

¹ An automated pipeline for LOFAR very-long baseline interferometry

³ **Matthijs van der Wild**  ¹¶, **Frits Sweijen**  ¹, **Jurjen M. G. H. J. de Jong**  ^{2,3}, **Alexander Drabent**  ⁴, **Emmy L. Escott**  ¹, **Neal Jackson**  ⁵, **Marcel Loose**  ³, **Vijay Mahatma**  ^{6,7}, **Leah K. Morabito**  ^{1,8}, and **James Petley** 

⁷ **1** Centre for Extragalactic Astronomy, Department of Physics, Durham University, South Road, Durham

⁸ DH1 3LE, UK **2** Leiden Observatory, Leiden University, PO Box 9513, 2300 RA Leiden, The Netherlands

⁹ **3** ASTRON, Oude Hoogeveensedijk 4, 7991 PD Dwingeloo, The Netherlands **4** Thüringer

¹⁰ Landessternwarte, Sternwarte 5, 07778 Tautenburg, Germany **5** University of Manchester, Jodrell Bank

¹¹ Centre for Astrophysics, Department of Physics and Astronomy, Oxford Rd, Manchester M13 9PL,

¹² United Kingdom **6** Cavendish Laboratory - Astrophysics Group, University of Cambridge, 19 JJ Thomson

¹³ Avenue, Cambridge CB3 0HE, United Kingdom **7** Kavli Institute for Cosmology, University of Cambridge,

¹⁴ Madingley Road, Cambridge CB3 0HA, United Kingdom **8** Institute for Computational Cosmology,

¹⁵ Department of Physics, Durham University, South Road, Durham DH1 3LE, United Kingdom ¶

¹⁶ Corresponding author

DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

Software

- [Review](#) ↗
- [Repository](#) ↗
- [Archive](#) ↗

Editor: ↗

Submitted: 02 December 2025

Published: unpublished

License

Authors of papers retain copyright ²⁵ and release the work under a ²⁶ Creative Commons Attribution 4.0 International License ([CC BY 4.0](#))³⁸

¹⁷ Summary

¹⁸ The Very-Long Baseline Interferometry (VLBI) Pipeline for the International Low-Frequency ¹⁹ ARray Telescope (PILoT) is an automated data reduction pipeline that produces calibrated ²⁰ radio data suitable for sub-second resolution imaging. It is a tool that facilitates calibrator and ²¹ source selection, self-calibration of data, and both postage-stamp and widefield imaging.

²² While a diverse ecosystem of processing and imaging tools exists for the LOFAR telescope, ²³ none of those tools have been designed with high-resolution imaging in mind. As a result, ²⁴ data reduction with the International LOFAR Telescope is a manual and error-prone process. ²⁵ Furthermore, owing to the distributed nature of software development in the LOFAR community, ²⁶ all of these tools have been developed with different input and output conventions.

²⁷ PILoT aims to incorporate these diverse software tools into a unified framework, making VLBI ²⁸ imaging with LOFAR accessible to a larger group of astronomers. Special care has been placed ²⁹ on ensuring that PILoT is interoperable and reusable, and that all of its software components ³⁰ are controlled through a consistent framework and that intermediate steps of the pipeline ³¹ can be consistently and safely resumed in the event of intermediate failure. Since typical ³² observations done with the International LOFAR Telescope result in datasets consisting of ³³ terabytes of data, the pipeline has been designed to integrate with job schedulers common ³⁴ in High-Performance Computing (HPC) clusters. This minimises manual intervention and ³⁵ optimises the use of available computing resources.

³⁶ Statement of need

³⁷ The International Low-Frequency ARray Telescope (ILT) ([van Haarlem et al., 2013](#)) currently ³⁸ comprises 38 Dutch stations and 14 international stations located in partner countries across ³⁹ Europe. It is a radio telescope operating at radio frequencies under 240 MHz with a sensitivity ⁴⁰ of up to 3 orders of magnitude better than previous telescopes operating at comparable ⁴¹ frequencies. By combining data from all stations with baselines of up to 1980 km, the ILT is

42 effectively a continent-sized telescope which is able to image astronomical radio sources at
43 sub-arcsecond resolution ([Leah K. Morabito et al., 2025; Varenius et al., 2015](#)).

44 The VLBI Pipeline for the International LOFAR Telescope (PILoT) is an implementation of
45 a data reduction pipeline which was designed to exploit the full imaging power of the ILT
46 ([L. K. Morabito et al., 2022](#)). PILoT addresses several critical issues the original reference
47 implementation had to face:

- 48 ▪ The original pipeline was implemented in an obsolete framework, which makes it difficult
49 to impossible to ensure that it would be functional on modern computing infrastructure.
50 In contrast, PILoT is implemented in the Common Workflow Language (CWL; Crusoe et
51 al. (2022)), which is an actively maintained framework in widespread use. This ensures
52 that the pipeline will be logically consistent and maintainable in the long term.
- 53 ▪ The original pipeline did not support modern scheduling systems such as Slurm or
54 TORQUE. The implementation in CWL allowed for optimisation of PILoT for the
55 workflow runner toil ([Vivian et al., 2017](#)), providing native support for these schedulers.
56 This reduces the runtimes of the pipeline by orders of magnitudes as individual processing
57 jobs can automatically be distributed to available nodes.

58 In addition, PILoT includes expanded functionality featuring implementations of state-of-the-art
59 advances in imaging techniques such as improvements in imaging resolution ([F. Sweijen et al.,
60 2022; Ye et al., 2024](#)), source selection ([de Jong et al., 2024; F. Sweijen et al., 2022](#)), and
61 wide-field imaging techniques ([de Jong et al., 2025; F. Sweijen et al., 2022](#)).

62 PILoT forms a natural part of the LOFAR software landscape and is designed to be used on
63 data that has been corrected for various instrumental and ionospheric effects ([de Gasperin et
64 al., 2019](#)) and calibrated for directional effects using the data obtained by the Dutch stations
65 using pipelines such as DDF-pipeline ([Hardcastle et al., 2019; Tasse et al., 2021](#)) or Rapthor
66 ([Rafferty et al., 2021](#)). It uses DP3 ([Dijkema et al., 2023](#)), WSclean ([Offringa et al., 2014](#)),
67 AOflagger ([Offringa, 2010](#)), and the LOFAR Initial Calibration (LINC) pipeline ([Drabent et al.,
68 2022](#)), which are developed by the ILT host institute ASTRON, as well as various codebases
69 developed by researchers in the LOFAR community such as the DDF-pipeline ([Tasse et al.,
70 2021](#)) for direction-dependent calibration, LOFAR facet-selfcal ([R. van Weeren, 2022; R. J. van
71 Weeren et al., 2021](#)) for self-calibration, and the LOFAR Helpers ([de Jong, 2021; de Jong et
72 al., 2022](#)) auxiliary library. Finally, it has been adapted and integrated into the FLoCs LOFAR
73 containers ([Frits Sweijen et al., 2025](#)) to ensure portability across computing facilities.

74 PILoT embeds the software tools above into a single cohesive framework and provides several
75 complementary modes with the aim of being a modular and fully automated imaging pipeline.
76 It is able to determine when intermediate results are no longer needed, and disposes of
77 intermediate results once they are no longer required. Furthermore, since the pipeline is
78 optimised to work with toil, processing steps can easily be resumed should the execution of a
79 mode be interrupted for any reason.

80 The pipeline provides the following main modes of operation:

81 **Postage stamp imaging**

82 The pipeline supports single-source imaging, in which the calibration is performed on an in-field
83 calibrator to correct for direction-independent phases and delays from the international stations.
84 PILoT selects the best calibrator based on the images and solutions from all performed self-
85 calibration cycles. Following this, multiple imaging targets can be specified at once; the pipeline
86 performs self-calibration and imaging on each target (with a resolution of 0.3 arcseconds) in
87 parallel. The data products are the calibrated data and images and the phase solutions in a
88 H5parm format¹.

¹The H5parm format is described in appendix C of ([de Gasperin et al., 2019](#)).

89 **Wide-field imaging**

90 The pipeline is capable of intermediate (1–2 arcseconds) and high-resolution (sub-arcsecond)
91 imaging. Before imaging the pipeline removes radio sources from observation data beyond
92 a square field of view of 6.25 degrees² centred on the imaging target, if given a 6 arcsecond
93 image generated by the DDF-pipeline. High-resolution imaging supports a resolution of 0.6 or
94 0.3 arcseconds, of which the former reduces the imaging time by a factor of 4 compared to the
95 latter. Intermediate resolution imaging speeds up the imaging time by a factor of 16 compared
96 to the 0.3 arcsecond imaging. The data products are FITS formatted images of the stated
97 resolution.

98 A list of ongoing research projects where PILoT is a central tool is provided in section 4 of
99 reference ([Leah K. Morabito et al., 2025](#)).

100 **Acknowledgements**

101 MvdW is supported by the Science and Technology Facilities Council via LOFAR-U.K.
102 [ST/V002406/1] and UKSRC [ST/T000244/1]. FS appreciates the support of STFC
103 [ST/Y004159/1]. JMGHJdJ acknowledges support from project CORTEX (NWA.1160.18.316)
104 of research programme NWA-ORC, which is (partly) financed by the Dutch Research Council
105 (NWO), and support from the OSCARS project, which has received funding from the
106 European Commission's Horizon Europe Research and Innovation programme under grant
107 agreement No. 101129751. LKM is grateful for support from a UKRI FLF [MR/Y020405/1]
108 and LOFAR-UK via STFC [ST/V002406/1]. AD acknowledges support by the BMBF
109 Verbundforschung under the grant 05A23STA. ELE is grateful for support from the Medical
110 Research Council [MR/T042842/1]. This research made use of the University of Hertfordshire
111 high-performance computing facility and the LOFAR-UK computing facility located at the
112 University of Hertfordshire and supported by STFC [ST/P000096/1].

113 The scripts developed for PILoT make use of the ASTROPY ([Astropy Collaboration et al., 2022](#)),
114 casacore ([Casacore Team, 2019](#)), LoSoTo ([de Gasperin et al., 2024](#)), numpy ([Harris et al., 2020](#)), pandas ([McKinney, 2010](#)), and PyBDSF ([Mohan & Rafferty, 2015](#)) libraries.

116 **References**

- 117 Astropy Collaboration, Price-Whelan, A. M., Lim, P. L., Earl, N., Starkman, N., Bradley, L.,
118 Shupe, D. L., Patil, A. A., Corrales, L., Brasseur, C. E., Nöthe, M., Donath, A., Tollerud,
119 E., Morris, B. M., Ginsburg, A., Vaher, E., Weaver, B. A., Tocknell, J., Jamieson, W., ...
120 Astropy Project Contributors. (2022). The Astropy Project: Sustaining and Growing a
121 Community-oriented Open-source Project and the Latest Major Release (v5.0) of the Core
122 Package. *935(2)*, 167. <https://doi.org/10.3847/1538-4357/ac7c74>
- 123 Casacore Team. (2019). *casacore: Suite of C++ libraries for radio astronomy data processing*.
124 Astrophysics Source Code Library, record ascl:1912.002.
- 125 Crusoe, M. R., Abeln, S., Iosup, A., Amstutz, P., Chilton, J., Tijanić, N., Ménager, H.,
126 Soiland-Reyes, S., Gavrilović, B., Goble, C., & Community, T. C. (2022). Methods
127 included: Standardizing computational reuse and portability with the common workflow
128 language. In *Communications of the ACM* (Vol. 65, pp. 54–63). <https://doi.org/https://doi.org/10.1145/3486897>
- 130 de Gasperin, F., Dijkema, T. J., Drabent, A., Mevius, M., Rafferty, D., van Weeren, R.,
131 Brüggen, M., Callingham, J. R., Emig, K. L., Heald, G., Intema, H. T., Morabito, L. K.,
132 Offringa, A. R., Oonk, R., Orrù, E., Röttgering, H., Sabater, J., Shimwell, T., Shulevski,
133 A., & Williams, W. (2024). *LoSoTo: LOFAR solutions tool*. Astrophysics Source Code
134 Library, record ascl:2401.006.

- 135 de Gasperin, F., Dijkema, T. J., Drabent, A., Mevius, M., Rafferty, D., Weeren, R. van,
 136 Brüggen, M., Callingham, J. R., Emig, K. L., Heald, G., Intema, H. T., Morabito, L. K.,
 137 Offringa, A. R., Oonk, R., Orrù, E., Röttgering, H., Sabater, J., Shimwell, T., Shulevski, A.,
 138 & Williams, W. (2019). Systematic effects in LOFAR data: A unified calibration strategy.
 139 *622*, A5. <https://doi.org/10.1051/0004-6361/201833867>
- 140 de Jong, J. M. G. H. J. (2021). LOFAR helpers. In *GitHub repository*. https://github.com/jurjen93/lofar_helpers; GitHub.
- 142 de Jong, J. M. G. H. J., van Weeren, R. J., Botteon, A., Oonk, J. B. R., Brunetti, G.,
 143 Shimwell, T. W., Cassano, R., Röttgering, H. J. A., & Tasse, C. (2022). Deep study of
 144 A399-401: Application of a wide-field facet calibration. *668*, A107. <https://doi.org/10.1051/0004-6361/202244346>
- 146 de Jong, J. M. G. H. J., Veefkind, L., Weeren, R. J. van, Oonk, J. B. R., Schlimbach, R. J.,
 147 Kampert, D. N. G., Wild, M. van der, Morabito, L. K., Sweijen, F., Offringa, A. R., &
 148 Röttgering, H. J. A. (2025). Scalable and robust wide-field facet calibration with LOFAR's
 149 longest baselines. <https://doi.org/10.1093/mnras/staf1373>
- 150 de Jong, J. M. G. H. J., Weeren, R. J. van, Sweijen, F., Oonk, J. B. R., Shimwell, T. W.,
 151 Offringa, A. R., Morabito, L. K., Röttgering, H. J. A., Kondapally, R., Escott, E. L.,
 152 Best, P. N., Bondi, M., Ye, H., & Petley, J. W. (2024). Into the depths: Unveiling
 153 ELAIS-N1 with LOFAR's deepest sub-arcsecond wide-field images. *A&A*, *689*, A80.
 154 <https://doi.org/10.1051/0004-6361/202450595>
- 155 Dijkema, T. J., Nijhuis, M., Diepen, G. van, Offringa, A., Krombein, L., Wever, M. de, Maljaars,
 156 J., & Loose, M. (2023). *DP3: Streaming processing pipeline for radio interferometric data*.
 157 Astrophysics Source Code Library, record ascl:2305.014.
- 158 Drabent, A., Rafferty, D., Mancini, M., Loose, M., Horneffer, A., de Gasperin, F., Iacobelli, M.,
 159 Orrù, E., Adebahr, B., Hardcastle, M., Heald, G., Mandal, S., Roskowinski, C., Sabater
 160 Montes, J., Shimwell, T., Sridhar, S., Weeren, R. van, & Williams, W. (2022). LOFAR
 161 initial calibration (LINC) pipeline. In *Gitlab repository*. <https://git.astron.nl/RD/LINC>;
 162 ASTRON.
- 163 Hardcastle, M., Shimwell, T., Tasse, T., & Williams, W. (2019). DDF pipeline. In *GitHub
 164 repository*. <https://github.com/mhardcastle/ddf-pipeline>; GitHub.
- 165 Harris, C. R., Millman, K. J., Walt, S. J. van der, Gommers, R., Virtanen, P., Cournapeau, D.,
 166 Wieser, E., Taylor, J., Berg, S., Smith, N. J., Kern, R., Picus, M., Hoyer, S., Kerkwijk,
 167 M. H. van, Brett, M., Haldane, A., Río, J. F. del, Wiebe, M., Peterson, P., ... Oliphant,
 168 T. E. (2020). Array programming with NumPy. *Nature*, *585*(7825), 357–362. <https://doi.org/10.1038/s41586-020-2649-2>
- 170 McKinney, Wes. (2010). Data Structures for Statistical Computing in Python. In Stéfan van
 171 der Walt & Jarrod Millman (Eds.), *Proceedings of the 9th Python in Science Conference*
 172 (pp. 56–61). <https://doi.org/10.25080/Majora-92bf1922-00a>
- 173 Mohan, N., & Rafferty, D. (2015). *PyBDSF: Python Blob Detection and Source Finder*.
 174 Astrophysics Source Code Library, record ascl:1502.007.
- 175 Morabito, L. K., Jackson, N. J., Mooney, S., Sweijen, F., Badole, S., Kukreti, P., Venkattu,
 176 D., Groeneveld, C., Kappes, A., Bonnassieux, E., Drabent, A., Iacobelli, M., Croston, J.
 177 H., Best, P. N., Bondi, M., Callingham, J. R., Conway, J. E., Deller, A. T., Hardcastle, M.
 178 J., ... Zucca, P. (2022). Sub-arcsecond imaging with the international LOFAR telescope - i.
 179 Foundational calibration strategy and pipeline. *A&A*, *658*, A1. <https://doi.org/10.1051/0004-6361/202140649>
- 181 Morabito, Leah K., Jackson, N., Jong, J. de, Escott, E., Groeneveld, C., Mahatma, V.,
 182 Petley, J., Sweijen, F., Timmerman, R., & Weeren, R. J. van. (2025). A decade of
 183 sub-arcsecond imaging with the International LOFAR Telescope. *370*(2), 19. <https://doi.org/10.1051/0004-6361/202450595>

- 184 [//doi.org/10.1007/s10509-025-04406-x](https://doi.org/10.1007/s10509-025-04406-x)
- 185 Offringa, A. R. (2010). *AOfagger: RFI Software*. Astrophysics Source Code Library, record
186 ascl:1010.017.
- 187 Offringa, A. R., McKinley, B., Hurley-Walker, N., Briggs, F. H., Wayth, R. B., Kaplan, D. L.,
188 Bell, M. E., Feng, L., Neben, A. R., Hughes, J. D., Rhee, J., Murphy, T., Bhat, N. D. R.,
189 Bernardi, G., Bowman, J. D., Cappallo, R. J., Corey, B. E., Deshpande, A. A., Emrich, D.,
190 ... Williams, C. L. (2014). WSClean: an implementation of a fast, generic wide-field imager
191 for radio astronomy. *MNRAS*, 444(1), 606–619. <https://doi.org/10.1093/mnras/stu1368>
- 192 Rafferty, D., Loose, M., Offringa, A., Dijkstra, A. G., Dijkema, T. J., Wever, M. de, Sweijen,
193 F., & Yatawatta, S. (2021). Raphor: LOFAR DDE pipeline. In *Gitlab repository*.
194 <https://git.astron.nl/RD/Raphor>; ASTRON.
- 195 Sweijen, Frits, Kurek, A., Weeren, R. van, & Wild, M. van der. (2025). *tikk3r/flocs: v6.0.0*
196 (Version v6.0.0). Zenodo. <https://doi.org/10.5281/zenodo.17038431>
- 197 Sweijen, F., Weeren, R. J. van, Röttgering, H. J. A., Morabito, L. K., Jackson, N., Offringa,
198 A. R., Tol, S. van der, Veenboer, B., Oonk, J. B. R., Best, P. N., Bondi, M., Shimwell,
199 T. W., Tasse, C., & Thomson, A. P. (2022). Deep sub-arcsecond wide-field imaging of
200 the Lockman Hole field at 144 MHz. *Nature Astronomy*, 6, 350–356. <https://doi.org/10.1038/s41550-021-01573-z>
- 201 Tasse, C., Shimwell, T., Hardcastle, M. J., O'Sullivan, S. P., Weeren, R. van, Best, P.
202 N., Bester, L., Hugo, B., Smirnov, O., Sabater, J., Calistro-Rivera, G., de Gasperin,
203 F., Morabito, L. K., Röttgering, H., Williams, W. L., Bonato, M., Bondi, M., Botteon,
204 A., Brüggen, M., ... Wiaux, Y. (2021). The LOFAR Two-meter Sky Survey: Deep
205 Fields Data Release 1. I. Direction-dependent calibration and imaging. 648, A1. <https://doi.org/10.1051/0004-6361/202038804>
- 206 van Haarlem, M. P., Wise, M. W., Gunst, A. W., Heald, G., McKean, J. P., Hessels, J. W.
207 T., Bruyn, A. G. de, Nijboer, R., Swinbank, J., Fallows, R., Brentjens, M., Nelles, A.,
208 Beck, R., Falcke, H., Fender, R., Hörandel, J., Koopmans, L. V. E., Mann, G., Miley, G.,
209 ... Zensus, A. (2013). LOFAR: The LOw-Frequency ARray. *Astron. Astrophys.*, 556, A2.
210 <https://doi.org/10.1051/0004-6361/201220873>
- 211 van Weeren, R. (2022). LOFAR facet-selfcal. In *GitHub repository*. https://github.com/rvweeren/lofar_facet_selfcal; GitHub.
- 212 van Weeren, R. J., Shimwell, T. W., Botteon, A., Brunetti, G., Brüggen, M., Boxelaar, J.
213 M., Cassano, R., Di Gennaro, G., Andrade-Santos, F., Bonnassieux, E., Bonafede, A.,
214 Cuciti, V., Dallacasa, D., de Gasperin, F., Gastaldello, F., Hardcastle, M. J., Hoeft, M.,
215 Kraft, R. P., Mandal, S., ... Wilber, A. G. (2021). LOFAR observations of galaxy clusters
216 in HETDEX. Extraction and self-calibration of individual LOFAR targets. 651, A115.
217 <https://doi.org/10.1051/0004-6361/202039826>
- 218 Varenius, E., Conway, J. E., Martí-Vidal, I., Beswick, R., Deller, A. T., Wucknitz, O., Jackson,
219 N., Adebahr, B., Pérez-Torres, M. A., Chyží, K. T., Carozzi, T. D., Moldón, J., Aalto, S.,
220 Beck, R., Best, P., Dettmar, R.-J., van Driel, W., Brunetti, G., Brüggen, M., ... White, G.
221 J. (2015). Subarcsecond international LOFAR radio images of the M82 nucleus at 118
222 MHz and 154 MHz. 574, A114. <https://doi.org/10.1051/0004-6361/201425089>
- 223 Vivian, J., Rao, A. A., Nothaft, F. A., Ketchum, C., Armstrong, J., Novak, A., Pfeil, J.,
224 Narkizian, J., Deran, A. D., Musselman-Brown, A., Schmidt, H., Amstutz, P., Craft, B.,
225 Goldman, M., Rosenbloom, K., Cline, M., O'Connor, B., Hanna, M., Birger, C., ... Paten,
226 B. (2017). Toil enables reproducible, open source, big biomedical data analyses. *Nature
227 Biotechnology*, 314–316. [https://doi.org/https://doi.org/10.1038/nbt.3772](https://doi.org/10.1038/nbt.3772)
- 228 Ye, H., Sweijen, F., Weeren, R. J. van, Williams, W., Jong, J. de, Morabito, L. K., Rottgering,
229 H., Shimwell, T. W., Best, P. N., Bondi, M., Brüggen, M., de Gasperin, F., & Tasse, C.
230

²³³ (2024). 1-arcsecond imaging of the ELAIS-N1 field at 144MHz using the LoTSS survey
²³⁴ with the international LOFAR telescope. *691*, A347. <https://doi.org/10.1051/0004-6361/202348103>
²³⁵

DRAFT