

# AUTUMATION, AUTUMATION, AUTUMATION:

# Approaches to Improving the Pre-Excavation Detection of Inhumations

## ASHELY GREEN

Supervisors: Paul Cheetham and Professor Timothy Darvill

ipzoli

Bournemouth

University

agreen@bournemouth.ac.uk | Postgraduate Researcher

Department of Archaeology, Anthropology and Forensic Science | Bournemouth University

## INTRODUCTION

As large scale landscape surveys continue to increase in commercial and research archaeogeophysics, there is still a markedly low ability to geophysically detect and interpret archaeological and forensic inhumations in some instances. The aim of this ongoing research project is to improve data acquisition by implementing an interactive ad hoc workflow model for determining appropriate methodologies for ground-penetrating radar (GPR) surveys, to improve data processing speed, and reduce observer error.

Can the confidence of manual interpretations of GPR data be improved by adapting machine learning libraries for automatic object extraction and classification to GPR data based on a training dataset comprised of ground-truthed real GPR data and simulated GPR data?



### St Catherine's Church, Temple, Cornwall



 Church on site of 12th century Templar chapel once part of the Preceptory of Trebeigh Post-medieval to modern gravemarkers







• 13/14th century square church with 17-19th century additions

Post-medieval gravemarkers



Yellow Steeple, St Mary's Abbey, Trim, Co Roscomroe Church, Roscrea, The Leap, Meath, Ireland



. Tower associated with the Augustinian abbey dedicated to St Mary

• Built before 1140





Co Offaly, Ireland

- . Late medieval rectangular church dedicated to St Molua
- Medieval and modern sectors in graveyard



## PRELIMINARY RESULTS

C:\Continuum\Anaconda3\envs\tensorflow) C:\Users\agreen\models\tutorials\image\ inagenst)python classify_inage.py	Ту
$2017-997-97$ , 11:21:22, 01:23; w c:\Xtensortiou_150(1907200-004,00+K)tensortiou-1.2; Xtensorflowicorevplationvcourfacture_guard.cc:451 The TensorFlow library wasn = copiled to use SEE instructions, but these are available on your machine $8017-998-97$ , 11:21:25, 01:25:31; W c:\Xtensorflowurflow	Lintel
(Xtensorflow/corevplatform/cpu/feature_guard.cc:45) The TensorFlow library wasn't compiled to use \$SE2 instructions, but these are available on your machine and could speed up CPU computations. 2017-09-07 11:21:25.012539: W c:\\tensorflow_1501907206084\work\tensorflow-1.2.	Cist
Altensorflow/core/platform/cpulfeature_guard.cc:45] The TensorFlow library wasn' c compiled to use \$853 instructions, but these are available on your machine and could speed up CPU computations. \$207-399-97 11:21:25,2015:253: W civiltensorflow_15 <u>01907</u> 206084\work\tensorflow-1.2.	Unlined
Altensorf low/core/platform/cpu_feature_guard.cc:45) The fensorflow Tibrary wasn' c compiled to use \$\$\$f41 instructions, but these are available on your machine a nd could speed up CPU computations. \$817-99-97 11:21:25,012:53): Uc:Nltensorflow_1581907286084\work\tensorflow-1.2,	Slab-lined
t consist iow core year of the constructions, but these are available on your machine a t consist iow core year of the constructions, but these are available on your machine a pd (7-99-67) ii (21:25, 01:253); U constructions of law 1501907206084 work/tensorFlour1.2 Ktensorf loukconstructions to available on the construction of the construction of the second of	Pit
c compiled to use AUX instructions, but these are available on your machine and pould speed up CPU computations. 2017-09-07 11:21:26.527008: W c:\\tensorflow_15019072060084\work\tensorflow-1.2. R\tensorflow\core\framework\op_def_util.cc:3321 Op_BatchNormWithGlobalNormalizat ion is deprecated. It will cease to work in GraphDef version 9. Use tf.nn.batch_	Later Medie
րդյալlization() yiant panda, panda, panda bear, coon bear, Ailuropoda melanoleuca (score = 0.896 32)	Later Medie
indri, indris, Indri indri, Indri brevicaudatus (score = 0.00766) Jesser panda, red panda, panda, bear cat, cat bear, Ailurus fulgens (score = 0.0 2065) uustard apple (score = 0.00138) iarthstar (score = 0.00104)	Later Medie
Example of output given by Tensorflow Inception-v3 trained on ImageNet data (Google 2017)	Massurad

Туре	Sample Size	Avg Length	Avg Width	Avg Depth	Soil Types
Lintel	23	1.895m	0.42m	0.295m	Clay, Sandy Clay, Clayey Loam
Cist	5	1.69m	0.40m	0.27m	Clay, Sandy Clay, Clayey Loam, Sand
Unlined	42	1.8m	0.67m	0.27m	Clay, Sandy Clay, Clayey Loam
Slab-lined	3	1.58m	0.88m	0.4m	Clay, Sandy Clay, Clayey Loam, Sand
Pit	6	1.94m	0.87m	0.33m	Clay, Sandy Clay, Clayey Loam
Later Medieval Simple	1	-	-	0.3m	Clay, Sandy Clay, Clayey Loam
Later Medieval Pit	1	0.7m	0.7m	0.3m	Clay, Sandy Clay, Clayey Loam
Later Medieval Coffin	2	1.775m	0.525m	0.45m	Clay, Sandy Clay, Clayey Loam

Measured characteristics used to simulate GPR data and train the networks (after Cahill and Sikora 2011)



Basalt (0.06%) Dolerite (0.20%) Gabbro (0.19%) Granite (4.54%) Greywacke (5.04%) Limestone (38.13%) Mudstone (6.49%) Quartz (2.02%) Rhvolite (0.04%) Sandstone (18.82%) Schists (3.71%) Shale (5.01%) Siltstone (1.93%) Slate (3.25%) Volcanic (1.71%) Other (8.88%) Bedrock geology coverage of Ireland (in square kilometres)

**OPW** 

### Have data vou would like to contribute to the training dataset? Let me know!



BIBLIOGRAPHY

- Alemi, A., 2016. Improving Inception and Image Classification in TensorFlow. Google Research Blog, https:// research.co.odeblog.com/2016/08/improving-inception-and-image.html
- Freienisch usd indering mit 2016/08/improving-inception-and-image html research google tocsdard todge learning Earth Sci Inform, 9, 553-545.
  Bonsail, J. Gaffney, C., Amit, 12014. Preparing for the Future: A reappraisal of archaeo-geophysical surveying on Irish National Road Schemes 2007-2010.
  Cahill, M. and Sikora, M., 2011. Breaking Ground, Finding Graves reports on the excavations of burials by the National Road Schemes 2007-2000.
  Long Tar. 2007. Schemes 2007-2000.
  Long Andread Schemes

- David, A, Lintord, N, and Lintord, P 2008. Geophysical Survey in Archaeological Field Evaluation. English Heritage Swindon, UK, Bruzzone, L, 2011. Batch-mode active learning methods for the interactive classification of remote sensing images, IEEE Trans Geoscience Remote Sensing, 49(3), 1014-1031.
  Eberle, D, Hutchins, D, Das, S, Majumdar, A, and Pasache, H 2015. Automated pattern recognition to support geological mepping and exploration target generation A case study from southern Namibia, Journal of African Earth Sciences, 106, 60-74.
- Huang, J., Rathod, V., Sun, C., Zhu, M., Korattikara, A., Fathi, A., Fischer, I., Wojna, Z., Song, Y., Guadarrama, S., Mumhu K. 2017. Sneed/Accuracy trade-offs for modern convolutional object detectors. Cornell University Library
- Hulp Iyn, X. 2017. Specifyed Actinety inder-ons on modern convolution updex declars. Come overeasy Lineary, https://anx.org/abs/1511.1002. afraséh, B and Fahimpour, N2017. Automatic extraction of geometrical characteristics hidden in ground-penetrating roots specifical images using simultaneous perturbation artificial bee codiny algorithm. Geophysical Prospecting, 65, 1999. radar s
- 224-336. Mediate can ge instancian be can also be an advanced to the second second responsible constraints and constraints

#### ACKNOWLEDGEMENTS





comhairle chontae na mí

