

A Quantitative Image Analysis Model for Evaluating Fiber Diameter Distribution & Surface Porosity of Electrospun Fibers

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1 Introduction

Lignin, the most abundant aromatic biopolymer on Earth, is a premier **renewable carbon source** for next-generation energy-storage electrodes. Its high carbon content and intrinsic polyphenolic structure make it an ideal "green-energy" precursor [1].

Through **electrospinning** followed by **CO₂ laser-induced graphitization**, lignin is converted into 3D porous, graphene-like scaffolds — a sustainable alternative to petroleum-based polymers [2].

The central challenge: bridging **synthesis** and **predictive supercapacitor performance**. Here we implement a **computational optimization framework** that quantifies structural descriptors — fiber diameter, porosity, orientation — directly from SEM imagery, enabling high-fidelity correlation between geometric design and electrochemical behaviour.

41.42 %

Global surface porosity

~500

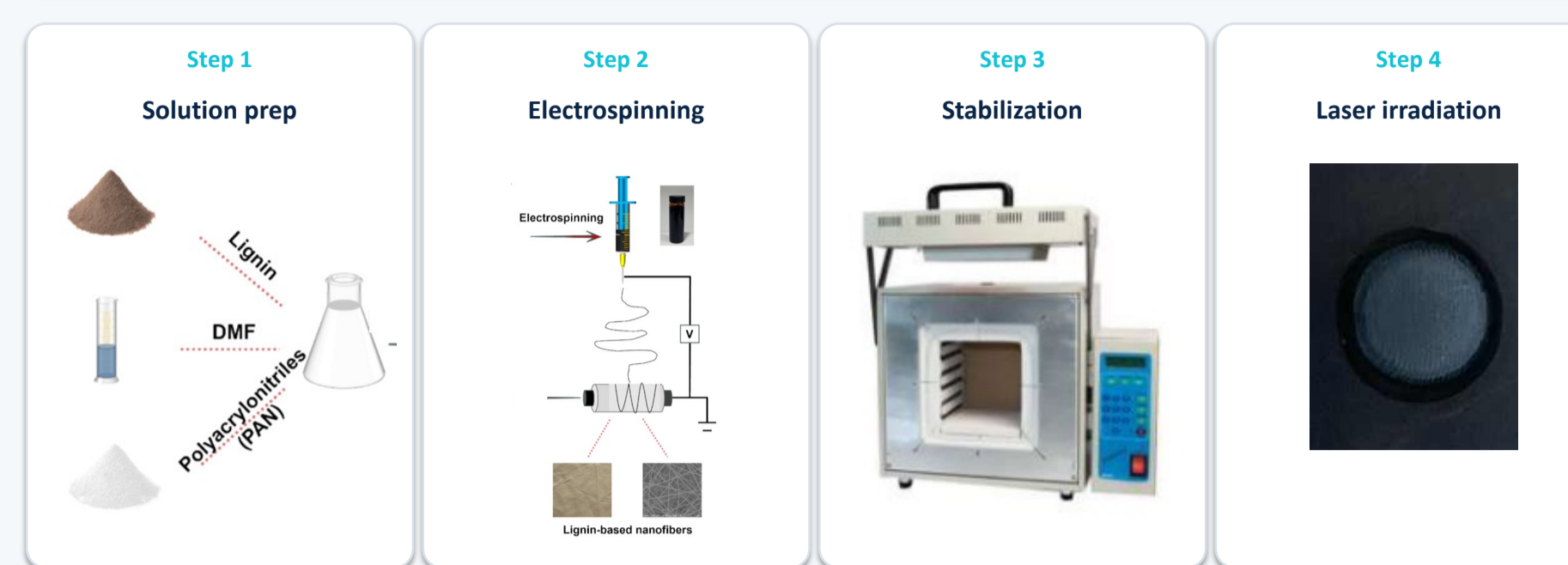
Watershed segments

320.7 nm

Mean fiber diameter

2 Methodology

EXPERIMENTAL — Fiber Production



→ In-situ graphitization via CO₂ laser (10.6 μm) → 3D porous graphene-like electrode

Binder-free Lignin-PAN carbon-fiber electrode

Flexible monolith ready for 3-electrode testing.

COMPUTATIONAL — Image Framework



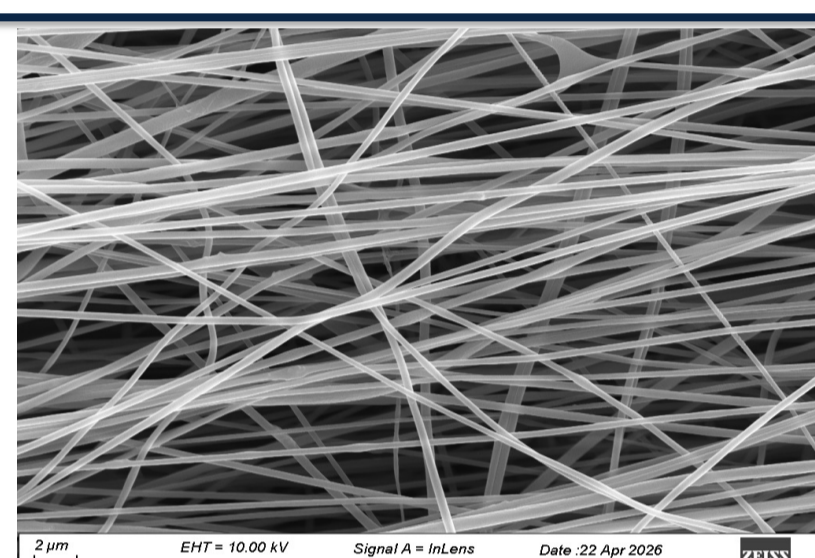
Integrated structural quantification

A custom pipeline couples SEM with adaptive segmentation, skeleton analysis and watershed isolation — yielding **high-fidelity descriptors** of diameter, porosity & orientation directly linked to electrochemistry.

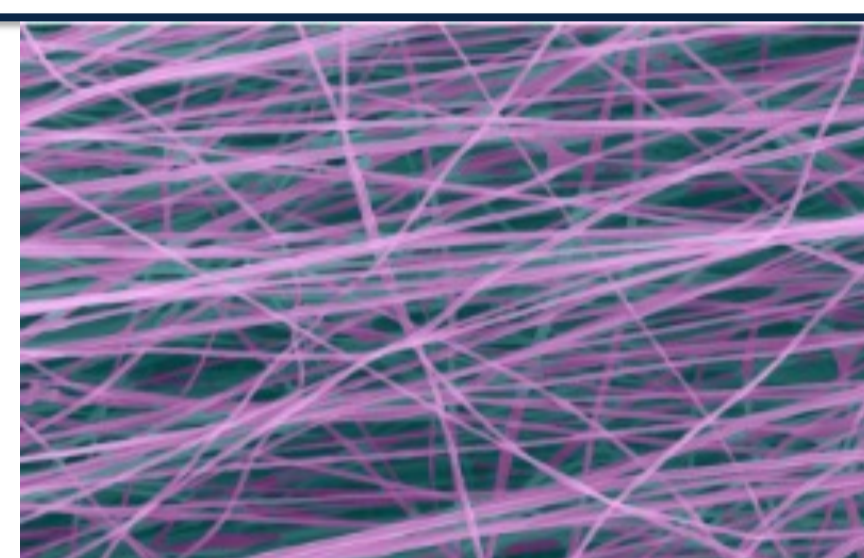
RESULTS

Integrated physicochemical, electrochemical & morphological analysis of Lignin-PAN carbon-fiber electrodes

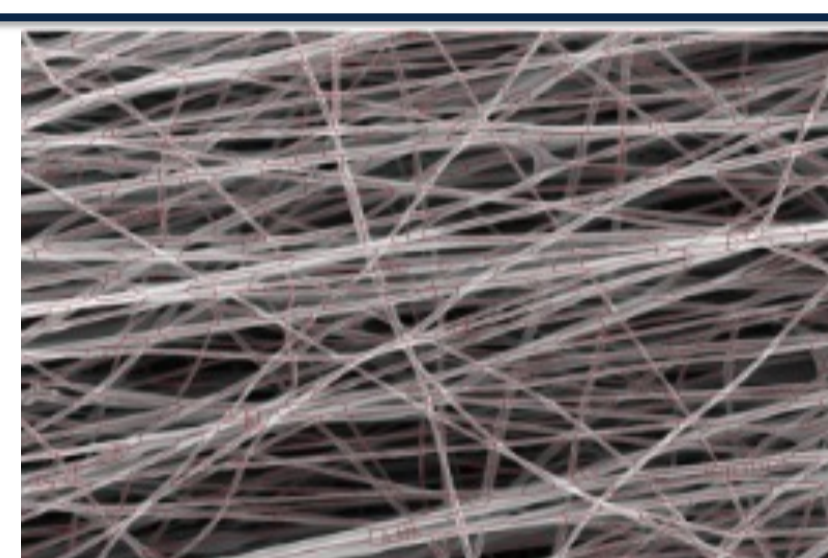
3 Quantitative Morphological Mapping · SEM and Image based Model



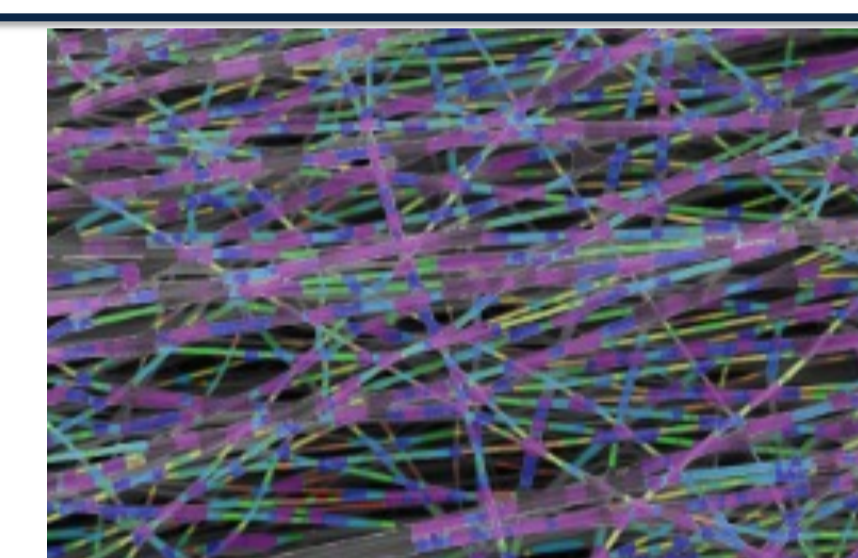
Raw SEM micrograph



Adaptive thresholding



Skeletonization



Watershed segmentation

Fiber diameter distribution

Statistical analysis across 16 independent fibers (N = 10,528 segments)

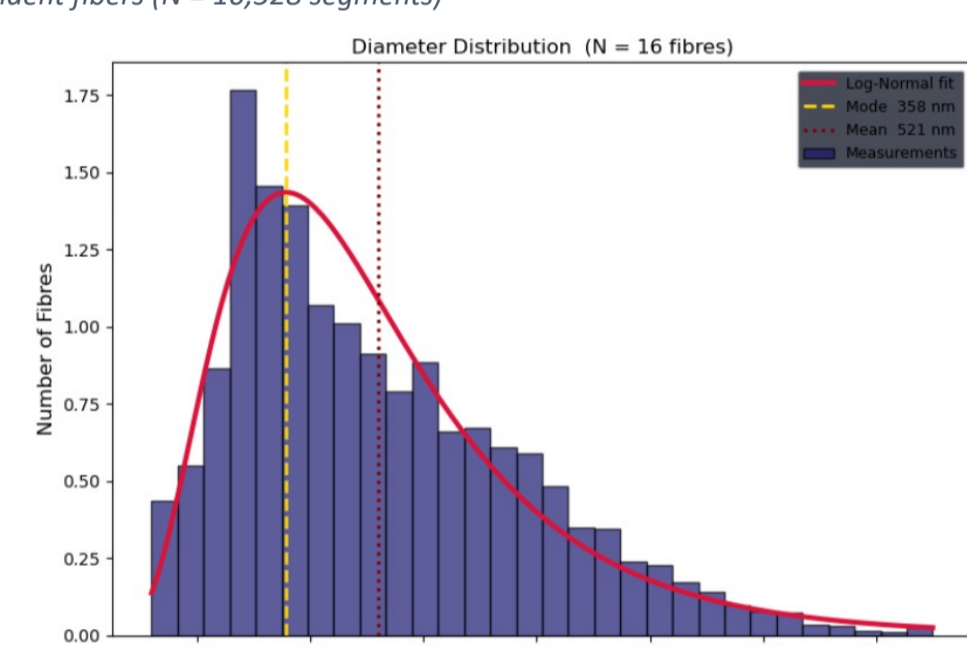


Figure 1 — Diameter histogram & log-normal fit

Fiber orientation analysis

Polar rose plot — randomness index R = 1.000

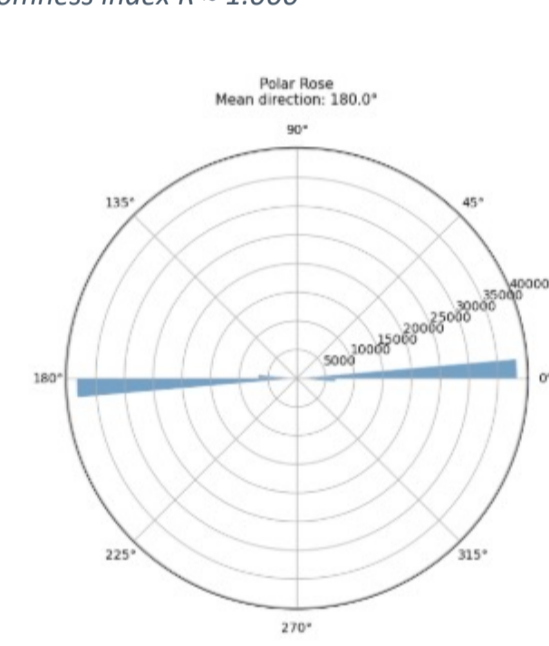


Figure 2 — Orientation rose plot

Key morphological metrics

Watershed segments
10,528
Independent fibers
16
Mean E SD
520.7 ± 256.7 nm
Mode (Log-Normal)
357.5 nm
Surface porosity
41.42 %
Mean pore ECD
980.9 nm

The computational image analysis pipeline developed in this study was successfully validated on as-spun electrospun Lignin-PAN fibers to establish a baseline for static morphological properties.

Future work will focus on evolving this framework into dynamic modeling simulations to track structural transformations in real-time.

The fiber networks will undergo thermal stabilization and subsequent carbonization processes; the algorithm will be extended to monitor the corresponding evolution of fiber shrinkage, pore closure, and orientation shifts throughout these thermal treatments.

The pipeline will be upgraded to support high-throughput batch processing, enabling the automated, simultaneous evaluation of large image datasets across various processing timeframes to ensure statistical significance.

Surface porosity across SEM samples

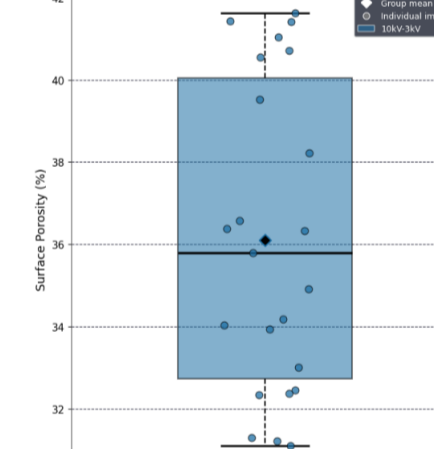


Figure 3 — Surface porosity boxplot

Robust architecture

41.42% global porosity ensures unhindered electrolyte access and rapid ion diffusion through the electrode.

Statistical reliability

Reported values backed by 10,528 watershed segments across 16 independent skeleton components.

Beyond qualitative SEM

Custom pipeline detects structural bottlenecks invisible to manual SEM screening — a diagnostic tool, not just visualization.

4 Phase Validation & Electrochemical Performance · 3-electrode setup

Raman spectra — Lignin-PAN

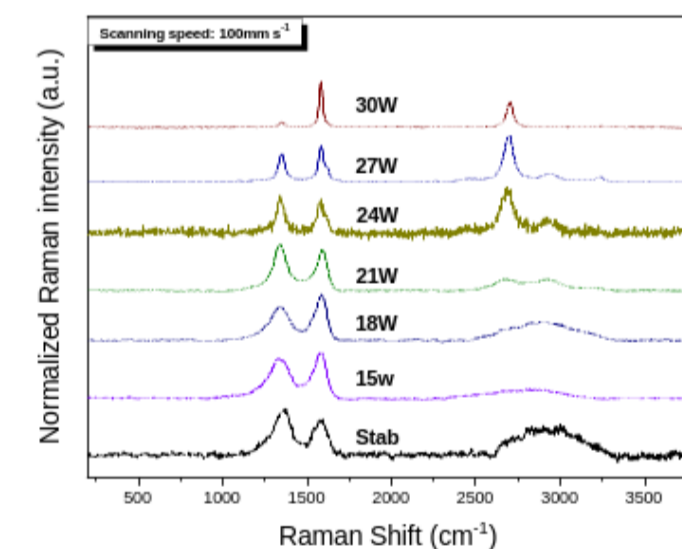


Figure 4 — Raman spectra — Lignin-PAN

Dominant D, G & 2D bands; different laser powers indicate different graphitization levels.

Lorentzian deconvolution

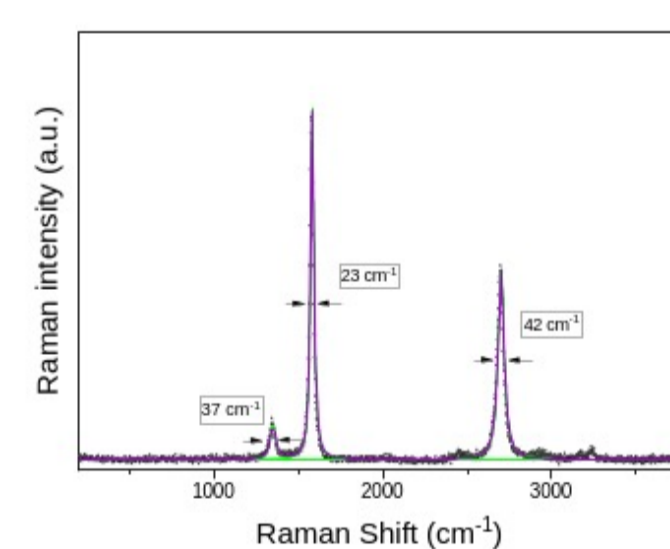


Figure 5 — Lorentzian deconvolution

Single Lorentzian 2D band → turbostratic arrangement of graphene layers.

Sheet resistance vs. laser power

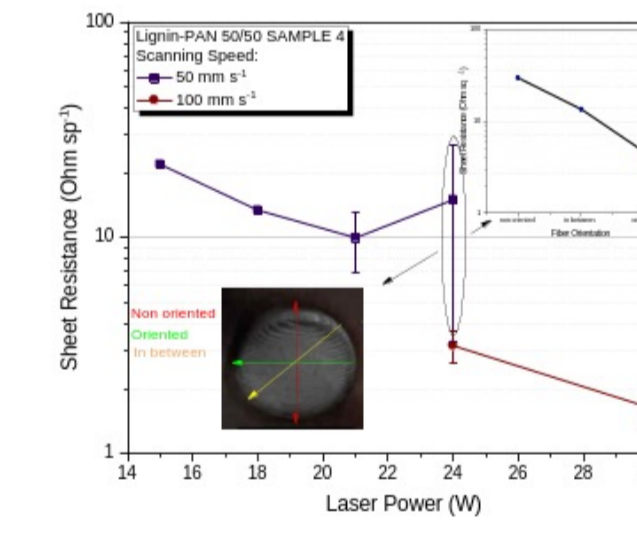


Figure 6 — Sheet resistance vs. laser power

Different laser parameters yield different graphitization & conductivity profiles.

Cyclic voltammetry

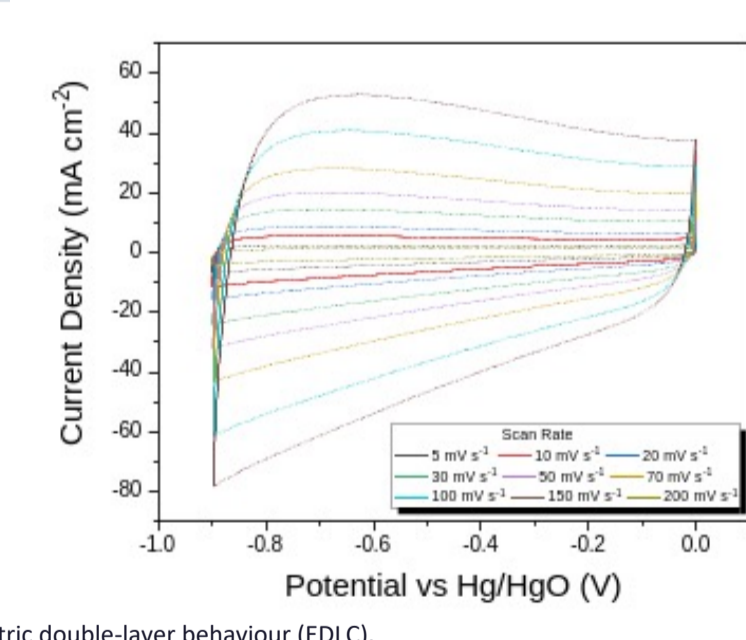


Figure 7 — Cyclic voltammetry

Quasi-rectangular CV → predominant electric double layer behaviour (EDLC).

Cycling / rate performance

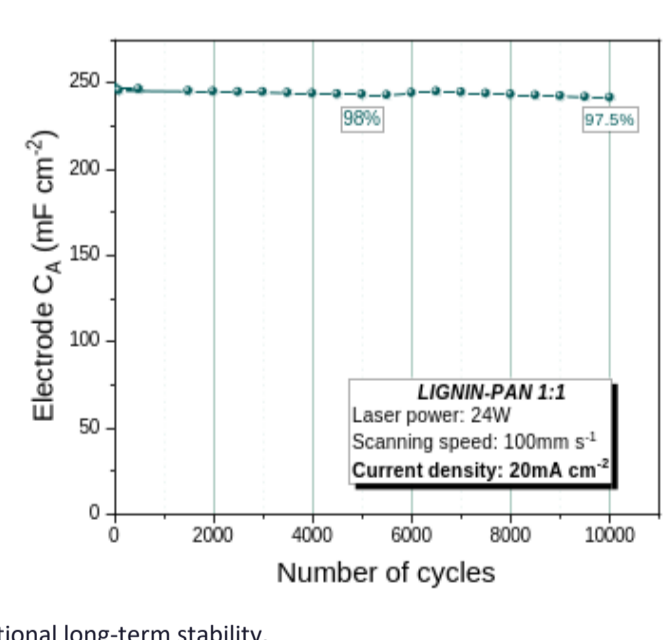


Figure 8 — Cycling / rate performance

Negligible capacitance fade after 10,000 cycles — exceptional long-term stability.

Electrochemical Impedance Spectroscopy

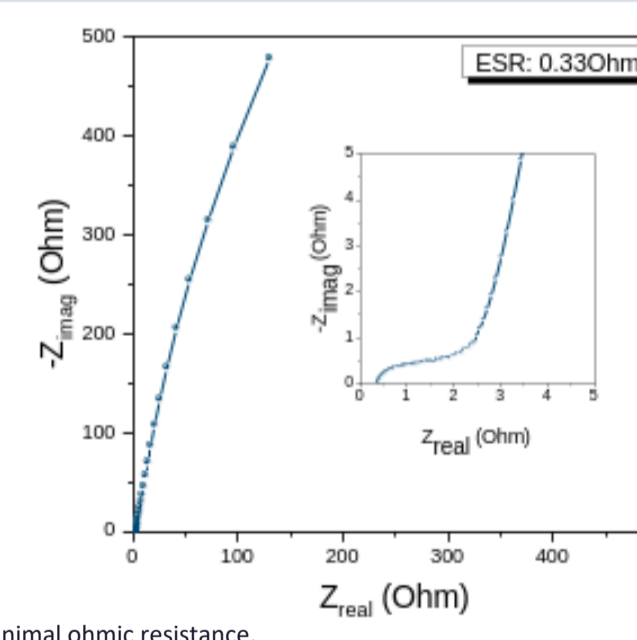


Figure 9 — Electrochemical Impedance Spectroscopy

Very low ESR (0.33 Ω sq⁻¹) — excellent conductivity & minimal ohmic resistance.

310 mF cm⁻²
Areal Electrode Capacitance at Scan Rate 5 mV s⁻¹
0.33 Ω sq⁻¹

Equivalent Series Resistance

97.5 %

Capacitance Retention after 10 000 cycles

EDLC

dominant charge-storage mechanism

5 Conclusions

1 Quantitative pipeline

Custom image-analysis framework reveals structural descriptors beyond qualitative SEM screening.

2 Optimised architecture

41.42% surface porosity → unhindered electrolyte access and rapid ion diffusion.

3 Statistical robustness

10,528 watershed segments across 16 independent components confirm representative metrics.

4 EDLC behaviour

Areal capacitance vs. scan rate confirms predominant electric double-layer mechanism.

5 Low ESR

0.33 Ω sq⁻¹ ESR reflects excellent conductivity & minimal ohmic resistance.

6 Outstanding cycling

Negligible capacitance fade over 10,000 cycles — strong candidate for sustainable supercapacitors.

6 References

- [1] Sahoo et al. (2024). Lignin-based materials for sustainable energy storage. *Renewable and Sustainable Energy Reviews*.
- [2] Ye et al. (2023). Laser-induced graphene from renewable precursors: progress and perspectives. *Advanced Materials*.
- [3] Edinburgh Instruments. What is Raman spectroscopy? edinst.com/blog/what-is-raman-spectroscopy.
- [4] Mevada, C., Mukhopadhyay, M. (2023). Handbook of Nanocomposite Supercapacitor Materials IV. Springer Series in Materials Science, vol 331. Springer, Cham.
- [5] Sinha, P., Kar, K.K. (2020). Introduction to Supercapacitors. In: Kar, K. (eds) Handbook of Nanocomposite Supercapacitor Materials II. Springer Series in Materials Science, vol 302. Springer, Cham.