, time schedules, etc.





# Decomposition

- ▶ Given a model, it often becomes easier to break a solution into smaller parts
- ▶ Understand each part in detail, and how they interact
- ▶ Then revisit the whole



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# Consider related problems

- ▶ Try to restate the problem, or create an equivalent problem
  - ▶ Consider different terminologies and representations
- ▶ See if your problem matches a general form already formalized
- ▶ Can you use the solution to a related problem?
- ▶ E.g. routing in circuit design relates to general path-finding algorithms



# Make analogies

- ▶ Make an analogy to another problem, then look at its solution
  - ▶ “structural transference”
  - ▶ look for “symmetries”, or interchangeable parts





# Change the problem to a solvable one

- ▶ Make simplifying assumptions
  - ▶ Violate some of the constraints of the problem
  - ▶ Define a sense of approximation to the ideal solution
- ▶ Then revisit the original problem
- ▶ Make the minimally-simpler problem; then relate the solutions to the two problems
  - ▶ “mathematical induction”



# Just start, with anything

- ▶ Start with a strawman solution, then modify as needed
  - ▶ e.g. (in algorithms): Propose a simple algorithm, check its correctness
  - ▶ e.g. (in data modeling): Look at simple statistics of a dataset, then dive into anomalies
  - ▶ e.g. (in systems): Just whip up some code
- ▶ A working simple solution is better than an intractable complex formulation



# Consider nature

- ▶ Introspection: How does a human naturally solve this problem?
- ▶ How does nature solve this problem?
- ▶ E.g. neural networks somewhat nature inspired, vision algorithms
- ▶ E.g. various optimization approaches: genetic algorithms, ant colony optimization, etc.



# Work backward from the goal

- ▶ Visualize the solution, and what it must look like, or probably looks like
- ▶ See what's needed to get there
- ▶ Consider all the solutions that can't work



# Solve a part, or each part

- ▶ Solve each part separately, then stitch the solutions together
  - ▶ Start with the part which is most tractable
  - ▶ “divide-conquer-merge”
  - ▶ Be careful: it’s always best to avoid separate objective functions when possible
- ▶ Perhaps finding a good solution to a part is a good problem in itself



# Think in speech or pictures

- ▶ Use dialogues with others
  - ▶ Or yourself
  - ▶ Talk to people who approach things differently from yourself
- ▶ Draw pictures
  - ▶ Add auxiliary elements, to be able to relate to other problems/solutions
- ▶ Maths is not the only way to reason about ideas





# Come from all angles

- ▶ Keep coming with a new twist on the problem
  - ▶ Break out of a thinking pattern or dead end
  - ▶ A new twist renews motivation
  - ▶ “Where there’s a will, there’s a way”
- ▶ Keep track of all your ideas and partially-completed paths



# Let your subconscious work

- ▶ Immersion
- ▶ Stay relaxed
- ▶ Or: use deadlines to force shortcuts
- ▶ Wider reading can stimulate new approaches to thinking



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# Look back at your solution

- ▶ Check that it really works
  - ▶ If it works, note the key to why, more abstractly
  - ▶ Were all of the constraints, difficulties, and facts used and accounted for
- ▶ Try to improve upon it
  - ▶ Can you achieve the same thing more directly or easily



# What else can your solution do?

- ▶ Now you have a hammer
- ▶ Can you use the solution for some other problem?
  - ▶ A more general form of the problem?
  - ▶ An interesting special case?
  - ▶ A related problem or analogous problem?



# Making a “theory”

- ▶ If you’re very successful, you may have a “theory” = a framework for characterizing problems and/or solutions
  - ▶ Says when it applies, when it doesn’t
  - ▶ Characterizes the hardness of different problems
    - ▶ May identify simple special cases
  - ▶ Characterizes the quality of different solutions
    - ▶ How long it takes, amount of resources it uses
  - ▶ Show/characterize solution meeting criteria
    - ▶ correctness, convergence, etc.

