

Biocomputing and Developmental Systems (BDS) Group



1 AUTOMATIC DESIGN OF DIGITAL CIRCUITS

AD/DC

Design Specification

Architecture Description

IO Description

Use Machine Learning to automatically design digital circuits from high level descriptions.

Grammar

```
(assign) ::= assign output = (expression);
(expression) ::= (expression) + (expression)
| (expression) * (expression)
| input | (c) | (x)
(c) ::= c0 | c1 | c2 | c3
(x) ::= x1 | x2 | x3
```

A grammar specifies the subset of a Hardware Description Language

Grammatical Evolution (GE)

Dataset

GE automatically generates a set of solutions with different trade-offs

Gatelevel Netlist

```
// Generated by Genetic Design (GD) Synthesis Solution 21.30.0002.0
// Generated on: Jan 4 2024 08:51:27 GMT (Jan 4 2024 00:51:27 UTC)
// WorldCacheDirectory: /root/.cache
module Counter_Module(clk, rst, sel, a);
input [2:0] clk;
input [2:0] rst;
input [2:0] sel;
output [2:0] a;
wire [2:0] q;
wire [2:0] n_q;
always @(posedge clk) begin
    if (rst == 1) begin
        q <= 0;
    end else begin
        q <= q + 1;
    end
end
endmodule
```

Place-and-Route (PnR)

A description of a digital circuit at the level of individual logic gates and their interconnections.

Backend design of the IC.

Fabricated IC using TSMC 65nm technology.

Functional Verification of the IC using FPGA (Field Programmable Gate Array).

Logic Synthesis

	BC	TC	WC
Temperature (Celsius)	0°	25°	125°
Core Voltage (Volts)	1.32	1.2	1.08
IO Voltage (Volts)	2.75	2.5	2.25
Area (Cell count)	6562	6566	6513
Total Power (in Milliwatt)	21.30	17.60	14.40
Datapath Delay (in Pico Second)	1691	5180	7986

Synthesizable HDL Code

```
module Counter_Module(input wire clk, input wire rst,
input wire sel, output reg [2:0] a);
parameter S0 = 2'b0;
parameter S1 = 1'b1;
reg state;
always @(posedge clk) begin
    if (rst == 1) begin
        state <= S0;
    end else begin
        state <= S1;
    end
end
endmodule
```

Transform the high-level hardware description into a gate-level netlist for efficient implementation in hardware.

These solutions can be automatically converted into hardware

2 UNLEASHING THE POWER OF AI AND DIGITAL TWINS FOR EMERGENCY CARE

ALTER
UNLEASHING THE POWER OF AI AND DIGITAL TWINS

Number of Patients

Year

Actual Total

Linear Prediction of Total

Total No. Trolleys

Total No. Beds

DATA

VALUE PROPOSITION

INTERCONNECTED

A dynamic and interconnected Digital Twin (DT) to streamline patient flow

ASSISTANT TECHNOLOGIES - EFFECTIVE USE OF 'WAIT' TIME

Waiting patients monitored by AI

Intervention when most appropriate

Patient vital signs deteriorate. AI notifies the appropriate staff for intervention.

Reduced burden on staff. Efficient use of patient wait time. Valuable patient information tracked.

Ospidéal OL UL Hospitals

ICHEC Irish Centre for High-end Computing

HE

Digital for Resilience Challenge

3 Explainable Safety and Security

Use Machine Learning to identify critical safety issues

Undamaged Damage I Damage II Damage III

Traditional Artificial Intelligence (AI)

Explainable Artificial Intelligence (XAI)

Black box models

Transparent models

Traditional AI only gives prediction.

XAI explains **why** a decision was made.

WACT

WIRELESS AUTOMATIC COMPILER TUNING

Improve Wireless software performance by identifying optimal compilation options

Treat compiler flag optimisation as a machine-learning problem in software engineering

HUAWEI

LLVM Compiler

Source Code

Code Binaries

Execute/Analyze/Instrument/Measure

Performance Measurements (CPU cycles, Cache misses, etc.)

Search Engine

- Genetic Algorithms
- Reinforcement Learning
- Deep Learning
- Bayesian Optimisation

Profiling and generating heatmaps

Minimize measurement bias

Coverage analysis

4 WIND TURBINE ROTOR SIDE UNDERPERFORMANCE THROUGH POWER CURVE ANALYSIS

Diagnosing poor turbine performance from gear box failures is simple, as energy lost is measured as heat dissipation.

Rotor issues are trickier; there's no direct way to spot problems in the alarm logs.

Hub

Blades

This research will develop a data-driven AI model to identify failures in rotor-side components of wind turbines after diagnosing underperformance.

Power Produced

Wind speed

A power curve (red) show the expected power at a certain wind speeds. If actual energy (black dots) falls below, it signals turbine underperformance. This approach reliably detects turbine efficiency.

The main hurdle in this diagnosis is inaccuracies in wind speed measurements, leading to false underperformance signs. Given anemometers' long lifespan, it's difficult to identify when they need replacing or maintenance.

Time

Not working correctly

Anemometer maintenance or replacement

Working correctly

Unknown status

When the power curve indicates normal performance, we can gauge the uncertainty level, hinting at the anemometer's accuracy. Normal Behaviour Models (NBM), a type of anomaly detection, learn from data representing typical operation. Anomalies are flagged when predictions deviate from actual outcomes.

Forecast start

Upcoming event

Real

Predicted

ENERGYPRO