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# Keeping It Fresh: New AI-based Strategy Can Assess the Freshness of Beef Samples

Scientists combine spectroscopy and deep learning in an efficient technique for detecting spoiled meat

Scientists at Gwangju Institute of Science and Technology, Korea, combine an inexpensive spectroscopy technique with artificial intelligence to develop a new way of assessing the freshness of beef samples. Their method is remarkably quicker and more cost-effective than conventional approaches while maintaining a relatively high accuracy, paving the way for mass-produced devices to identify spoiled meat both in the industry and at home.



Consuming spoiled beef is dangerous, but there are currently no simple and efficient methods to assess beef freshness. Photo courtesy: Unsplash

Although beef is one of the most consumed foods around the world, eating it when it's past its prime is not only unsavory, but also poses some serious health risks. Unfortunately, available methods to check for beef freshness have various disadvantages that keep them from being useful to the public. For example, chemical analysis or microbial population evaluations take too much time and require the skills of a professional. On the other hand, non-destructive approaches based on near-infrared spectroscopy require expensive and sophisticated equipment. Could artificial intelligence be the key to a more cost-effective way to assess the freshness of beef?

At Gwangju Institute of Science and Technology (GIST), Korea, a team of scientists led by Associate Processors Kyoobin Lee and Jae Gwan Kim have developed a new strategy that combines deep learning with diffuse reflectance spectroscopy (DRS), a relatively inexpensive

optical technique. "Unlike other types of spectroscopy, DRS does not require complex calibration; instead, it can be used to quantify part of the molecular composition of a sample using just an affordable and easily configurable spectrometer," explains Lee. The findings of their study are now <u>published in Food Chemistry</u>.

To determine the freshness of beef samples, they relied on DRS measurements to estimate the proportions of different forms of myoglobin in the meat. Myoglobin and its derivatives are the proteins mainly responsible for the color of meat and its changes during the decomposition process. However, manually converting DRS measurements into myoglobin concentrations to finally decide upon the freshness of a sample is not a very accurate strategy—and this is where deep learning comes into play.

Convolutional neural networks (CNN) are widely used artificial intelligence algorithms that can learn from a pre-classified dataset, referred to as 'training set,' and find hidden patterns in the data to classify new inputs. To train the CNN, the researchers gathered data on 78 beef samples during their spoilage process by regularly measuring their pH (acidity) alongside their DRS profiles. After manually classifying the DRS data based on the pH values as 'fresh,' 'normal,' or 'spoiled,' they fed the algorithm the labelled DRS dataset and also fused this information with myoglobin estimations. "By providing both myoglobin and spectral information, our trained deep learning algorithm could correctly classify the freshness of beef samples in a matter of seconds in about 92% of cases," highlights Kim.

Besides its accuracy, the strengths of this novel strategy lie in its speed, low cost, and nondestructive nature. The team believes it may be possible to develop small, portable spectroscopic devices so that everyone can easily assess the freshness of their beef, even at home. Moreover, similar spectroscopy and CNN-based techniques could also be extended to other products, such as fish or pork. In the future, with any luck, it will be easier and more accessible to identify and avoid questionable meat.

#### Reference

Authors:	Sungho Shin <sup>1</sup> , Youngjoo Lee <sup>2</sup> , Sungchul Kim <sup>2</sup> , Seungjun Choi <sup>1</sup> , Jae Gwan Kim <sup>2</sup> *, Kyoobin Lee <sup>1</sup> *
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Affiliations:	<sup>1</sup> School of Integrated Technology, Gwangju Institute of Science and Technology (GIST)
	<sup>2</sup> Department of Biomedical Science & Engineering, Gwangju Institute of Science and Technology (GIST)

\*Corresponding authors' emails: jaekim@gist.ac.kr (J.G. Kim), kyoobinlee@gist.ac.kr (K. Lee)

## About Gwangju Institute of Science and Technology (GIST)

Gwangju Institute of Science and Technology (GIST) is a research-oriented university situated in Gwangju, South Korea. One of the most prestigious schools in South Korea, it was founded in 1993. The university aims to create a strong research environment to spur advancements in science and technology and to promote collaboration between foreign and domestic research programs. With its motto, "A Proud Creator of Future Science and Technology," the university has consistently received one of the highest university rankings in Korea.

Website: http://www.gist.ac.kr/

# About the authors

Kyoobin Lee is an Associate Professor and Director of the AI laboratory at GIST. His group is developing AI-based robot vision and deep learning-based bio-medical analysis methods. Before joining GIST, he obtained a PhD in Mechatronics from KAIST and completed a postdoctoral training program at Korea Institute of Science and Technology (KIST).



Jae Gwan Kim is an Associate Professor at the Department of Biomedical Science and Engineering at GIST since 2011. His current research topics include brain stimulation by transcranial ultrasound, anesthesia depth monitoring, and screening the stage of Alzheimer's disease via brain functional connectivity measurements. Before joining GIST, he completed a postdoctoral training program at the Beckman Laser Institute and Medical Clinic at UC Irvine, USA. In 2005, he received a PhD in Biomedical Engineering from a joint program between the University of Texas at Arlington and the University of Texas Southwestern Medical Center at Dallas, USA.

