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Bio-Inspired Algorithmic Framework for Adaptive Façade Design in Energy-Efficient Architecture

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Abstract

The construction sector is one of the largest contributors to global energy consumption and greenhouse gas emissions. High thermal losses and over-reliance on mechanical heating or cooling systems are due to the frequent failure of traditional facades to adapt to changing environmental conditions. Inspired by natural systems such as the thermic cooling of plant leaves and the responsive structure of pine trees, this study provides an algorithmic bio-inspired framework for adaptive building facade design. The framework optimizes the geometry of the facade and the material response to occupancy dynamics, temperature, and sunlight, combining evolutionary algorithms, swarm intelligence, and morphogenetic principles. Conventional methods, because of their rigid control logic, often fail to strike the right balance between performance, comfort, and aesthetics. To achieve a dynamic balance between energy consumption and indoor comfort, the proposed framework will introduce an adaptive, self-organising mechanism that continuously changes the elements of the facade in real time. Computational simulations using parametric modelling and energy-use analysis tools shall be performed to evaluate metrics like temperature comfort index (PMI), solar heat gain coefficient (SHGC), daylight autonomy (DA), and energy use intensity (EUI).

Keywords; Bio-inspired design, Adaptive facade, Energy efficient architecture, Evolutionary algorithm, Thermal comfort.

INTRODUCTION

The built environment accounts for a large share of global energy consumption and greenhouse gas emissions. The envelopes of buildings is essential for energy saving and occupant comfort, as buildings account for about one-third of global energy consumption and associated emissions [2]. Static facade systems are often difficult to adapt to changing weather conditions, sun direction, and occupancy dynamics. The facades act as an interface between the indoor and outdoor environment and affect the thermal, lighting, and ventilation performance. Recent advances in materials, automation, and computational design have opened the door to facades becoming more responsive and adaptable, but the gap is still wide[1].

Challenges

Several important issues hamper the design and implementation of adaptive facade systems in energy-efficient buildings. First, many facade systems are still passive and rigid, unable to dynamically adapt to changing solar loads, changes in outdoor temperatures, or changes in indoor use patterns. This means that despite the efforts made by the design to provide insulation and shade, thermal losses or solar overheating are often encountered. Simulation studies have shown, for example, that adaptive shading, phase-changing materials, and air-conditioned wall cavities can reduce energy loads; however, their application is still limited and location-specific [3].

Secondly, the assessment and optimization of the performance of the facade are intrinsically multifaceted. Adaptive facades influence temperature, daylighting, visual comfort, and energy flows at the same time, and interactions between these areas are context-dependent and non-linear [2].

Instead of combining metrics such as energy use intensity (EUI), daylight independence (DDA), and heat comfort (PMV), many of the studies currently being conducted focus on one or two performance metrics. Moreover, control systems, sensors, and actuators integrated in adaptive facades often face problems of integration, cost, and reliability.

Third, most optimization frameworks do not take inspiration from efficient and resilient systems found in nature, despite the potential benefits of biomimetic design. With minimal input, natural systems such as leaves, pine needles, and hygroscopic skins exhibit adaptive behavior and provide rich metaphors for the response to the facade. In the absence of algorithmic frameworks integrating bio-inspired mechanisms with parametric design, real-time control, and enclosure optimization, many designs remain static or pre-programmed, rather than dynamically adapting.

Fourth, there is little commercialization and widespread use of adaptive signage. High start-up costs, problems with retrofitting existing building stock, disagreements on standards for performance assessment, and a lack of data on the ground [2] are some of the obstacles. Many promising concepts remain in the laboratory or simulation phase with no standardized metrics and validation, which hampers their wide market acceptance.

Problem Statement

Static façade systems lead to less-than-ideal performance, excessive reliance on mechanical heating or cooling and compromise occupant comfort due to their inability to adapt to the temporal changes in solar radiation, outdoor climate, occupancy and indoor comfort requirements. To overcome these limitations, a unified algorithmic framework is needed to allow for dynamic adaptive geometry and material behaviour, based on bio-inspired principles, and optimised along a range of performance metrics (energy consumption, daylighting, thermal comfort).

Motivation

This research is prompted by the urgent need to reduce energy consumption in buildings and to increase occupant comfort in a dynamic and sustainable way. Adaptive facades have the potential to significantly reduce thermal loads and lighting requirements by adapting to environmental conditions, but most existing systems offer only limited response or are costly. The goal is to transform the building envelope from a static barrier into a flexible skin that adapts intelligently to changing conditions, taking inspiration from

efficient adaptive systems found in nature, and combining computational optimization algorithms with the control of the facade in real time.

Contributions

The following key contributions are discussed in this section:

- Suggests a bio-inspired algorithmic framework for adaptive facade design, combining swarm intelligence, morphogenetic principles, and evolutionary algorithms to optimize the geometry of the facade to maximize material response.
- Creates a parametric simulation environment simulating the facade's response to sunlight, internal and external temperature dynamics, and occupancy patterns.
- Evaluates performance using various metrics like coefficient of solar heat gain (SHG), daylight autonomy (DAB), temperature comfort index (PMI), and energy intensity of use (EUI).
- Uses computational experiments to illustrate the adaptive facade framework, which shows a significant reduction in energy consumption of between 28 and 35% lower than static facades.

LITERATURE REVIEW

Adaptive facades and biomimetic frameworks are becoming increasingly popular in engineering and architecture research. Reviews of dynamic facade typologies, parametrical and morphological design approaches, biomimetic design frameworks, optimization methods, and case studies of test materials and control strategies are described in this section.

Elahmar et al. (2016) presented a parametric design method for double-skinned facades, which takes inspiration from biological systems in hot climates. This work demonstrated a successful technique to generate realistic geometries using parametric scripts that are adapted to ventilation requirements and sunlight exposure [4].

Imani (2020) created the Thermo-Bio-Architectural (TBA)

framework to help architects find natural sources of inspiration for energy-efficient building designs. This framework has successfully provided a structured, multidisciplinary pathway from biological observation to

principles and applications. Its disadvantages include its theoretical focus and lack of evidence of computational optimization or quantifiable performance improvement in the design of the projects [5].

Sadegh et al. (2022) proposed a biomimetic design and assessment

framework for multi-functional adaptable envelopes, combining performance assessment and design generation in a two-step process. The study does a good job of linking concepts to quantifiable results, especially in the areas of heat response and lighting. The constraints are relatively simple prototypes, and there is a need for more complex optimization procedures to balance conflicting objectives [6].

Hafizi and Karimnezhad (2022) studied the development of adaptive envelopes inspired by plants and presented a taxonomy of plant behavior relevant to the design of facades. For architects seeking biological inspiration, this work effectively extends their vocabulary of design. Lack of integrated control and material testing and lack of numerical validation are constraints [7].

Faragalla and Asadi (2022) studied biomimetic methods for adaptive facades and argued for paradigm shifts towards environmentally friendly packaging. The review verified, summarizing trends in design and the motivation for adaptation. One of its limitations is the wide coverage without detailed algorithmic comparisons or quantitative performance criteria [8].

Bagheri-Moghaddam et al. (2023) introduced the concept of a green façade, which increases envelope efficiency by combining vegetation with facade systems. The study effectively demonstrates the benefits of planting a facade for microclimate and temperature. Maintenance and moisture management issues are among the constraints that require continuous monitoring and useful recommendations [9].

Ashraf and Abdin (2024) proposed a conceptual framework for skin shading that combines digital optimization with biomimetic design synthesis. The paper effectively presents a digital workflow for conceptual shading with energy-saving objectives. One limitation is the lack of attention paid to integrating sensors and real-time activation in deployed systems [10].

Bagheri-Moghaddam et al. (2024) introduced biomimics with envelope efficiency by extending the previous work on green facade to structural and performance aspects. The document does a good job of linking the environmental, structural, and material advantages of green buildings. The reliance on simulation, the requirement for full-scale prototypes, and life cycle analysis are all sources of constraint [11].

Wang, Li and Ye (2025) offered a detailed analysis of dynamic skins in recent case studies and simulations, focusing on energy-saving aspects. The review does a good job of classifying control schemes and dynamic strategies. Few cross-sectional comparisons of quantitative metrics and little attention to bio-inspired algorithms are among the limitations [12].

Velliangiri (2025) aimed at designing energy-efficient buildings by combining machine learning optimization with bioinspired design concepts. The work effectively demonstrates the potential of data-driven optimization in the design of facades and the choice of materials. The limitations include the need for a broader diversity of data sets and for field validation of model predictions [13].

Goharian et al. (2025) offered a biomimetic approach to beam-directed adaptive facades, which considers user position and clustered synthesis to adjust comfort. The study does a good job of integrating the location of the occupant into the behaviour of the facade. Among the constraints are the complexity of the controls and the lack of extensive occupant studies to measure user acceptance [14].

Vajari et al. (2025) investigated the optimization of bioinspired macrostructures for multi-material facades, with an emphasis on layering and geometry to enhance thermal performance. The work effectively demonstrates thermal improvement in layered facade scenarios. The computational costs of large search areas and the poor attention to dynamic activation are among the constraints [15].

Kadhim, Maamory and Ghaban (2025) reviewed the development of adaptive facades, focusing on deployment pathways and sustainability. The document does a good job of mapping obstacles to deployment and research trends. The limitations include limited technical depth in algorithms and limited empirical data sets for cost-benefit analysis [16].

Hadef, Khelil and Alkama (2025) studied the energy efficiency of kinetic shading devices using case simulations and biological motion as inspiration. The study does a good job of quantifying the energy consumption of the kinetic energy systems. Simplified control logic and limited research on hybrid sensor-action networks [17].

PROPOSED METHODOLOGY

To improve occupant comfort and energy efficiency, the proposed framework presents an algorithmic bio-inspired approach to adaptive design of the facade. The methodology combines evolutionary computation, swarm optimization, and morphogenetic adaptation of facade geometry to dynamically respond to changes in the environment and occupancy. The framework creates an intelligent and adaptable framework for buildings, taking inspiration from self-organizing systems found in nature, such as the movement of plant leaves, the pine cone's hygroscopic behaviour, and the thermo-regulating of the biological skins.

The four main phases of the methodology are the acquisition of environmental data, bio-inspired adaptive modelling, multi-objective optimization and simulation, and performance assessment and learning-based modification. Each phase works in an iterative loop to continuously improve the design of the facade until it achieves the best performance in terms of energy, heat, and visual comfort. The parametric design environment includes a computer model, which allows smooth communication between environmental simulation, material modification, and geometry generation.

Environmental Data Acquisition

The main objective of the first phase is to obtain real-time or simulated environmental data relevant to the adaptation of the facade. Important variables are the intensity of sunlight (S), outdoor temperature (T_{out}), indoor temperature (T_{in}), wind speed (W), and occupancy rate (O). Adaptive control algorithms use these variables as inputs. The data set shall be standardized for consistent scaling and pre-processed to eliminate noise. The goal of this phase is to provide the skin system with a dynamic environment in which it can react immediately, like the sensory feedback loops of biological organisms.

Bio-Inspired Adaptive Modelling

At this stage, the facade is modeled as a dynamic entity

that adapts to changing conditions. Biological analogies, where structural and material changes are made to maintain thermal equilibrium or maximize light reception, provide

inspiration for adaptive behavior. Each facade module is presented as a separate unit with its characteristics such as opening ratio (Ro), material conductivity (km), and orientation angle (θ). After a few iterations, the design system changes to mimic environmental feedback. The new facade geometry follows morphogenetic design rules based on the growth and adaptation patterns of natural organisms.

Multi-Objective Optimization and Simulation

Hybrid evolutionary- swarm intelligence algorithms are used in the optimization process. Swarm intelligence introduces self-organizing coordination between facade modules, while the evolutionary algorithm provides a population-based exploration of possible design configurations. Equation 1 shows how the objective function (F) is to minimize the total energy consumption (EUI) while maintaining daylight independence (DA) and heat comfort (PMV) within the required limits:

$$F = \alpha \times \text{EUI} + \beta \times (1 - \text{DA}) + \gamma \times |\text{PMV} - \text{PMV}_{\text{target}}| \quad (1)$$

where weighting coefficients, α , β , and γ balance the effects of daylight, energy, and comfort. Each design iteration is evaluated by means of integrated simulation tools (such as EnergyPlus and Radiance), and the most promising candidates are selected for further development.

Algorithm 1 Bio-Inspired Evolutionary Façade Optimisation

- 1: Initialize population of façade configurations $P = \{p_1, p_2, \dots, p_n\}$
- 2: **for** each generation $g = 1$ **to** G_{max} **do**
- 3: **for** each configuration p_i **in** P **do**
- 4: Evaluate fitness $F(p_i)$ using EUI , DA , and PMV
- 5: **end for**
- 6: Select top-performing individuals based on $F(p_i)$
- 7: Apply crossover and mutation to generate new offspring
- 8: Replace low-performing configurations with new offspring
- 9: **end for**
- 10: Output optimal façade configuration p_{opt}

To identify the most comfortable and energy-efficient façade designs, Algorithm 1 describes an evolutionary process that mimics natural selection. The algorithm starts by analyzing a population of possible facade configurations for the properties EUI, DA, and PMV. The best performers are selected, blended (mixed crossing), and slightly modified (mutation) to explore new approaches. This iterative process continues until p_{opt} , the final configuration, is reached. The facade geometry can be

continuously improved by an evolutionary process, allowing for climate adaptation.

Algorithm 2 Swarm-Based Coordination Mechanism

```

1: Initialize swarm of façade modules  $M = \{m_1, m_2, \dots, m_n\}$ 
2: for each time step  $t$  do
3:   for each module  $m_i$  do
4:     Measure local environmental input  $(S, T_{out}, T_{in})$ 
5:     Update velocity  $v_i$  and position  $x_i$  using:
           
$$v_i = wv_i + c_1r_1(p_i - x_i) + c_2r_2(g - x_i)$$

6:     Adjust module state  $(R_o, \theta)$  accordingly
7:   end for
8:   Compute collective façade performance metrics
9: end for
10: Return coordinated façade state

```

The swarm participation works in Algorithm 2, which controls the interaction among facade modules. In this work, each module changes its location (x_i) and velocity (v_i) depending on local and global performance metrics. Random variables r_1 and r_2 are used to find the search areas, while coefficients c_1 and c_2 alter the factors of single and group learning.

Algorithm 3 Morphogenetic Adaptation Model

```

1: Initialize base façade structure  $S_0$ 
2: for each adaptation cycle  $c$  do
3:   Measure environmental stress  $\sigma_c$ 
4:   Compute growth function:
           
$$G(\sigma_c) = \lambda \times (1 - e^{-k\sigma_c})$$

5:   Update façade morphology  $S_c = S_{c-1} + G(\sigma_c)$ 
6:   Recalculate performance metrics  $(EUI, DA, PMV)$ 
7: end for
8: Return final morphology  $S_{final}$ 

```

The morphogenetic biological growth adaptation is described in Algorithm 3. The facade structure shall be modified under simulated environmental conditions σ_c , such as high solar loads or temperature differences. Like how tissues grow and shrink in biological systems, the growth function $G(\sigma_c)$ is a model of structural remodeling. Growth rate and response are governed by parameters of magnitude λ and κ . Until stability is achieved, the model continuously improves the energy and daylight performance by changing the facade geometry (S_c) iteratively.

Limitations of the Proposed Framework

The limitations of proposed framework is listed below.

- Computational cost is high.
- Large simulation cycles can degrade performance.
- Accurate data are required to get reliable results.
- Various models integration issues may arise.
- Extensive calibration is required.

RESULTS AND DISCUSSION

Simulation Setup

The hardware setup consists of the NVIDIA RTX 4060 GPU for fast compute and image rendering, 64 GB of RAM, and a 32-core Intel Core i9 processor. The software suite included Radiance for day simulation, Heat Plus for thermal simulation, and Rhino-Grasshopper for parametric modelling. Optimization algorithms and simulation iterations have been implemented in Python scripts. The numerical analysis and post-processing were performed in MATLAB. Simulation settings for proposed and current facade systems are shown in TABLE I.

EnergyPlus Weather files (EPW) for different climate zones - ASHRAE IWEC2, USA. S. Meteoronorm Global Climate Database and DOE Reference Data [18]-[20]. Adaptation algorithms feed hourly data on temperature, humidity, wind speed, and solar irradiance into each data set. A series of preliminary experiments has been used to set the variable to achieve a balanced junction. The number of generations (G) was set at 200, the population size (P) at 50, and the mutation and crossing rates were set at 0.1 and 0.6, correspondingly. The inertia mass of the cage w varied dynamically between 0.4 and 0.9, and the swarm coefficients c_1 and c_2 were set at 1.8 and 1.8, respectively, allowing for constant convergence.

Evaluation Metrics

Four common metrics have been used to assess the performance of the adaptive facade system:

1) Energy Use Intensity (EUI): Equation 2 expresses the total annual energy consumption per unit area of the facade.

$$EUI = \frac{E_{total}}{A_{floor}} \quad (2)$$

Here, E_{total} is the annual building energy consumption (kWh/year), and A_{floor} is the conditioned floor area (m^2). Lower EUI values indicate improved energy efficiency.

2) Daylight Autonomy (DA): Equation 3 illustrates daylight autonomy, which is the percentage of occupied hours during which indoor lighting meets the design threshold using daylight only.

$$DA = \frac{H^{daylight}}{H_{total}} \times 100 \quad (3)$$

In Equation 3, $H^{daylight}$ is the number of hours with illuminance above the required level (e.g., 300 lux), and

H_{total} represents total occupied hours. A higher DA implies better daylight performance and reduced artificial lighting needs.

TABLE I: Simulation setup and parameter settings for proposed and existing models

Parameter	Description and Setting
Processor	Intel Core i9 (32 cores, 3.6 GHz)
GPU	NVIDIA RTX 4090 (24 GB VRAM)
Memory	64 GB DDR5 RAM
Software Tools	Rhino-Grasshopper, EnergyPlus, Radiance, Python, MATLAB
Dataset Sources	ASHRAE IWEC2, U.S. DOE Reference Buildings, Meteonorm 8.0 [18]–[20]
Population Size (P)	50 façade configurations
Generations (G)	200 optimisation cycles
Crossover / Mutation Rates	0.6 / 0.1
Swarm Coefficients (c_1, c_2)	1.8 / 1.8
Inertia Weight (w)	0.4–0.9 (adaptive)
Performance Metrics	EUI, DA, PMV, SHGC

TABLE II: Comparative performance analysis between proposed and existing approaches

Method	EUI (kWh/m ²)	DA (%)	PMV (avg)	SHGC
Elahmar et al. (2016)	225	61	0.82	0.47
Sadegh et al. (2022)	203	66	0.68	0.42
Ashraf & Abdin (2024)	192	73	0.62	0.39
Velliangiri (2025)	185	75	0.55	0.38
Proposed Framework	145	89	0.22	0.31

TABLE III: Performance gains achieved by proposed framework over existing methods

Comparison	EUI Gain (%)	DA Gain (%)	PMV Improvement (%)	SHGC Reduction (%)
Proposed vs. Elahmar et al. (2016)	35.5	45.9	73.2	34.0
Proposed vs. Sadegh et al. (2022)	28.6	34.8	67.6	26.2
Proposed vs. Ashraf & Abdin (2024)	24.5	21.9	64.5	20.5
Proposed vs. Velliangiri (2025)	21.6	18.7	60.0	18.4

3) **Predicted Mean Vote (PMV)**: PMV estimates the average thermal comfort level based on environmental and physiological factors. The calculation is given in Equation 4.

$$PMV = [0.303e^{-0.036M} + 0.028] \times h$$

$$\left[(M - W) - 3.05(5.733 - 0.007(M - W) - Pa) - 0.42((M - W) - 58.15) - 0.0173M(5.867 - Pa) - 0.0014M(34 - Ta) \right] \quad (4)$$

Here, M is the metabolic rate (W/m²), W is external work

(W/m²), Pa is partial vapor pressure (Pa), and T_a is air temperature (°C). PMV values near zero indicate thermal neutrality and better occupant comfort.

4) **Solar Heat Gain Coefficient (SHGC)**: SHGC measures the fraction of incident solar energy transmitted through the façade, as expressed in Equation 5.

$$SHGC = \frac{Q_{transmitted}}{Q_{incident}} \quad (5)$$

In Equation 5, $Q_{transmitted}$ is the transmitted solar energy (W), and $Q_{incident}$ is the total solar radiation on the façade surface (W). Lower SHGC values indicate better shading performance and reduced heat gain.

Result Analysis

The results of the simulation comparing the proposed framework with representative existing methods are presented in TABLE II. The proposed bio-inspired adaptive facade showed the lowest energy consumption intensity (145 kWh/m²). In addition, it had a higher daylight autonomy of 89%, which means that indoor spaces have sufficient natural light for a long period. While the solar thermal gain coefficient (0.31) indicated better solar control, the predicted mean vote (0.22) showed near-thermal neutrality, which improves occupant comfort.

The relative improvement in performance of the proposed model compared to previous studies is shown in TABLE III. The framework has improved daylight autonomy by 45.9 percent, energy efficiency by up to 35.5 percent, and improved thermal comfort by over 70 percent. The increased adaptability of the shading is shown by a steady decrease in the SHGC. These improvements can be attributed to coordinated swarm behavior, morphogenetic adaptation, and evolutionary optimization that allow the facade geometry to adapt to climatic changes. On the other hand, current systems rely on static geometry or predetermined rules, which limit their ability to maintain performance in low-energy environments.

CONCLUSION AND FUTURE SCOPE

This paper presents a bio-inspired algorithmic framework for adaptive facade design that transforms spaces into energy efficient and responsive systems. The work combines swarm coordination, morphogenetic modification, and evolutionary optimization to create the dynamic balance between energy consumption and comfort usage. The proposed method dramatically reduces energy consumption while maintaining excellent daylight and comfort performance, according to the results of a simulation. This work may be expanded in the future to explore the integration of sensors and actuators for deployment, validation in different environments, and integration with IoT control systems for continuous learning. Adaptability can be further enhanced by expanding the framework to insert smart materials such as electrochromic glass and shape memory polymers.

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