

探討光學檢測豬肉中揮發性鹽基態氮含量

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一、前言

二、使用光散射成像技術檢測揮發性鹽基態氮含量

三、以高光譜系統為基礎的多成像技術檢測豬肉中揮發性鹽基態氮含量

四、利用近紅外光多光譜成像系統檢測豬肉揮發性鹽基態氮含量

五、結論

摘要

傳統檢測揮發性鹽基態氮 (Total volatile basic nitrogen, TVB-N) 作為肉品新鮮度的指標，一般使用化學分析的方法，但檢測耗時，在快速生產線上，傳統的檢測方法已不適用，故本篇比較各項光學成像系統，以找尋最適用之肉品新鮮度光學檢測法。利用光散射成像技術，從感興趣區域 (Region Of Interest, ROI) 提取特徵變數，接著使用正交線性判別分析 (Algorithm adaptive boosting Orthogonal Linear Discriminant Analysis, AdaBoost-OLDA) 建立模組，與線性判別 (Linear Discriminant Analysis, LDA) 和支持向量機 (Support Vector Machine, SVM) 做比較，結果顯示 AdaBoost-OLDA 達到 100% 校準及預測的分類率，皆優於 LDA 和 SVM。ROI 選定區域後，從高光譜圖像 (Hyperspectral Imaging, HSI) 得到質地分析，再使用最小二乘法預測 TVB-N，並比較二維主成分分析 (2D principal component analysis, 2DPCA)、主成分分析 (principal component analysis, PCA)，結果得到 2DPCA 的預測相關係數與預測均方根誤差分別為 0.995 與 1.86 mg/100g，相較於 PCA 分析法證明 2DPCA 更能展現非破壞性檢測豬肉 TVB-N 含量。利用近紅外線 (near infrared, NIR) 多光譜成像技術 (multispectral imaging technique, MSI) 在不同波長光源下進行特徵成像，並和人工神經網路 (modeling-back propagation artificial neural network, BP-ANN) 及 AdaBoost 演算法 BP-AdaBoost 並比較。綜上，以光學成像系統搭配不同演算法檢測 TVB-N，相較於傳統的檢測方式更加快速準確，在市場上的使用將更有潛力。

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