# A rapid machine-learning approach for detecting fish species and body parts

using rapid evaporative ionisation mass spectrometry

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# GP [1, 2, 3] inspired by reproductive behaviour of animals





## ECRG, VUW, New Zealand





## **Topics**

- Catfishing
- 2 Fish Oil
- Mass Spectrometry
- 4 Classification
- Transformer
- 6 Intepretable



# Have you been catfished? [4]



#### Popular restaurant accused of serving cheap Vietnamese catfish to customers who thought they were getting Australian dory

- · A Melbourne restaurant has been accused of serving catfish to customers
- · Hunky Dory has allegedly been selling frozen fillets of basa as dory
- · Owner Greg Robotis has denied allegations he is misleading customers
- The City of Port Phillip is investigating Hunky Dory's Port Melbourne store

By HARRY PEARL FOR DAILY MAIL AUSTRALIA PUBLISHED: 14:31 AEDT, 27 May 2016 | UPDATED: 16:08 AEDT, 27 May 2016

















A Melbourne restaurant has been accused of serving a Vietnamese catfish to customers who believe they are ordering Dory.

A whistleblower has alleged that Hunky Dory outlets have been selling frozen fillets of basa, a species of catfish native to the Mekong basin, as fish-of-the-day dory, The Age reports.

Owner Greg Robotis has denied the claims and said inexperienced staff may have been calling the fish the wrong name.



## Catfishing [4], Mislabelling [5], and Quality Assurance [6]

<b>Nutrition F</b>	acts
6 servings per container Serving size 4-5 ound	ces(187g
Amount per serving Calories	200
% !	Daily Value
Total Fat 5g	69
Saturated Fat 0.5g	39
Trans Fat 0g	
Cholesterol 80mg	279
Sodium 610mg	279
Total Carbohydrate 10g	49
Dietary Fiber 0g	09
Total Sugars 3g	
Includes 0g Added Sugars	09
Protein 27g	
Vitamin D 2mcg	109
Calcium 79mg	69
Iron 3mg	159
Potassium 519mg	109

<sup>\*</sup>The % Daily Value tells you how much a nutrient in a serving of food contributes to a daily diet. 2,000 calories a day is used for general nutrition advice.

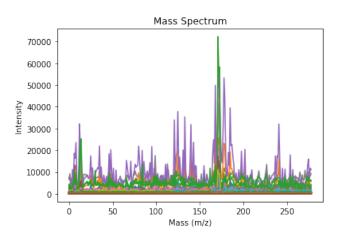


# Fish oil is brain food! [7, 8]





# Fish oil analyzed with Mass Spectrometry! [6]





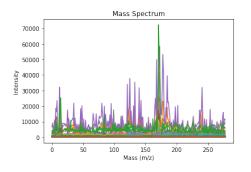


# Fish oil analysis can't be blackbox! [9, 10]





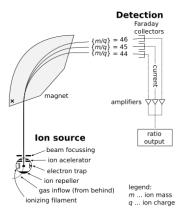
#### Mass Spectrometry [11, 6, 12] $\approx$ Chemical Fingerprint





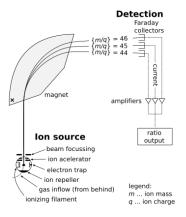


- Laser Pen
- 2 Vacuum
- Selectromagnetic Field (EMF)
- Operation
  Operation



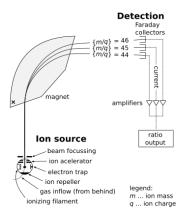


- Laser Pen
- Vacuum
- Selectromagnetic Field (EMF)
- 4 Detector



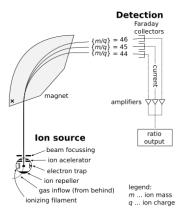


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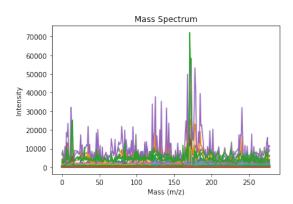


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- Laser Pen
- Vacuum
- Electromagnetic Field (EMF)
- Operation



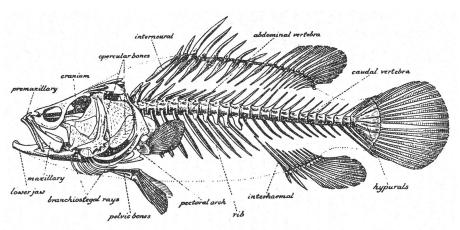


#### Fish Species





#### Fish Body Parts

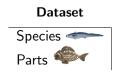




fillets, heads, livers, skins, guts and frames



#### Classification: Datasets





#### Classification: Methods

Dataset	Method
Species Parts	RF [13] KNN [14] DT [15] NB [16] LR [17] SVM [18] LDA [19] Ensemble [20] Transformer [21, 22]
	MCIFC [2, 3]



## Classification: Fish Species

Dataset	Method	Train	Test
Species <	RF [13]	$100.0\% \pm 0.00\%$	95.88% ± 4.47%
	KNN [14]	$93.24\% \pm 2.43\%$	$83.69\% \pm 6.91\%$
	DT [15]	$100.0\% \pm 0.00\%$	$99.13\% \pm 1.72\%$
	NB [16]	$100.0\% \pm 0.00\%$	$87.97\% \pm 9.57\%$
	LR [17]	$100.0\% \pm 0.00\%$	$96.72\% \pm 4.75\%$
	SVM [18]	$100.0\% \pm 0.00\%$	$95.97\% \pm 5.06\%$
	LDA [19]	$98.67\% \pm 0.77\%$	96.47% ± 3.67%
	Ensemble [20]	$100.0\% \pm 0.00\%$	$98.16\% \pm 3.00\%$
	Transformer [21, 22]	$100.0\%\pm0.00\%$	99.58% ± 1.31%
	MCIFC [2, 3]	$99.97\% \pm 0.15\%$	$94.72\% \pm 10.25\%$



#### Classification: Fish Body Parts

Dataset	Method	Train	Test
Parts 🗪	RF [13]	$100.0\% \pm 0.00\%$	$40.00\% \pm 15.27\%$
	KNN [14]	$42.88\% \pm 5.37\%$	$31.66\% \pm 14.49\%$
	DT [15]	$100.0\% \pm 0.00\%$	$27.22\% \pm 13.25\%$
	NB [16]	$100.0\% \pm 0.00\%$	$45.00\% \pm 15.60\%$
	LR [17]	$100.0\% \pm 0.00\%$	$56.66\% \pm 15.27\%$
	SVM [18]	$100.0\% \pm 0.00\%$	$56.11\% \pm 14.58\%$
	LDA [19]	$75.61\% \pm 3.20\%$	$45.55\% \pm 16.06\%$
	Ensemble [20]	$100.0\% \pm 0.00\%$	$51.66\% \pm 15.72\%$
	Transformer [21, 22]	$100.0\%\pm0.00\%$	63.33% ± 24.59%
	MCIFC [2, 3]	$97.93\% \pm 1.59\%$	$55.83\% \pm 18.97\%$

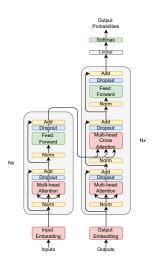


## Classification: Avoid Catfishing [4] & Mislabelling [5]



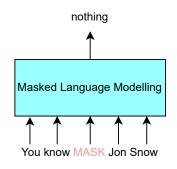


#### Transformer Architecture [21]



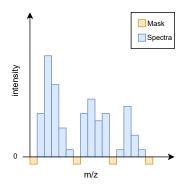


# Pre-Training: Masked Spectra Modelling [22]





# Pre-Training: Masked Spectra Modelling [22]





# Pre-Training: Next Spectra Prediction [22]

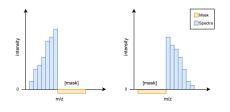
Sentence 1 Sentence 2 Next Sentence?

The quick brown fox jumped over the lazy dog.

The quick brown fox You know nothing Jon Snow.



# Pre-Training: Next Spectra Prediction [22]



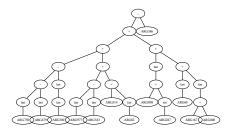


#### **Decision Tree**

```
110 122786584657 <= 19 426
                 qini = 0.496
                samples = 187
               value = [85, 102]
                 class = Hoki
                        439.163087160249 <= 300.837
  qini = 0.0
                                  qini = 0.023
samples = 101
                                 samples = 86
value = [0, 101]
                                value = [85, 1]
 class = Hoki
                               class = Mackerel
                   qini = 0.0
                                                   gini = 0.0
                 samples = 1
                                                 samples = 85
                                                value = [85, 0]
                 value = [0, 1]
                 class = Hoki
                                                class = Mackerel
```



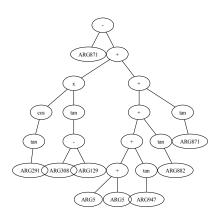
#### Genetic Programming Tree - Hoki



Genetic Programming Tree - Fish Species Hoki



## Genetic Programming Tree - Mackerel



Genetic Programming Tree - Fish Species Hoki



#### TLDR;

**Transformer** can predict fish species with near-perfect accuracy, **DT** and **GP** provide **accurate**, **interpretable** and **efficient** models for **Rapid Evaporative Ionisation Mass Spectrometry**.



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