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RESEARCH ARTICLE

Transcriptome‐wide association study of breast cancer risk by estrogen‐receptor status

Helian Feng^{1,2,3} | Alexander Gusev⁴ | Bogdan Pasaniuc⁵ | Lang Wu⁶ | Jirong Long⁷ | Zomoroda Abu-full⁸ | Kristiina Aittomäki⁹ | Irene L. Andrulis^{10,11} | Hoda Anton-Culver¹² | Antonis C. Antoniou¹³ | Adalgeir Arason^{14,15} | Volker Arndt¹⁶ | Kristan J. Aronson¹⁷ | Banu K. Arun¹⁸ | Ella Asseryanis¹⁹ | Paul L. Auer^{20,21} | Jacopo Azzollini²² | Judith Balmaña²³ | Rosa B. Barkardottir^{14,15} | Daniel R. Barnes¹³ | Daniel Barrowdale¹³ | Matthias W. Beckmann²⁴ | Sabine Behrens²⁵ | Javier Benitez^{26,27} | Marina Bermisheva²⁸ | Katarzyna Białkowska²⁹ | Ana Blanco^{26,30,31} Carl Blomqvist^{32,33} | Bram Boeckx^{34,35} | Natalia V. Bogdanova</sub>^{36,37,38} | Stig E. Bojesen^{39,40,41} | Manjeet K. Bolla¹³ | Bernardo Bonanni⁴² | Ake Borg⁴³ | Hiltrud Brauch^{44,45,46} | Hermann Brenner^{16,46,47} | Ignacio Briceno^{48,49} | Annegien Broeks⁵⁰ | Thomas Brüning⁵¹ | Barbara Burwinkel^{52,53} | Qiuyin Cai⁷ | Trinidad Caldés⁵⁴ | Maria A. Caligo⁵⁵ | Ian Campbell^{56,57} | Sander Canisius^{50,58} | Daniele Campa⁵⁹ | Brian D. Carter⁶⁰ | Jonathan Carter⁶¹ | Jose E. Castelao⁶² | Jenny Chang-Claude^{25,63} | Stephen J. Chanock⁶⁴ | Hans Christiansen³⁶ | Wendy K. Chung⁶⁵ | Kathleen B. M. Claes⁶⁶ | Christine L. Clarke⁶⁷ | GEMO Study Collaborators^{68,69,70} | EMBRACE Collaborators¹³ | GC-HBOC study Collaborators⁷¹ | Fergus J. Couch⁷² | Angela Cox⁷³ | Simon S. $Cross^{74}$ | Cezary Cybulski²⁹ | Kamila Czene⁷⁵ | Mary B. Daly⁷⁶ | Miguel de la Hoya⁵⁴ | Kim De Leeneer⁶⁶ | Joe Dennis¹³ | Peter Devilee^{77,78} | Orland Diez^{79,80} | Susan M. Domchek⁸¹ | Thilo Dörk³⁷ | Isabel dos-Santos-Silva⁸² | Alison M. Dunning⁸³ | Miriam Dwek⁸⁴ | Diana M. Eccles⁸⁵ | Bent Ejlertsen⁸⁶ | Carolina Ellberg⁸⁷ | Christoph Engel^{88,89} | Mikael Eriksson⁷⁵ | Peter A. Fasching^{24,90} | Olivia Fletcher⁹¹ | Henrik Flyger⁹² | Florentia Fostira⁹³ | Eitan Friedman^{94,95} | Lin Fritschi⁹⁶ | Debra Frost¹³ | Marike Gabrielson⁷⁵ | Patricia A. Ganz⁹⁷ | Susan M. Gapstur⁶⁰ | Judy Garber⁹⁸ | Montserrat García-Closas^{64,99,100} | José A. García-Sáenz⁵⁴ | Mia M. Gaudet⁶⁰ | Graham G. Giles^{101,102} | Gord Glendon¹⁰ | Andrew K. Godwin¹⁰³ | Mark S. Goldberg^{104,105} | David E. Goldgar¹⁰⁶ | Anna González-Neira²⁷ |

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Mark H. Greene¹⁰⁷ | Jacek Gronwald²⁹ | Pascal Guénel¹⁰⁸ | Christopher A. Haiman¹⁰⁹ | Per Hall^{75,110} | Ute Hamann¹¹¹ | Christopher Hake¹¹² | Wei He⁷⁵ | Jane Heyworth¹¹³ | Frans B.L. Hogervorst¹¹⁴ | Antoinette Hollestelle¹¹⁵ | Maartie J. Hooning¹¹⁵ | Robert N. Hoover⁶⁴ | John L. Hopper¹⁰¹ | Guanmengqian Huang¹¹¹ | Peter J. Hulick^{116,117} | Keith Humphreys⁷⁵ | Evgeny N. Imyanitov¹¹⁸ | ABCTB Investigators¹¹⁹ | HEBON Investigators¹²⁰ | BCFR Investigators¹²¹ | OCGN Investigators¹²² | Claudine Isaacs¹²³ | Milena Jakimovska¹²⁴ | Anna Jakubowska^{29,125} | Paul James^{57,126} | Ramunas Janavicius¹²⁷ | Rachel C. Jankowitz¹²⁸ | Esther M. John¹²¹ | Nichola Johnson⁹¹ | Vijai Joseph¹²⁹ | Audrey Jung²⁵ | Beth Y. Karlan¹³⁰ | Elza Khusnutdinova^{28,131} | Johanna I. Kiiski¹³² | Irene Konstantopoulou⁹³ | Vessela N. Kristensen^{133,134} | Yael Laitman⁹⁴ | Diether Lambrechts^{34,35} | Conxi Lazaro¹³⁵ | Dominique Leroux¹³⁶ | Goska Leslie¹³ | Jenny Lester¹³⁰ | Fabienne Lesueur^{69,70,137} | Noralane Lindor¹³⁸ | Sara Lindström^{139,140} | Wing-Yee Lo^{44,45} | Jennifer T. Loud¹⁰⁷ | Jan Lubiński²⁹ | Enes Makalic¹⁰¹ | Arto Mannermaa^{141,142,143} | Mehdi Manoochehri¹¹¹ | Siranoush Manoukian²² | Sara Margolin^{110,144} | John W.M. Martens¹¹⁵ | Maria E. Martinez^{145,146} | Laura Matricardi¹⁴⁷ | Tabea Maurer⁶³ | Dimitrios Mavroudis¹⁴⁸ | Lesley McGuffog¹³ | Alfons Meindl¹⁴⁹ | Usha Menon¹⁵⁰ | Kyriaki Michailidou^{13,151} | Pooja M. Kapoor^{25,152} | Austin Miller¹⁵³ | Marco Montagna¹⁴⁷ | Fernando Moreno⁵⁴ | Lidia Moserle¹⁴⁷ | Anna M. Mulligan^{154,155} | Taru A. Muranen¹³² | Katherine L. Nathanson⁸¹ | Susan L. Neuhausen¹⁵⁶ | Heli Nevanlinna¹³² | Ines Nevelsteen¹⁵⁷ | Finn C. Nielsen¹⁵⁸ | Liene Nikitina-Zake¹⁵⁹ | Kenneth Offit^{129,160} | Edith Olah¹⁶¹ | Olufunmilayo I. Olopade¹⁶² | Håkan Olsson⁸⁷ | Ana Osorio^{26,27} | Janos Papp¹⁶¹ | Tioung-Won Park-Simon³⁷ | Michael T. Parsons¹⁶³ | Inge S. Pedersen¹⁶⁴ | Ana Peixoto¹⁶⁵ | Paolo Peterlongo¹⁶⁶ | Julian Peto⁸² | Paul D.P. Pharoah^{13,83} | Kelly-Anne Phillips^{56,57,101,167} | Dijana Plaseska-Karanfilska¹²⁴ | Bruce Poppe⁶⁶ | Nisha Pradhan¹²⁹ | Karolina Prajzendanc²⁹ | Nadege Presneau⁸⁴ | Kevin Punie¹⁵⁷ | Katri Pylkäs^{168,169} | Paolo Radice¹⁷⁰ | Johanna Rantala¹⁷¹ | Muhammad Usman Rashid^{111,172} | Gad Rennert⁸ | Harvey A. Risch¹⁷³ | Mark Robson¹⁶⁰ | Atocha Romero¹⁷⁴ | Emmanouil Saloustros¹⁷⁵ | Dale P. Sandler¹⁷⁶ | Catarina Santos¹⁶⁵ | Elinor J. Sawyer¹⁷⁷ | Marjanka K. Schmidt^{50,178} | Daniel F. Schmidt^{101,179} | Rita K. Schmutzler^{71,180} | Minouk J. Schoemaker⁹⁹ | Rodney J. Scott^{181,182,183} | Priyanka Sharma¹⁸⁴ | Xiao-Ou Shu⁷ | Jacques Simard¹⁸⁵ | Christian F. Singer¹⁹ | Anne-Bine Skytte¹⁸⁶ | Penny Soucy¹⁸⁵ | Melissa C. Southey^{187,188} | John J. Spinelli^{189,190} | Amanda B. Spurdle¹⁶³ | Jennifer Stone^{101,191} | Anthony J. Swerdlow^{99,192} |

FENG ET AL. $\begin{array}{c|c} \hline \text{FENG ET AL.} \end{array}$

William J. Tapper¹⁹³ | Jack A. Taylor^{176,194} | Manuel R. Teixeira^{165,195} Mary Beth Terry¹⁹⁶ | Alex Teulé¹⁹⁷ | Mads Thomassen¹⁹⁸ | Kathrin Thöne⁶³ | Darcy L. Thull¹⁹⁹ | Marc Tischkowitz^{200,201} | Amanda E. Toland²⁰² | Rob A. E. M. Tollenaar²⁰³ | Diana Torres^{48,111} | Thérèse Truong¹⁰⁸ Nadine Tung²⁰⁴ | Celine M. Vachon²⁰⁵ | Christi J. van Asperen²⁰⁶ Ans M. W. van den Ouweland²⁰⁷ | Elizabeth J. van Rensburg²⁰⁸ | Ana Vega^{26,30,31} Alessandra Viel²⁰⁹ | Paula Vieiro-Balo²¹⁰ | Qin Wang¹³ | Barbara Wappenschmidt^{71,180} | Clarice R. Weinberg²¹¹ | Jeffrey N. Weitzel²¹² Camilla Wendt¹⁴⁴ | Robert Winqvist^{168,169} | Xiaohong R. Yang⁶⁴ | Drakoulis Yannoukakos⁹³ | Argyrios Ziogas¹² | Roger L. Milne^{100,101,187} | Douglas F. Easton^{13,83} | Georgia Chenevix-Trench¹⁶³ | Wei Zheng⁷ | Peter Kraft^{1,2} | Xia Jiang $1,2$

- ²Department of Epidemiology, Harvard T. H. Chan School of Public Health, Boston, Massachusetts
- ³Department of Biostatistics, Harvard T. H. Chan School of Public Health, Boston, Massachusetts
- 4 Dana‐Farber Cancer Institute, Boston, Massachusetts
- 5 UCLA Path & Lab Med, Los Angeles, California
- 6 Epidemiology Program, University of Hawaii Cancer Center, Honolulu, Hawaii
- 7 Division of Epidemiology, Department of Medicine, Vanderbilt Epidemiology Center, Vanderbilt‐Ingram Cancer Center, Vanderbilt University School of Medicine, Nashville, Tennessee
- 8 Clalit National Cancer Control Center, Carmel Medical Center and Technion Faculty of Medicine, Haifa, Israel
- 9 Department of Clinical Genetics, Helsinki University Hospital, University of Helsinki, Helsinki, Finland
- ¹⁰Fred A, Litwin Center for Cancer Genetics, Lunenfeld-Tanenbaum Research Institute of Mount Sinai Hospital, Toronto, Ontario, Canada
- ¹¹Department of Molecular Genetics, University of Toronto, Toronto, Ontario, Canada
- ¹²Department of Epidemiology, Genetic Epidemiology Research Institute, University of California Irvine, Irvine, California
- ¹³Centre for Cancer Genetic Epidemiology, Department of Public Health and Primary Care, University of Cambridge, Cambridge, UK
- ¹⁴Department of Pathology, Landspitali University Hospital, Reykjavik, Iceland
- ¹⁵BMC (Biomedical Centre), Faculty of Medicine, University of Iceland, Reykjavik, Iceland
- ¹⁶Division of Clinical Epidemiology and Aging Research, German Cancer Research Center (DKFZ), Heidelberg, Germany
- ¹⁷Department of Public Health Sciences, and Cancer Research Institute, Queen's University, Kingston, Ontario, Canada
- ¹⁸Department of Breast Medical Oncology, University of Texas MD Anderson Cancer Center, Houston, Texas
- ¹⁹Department of OB/GYN and Comprehensive Cancer Center, Medical University of Vienna, Vienna, Austria
- ²⁰Cancer Prevention Program, Fred Hutchinson Cancer Research Center, Seattle, Washington
- ²¹Zilber School of Public Health, University of Wisconsin-Milwaukee, Milwaukee, Wisconsin
- ²²Unit of Medical Genetics, Department of Medical Oncology and Hematology, Fondazione IRCCS Istituto Nazionale dei Tumori di Milano, Milan, Italy
- ²³High Risk and Cancer Prevention Group, Vall d'Hebron Institute of Oncology, Barcelona, Spain
- ²⁴Department of Gynecology and Obstetrics, Comprehensive Cancer Center ER‐EMN, University Hospital Erlangen, Friedrich‐Alexander‐University Erlangen‐Nuremberg, Erlangen, Germany
- ²⁵Division of Cancer Epidemiology, German Cancer Research Center (DKFZ), Heidelberg, Germany
- ²⁶Centro de Investigaci-n en Red de Enfermedades Raras (CIBERER), Madrid, Spain
- ²⁷Human Cancer Genetics Programme, Spanish National Cancer Research Centre (CNIO), Madrid, Spain
- ²⁸Institute of Biochemistry and Genetics, Ufa Federal Research Centre of the Russian Academy of Sciences, Ufa, Russia
- ²⁹Department of Genetics and Pathology, Pomeranian Medical University, Szczecin, Poland
- 30Fundaci—n Pœblica Galega Medicina Xen—mica, Santiago De Compostela, Spain

¹Program in Genetic Epidemiology and Statistical Genetics, Harvard T. H. Chan School of Public Health, Boston, Massachusetts

4 | **IVALUE V FENG** ET AL.

³¹Instituto de Investigacion Sanitaria de Santiago de Compostela, Santiago de Compostela, Spain

³²Department of Oncology, Helsinki University Hospital, University of Helsinki, Helsinki, Finland

³³Department of Oncology, University Hospital, Karolinska Institute, Stockholm, Sweden

34VIB Center for Cancer Biology, VIB, Leuven, Belgium

³⁵Laboratory for Translational Genetics, Department of Human Genetics, University of Leuven, Leuven, Belgium

³⁶Department of Radiation Oncology, Hannover Medical School, Hannover, Germany

³⁷Gynaecology Research Unit, Hannover Medical School, Hannover, Germany

³⁸NN Alexandrov Research Institute of Oncology and Medical Radiology, Minsk, Belarus

³⁹Copenhagen General Population Study, Herlev and Gentofte Hospital, Copenhagen University Hospital, Herlev, Denmark

40Department of Clinical Biochemistry, Herlev and Gentofte Hospital, Copenhagen University Hospital, Herlev, Denmark

⁴¹Faculty of Health and Medical Sciences, University of Copenhagen, Copenhagen, Denmark

⁴²Division of Cancer Prevention and Genetics, IEO, European Institute of Oncology IRCCS, Milan, Italy

43Department of Oncology, Lund University and Skåne University Hospital, Lund, Sweden

44Dr. Margarete Fischer‐Bosch‐Institute of Clinical Pharmacology, Stuttgart, Germany

45iFIT-Cluster of Excellence, University of Tuebingen, Tuebingen, Germany

46German Cancer Consortium (DKTK), German Cancer Research Center (DKFZ), Heidelberg, Germany

⁴⁷Division of Preventive Oncology, German Cancer Research Center (DKFZ) and National Center for Tumor Diseases (NCT), Heidelberg, Germany

48Institute of Human Genetics, Pontificia Universidad Javeriana, Bogota, Colombia

49Medical Faculty, Universidad de La Sabana, Bogota, Colombia

50Division of Molecular Pathology, The Netherlands Cancer Institute—Antoni van Leeuwenhoek Hospital, Amsterdam, The Netherlands

⁵¹Institute for Prevention and Occupational Medicine of the German Social Accident Insurance, Institute of the Ruhr University Bochum (IPA), Bochum, Germany

52Molecular Epidemiology Group, German Cancer Research Center (DKFZ), Heidelberg, Germany

⁵³Molecular Biology of Breast Cancer, University Womens Clinic Heidelberg, University of Heidelberg, Heidelberg, Germany

⁵⁴Medical Oncology Department, Hospital Cl'nico San Carlos, Instituto de Investigaci—n Sanitaria San Carlos (IdISSC), Centro Investigación Biomédica en Red de Cáncer (CIBERONC), Madrid, Spain

⁵⁵Section of Molecular Genetics, Dept, of Laboratory Medicine, University Hospital of Pisa, Pisa, Italy

56Research Department, Peter MacCallum Cancer Center, Melbourne, Victoria, Australia

⁵⁷Sir Peter MacCallum Department of Oncology, The University of Melbourne, Melbourne, Victoria, Australia

⁵⁸Division of Molecular Carcinogenesis, The Netherlands Cancer Institute—Antoni van Leeuwenhoek hospital, Amsterdam, The Netherlands

⁵⁹Genomic Epidemiology Group, German Cancer Research Center (DKFZ), Heidelberg, Germany

60Behavioral and Epidemiology Research Group, American Cancer Society, Atlanta, Georgia

⁶¹Department of Gynaecological Oncology, Chris OÕBrien Lifehouse and The University of Sydney, Camperdown, New South Wales, Australia

⁶²Oncology and Genetics Unit, Instituto de Investigacion Sanitaria Galicia Sur (IISGS), Xerencia de Xestion Integrada de Vigo-SERGAS, Vigo, Spain

63Cancer Epidemiology Group, University Cancer Center Hamburg (UCCH), University Medical Center Hamburg-Eppendorf, Hamburg, Germany

⁶⁴Division of Cancer Epidemiology and Genetics, Department of Health and Human Services, National Cancer Institute, National Institutes of Health, Bethesda, Maryland

65Departments of Pediatrics and Medicine, Columbia University, New York, New York

⁶⁶Centre for Medical Genetics, Ghent University, Gent, Belgium

⁶⁷Westmead Institute for Medical Research, University of Sydney, Sydney, New South Wales, Australia

68Department of Tumour Biology, INSERM U830, Paris, France

69Institut Curie, Paris, France

70Mines ParisTech, Fontainebleau, France

 $⁷¹$ Center for Hereditary Breast and Ovarian Cancer, Faculty of Medicine and University Hospital Cologne, University of Cologne, Cologne, Germany</sup>

⁷²Department of Laboratory Medicine and Pathology, Mayo Clinic, Rochester, Minnesota

⁷³Department of Oncology and Metabolism, Sheffield Institute for Nucleic Acids (SInFoNiA), University of Sheffield, Sheffield, UK

⁷⁴ Academic Unit of Pathology, Department of Neuroscience, University of Sheffield, Sheffield, UK

⁷⁵Department of Medical Epidemiology and Biostatistics, Karolinska Institutet, Stockholm, Sweden

⁷⁶Department of Clinical Genetics, Fox Chase Cancer Center, Philadelphia, Pennsylvania

⁷⁷Department of Pathology, Leiden University Medical Center, Leiden, The Netherlands

⁷⁸Department of Human Genetics, Leiden University Medical Center, Leiden, The Netherlands

 79 Hereditary Cancer Genetics Group, Area of Clinical and Molecular Genetics, Vall dHebron Institute of Oncology (VHIO), University Hospital Vall d'Hebron, Barcelona, Spain

80Clinical and Molecular Genetics Area, University Hospital Vall dHebron, Barcelona, Spain

⁸¹Department of Medicine, Abramson Cancer Center, Perelman School of Medicine at the University of Pennsylvania, Philadelphia, Pennsylvania

82Department of Non-Communicable Disease Epidemiology, London School of Hygiene and Tropical Medicine, London, UK

83Centre for Cancer Genetic Epidemiology, Department of Oncology, University of Cambridge, Cambridge, UK

84Department of Biomedical Sciences, Faculty of Science and Technology, University of Westminster, London, UK

⁸⁵Cancer Sciences Academic Unit, Faculty of Medicine, University of Southampton, Southampton, UK

86Department of Oncology, Rigshospitalet, Copenhagen University Hospital, Copenhagen, Denmark

87Department of Cancer Epidemiology, Clinical Sciences, Lund University, Lund, Sweden

88Institute for Medical Informatics, Statistics and Epidemiology, University of Leipzig, Leipzig, Germany

89LIFE ‐ Leipzig Research Centre for Civilization Diseases, University of Leipzig, Leipzig, Germany

⁹⁰David Geffen School of Medicine, Department of Medicine Division of Hematology and Oncology, University of California at Los Angeles, Los Angeles, California

⁹¹The Breast Cancer Now Toby Robins Research Centre, The Institute of Cancer Research, London, UK

⁹²Department of Breast Surgery, Herlev and Gentofte Hospital, Copenhagen University Hospital, Herlev, Denmark

93Molecular Diagnostics Laboratory, INRASTES, National Centre for Scientific Research 'Demokritos', Athens, Greece

94The Susanne Levy Gertner Oncogenetics Unit, Chaim Sheba Medical Center, Ramat Gan, Israel

⁹⁵Sackler Faculty of Medicine, Tel Aviv University, Ramat Aviv, Israel

⁹⁶School of Public Health, Curtin University, Perth, Western Australia, Australia

⁹⁷Schools of Medicine and Public Health, Division of Cancer Prevention & Control Research, Jonsson Comprehensive Cancer Centre, UCLA, Los Angeles, California

98Cancer Risk and Prevention Clinic, Dana‐Farber Cancer Institute, Boston, Massachusetts

⁹⁹Division of Genetics and Epidemiology, Institute of Cancer Research, London, UK

100Cancer Epidemiology Division, Cancer Council Victoria, Melbourne, Victoria, Australia

¹⁰¹Centre for Epidemiology and Biostatistics, Melbourne School of Population and Global Health, The University of Melbourne, Melbourne, Victoria, Australia

¹⁰²Department of Epidemiology and Preventive Medicine, Monash University, Melbourne, Victoria, Australia

¹⁰³Department of Pathology and Laboratory Medicine, Kansas University Medical Center, Kansas City, Kanas

¹⁰⁴Department of Medicine, McGill University, Montreal, Quebec, Canada

¹⁰⁵Division of Clinical Epidemiology, Royal Victoria Hospital, McGill University, Montreal, Quebec, Canada

¹⁰⁶Department of Dermatology, Huntsman Cancer Institute, University of Utah School of Medicine, Salt Lake City, Utah

¹⁰⁷Clinical Genetics Branch, Division of Cancer Epidemiology and Genetics, National Cancer Institute, Bethesda, Maryland

¹⁰⁸Cancer & Environment Group, Center for Research in Epidemiology and Population Health (CESP), INSERM, University Paris-Sud, University Paris‐Saclay, Villejuif, France

¹⁰⁹Department of Preventive Medicine, Keck School of Medicine, University of Southern California, Los Angeles, California

¹¹⁰Department of Oncology, Sšdersjukhuset, Stockholm, Sweden

111 Molecular Genetics of Breast Cancer, German Cancer Research Center (DKFZ), Heidelberg, Germany

¹¹²City of Hope Clinical Cancer Genetics Community Research Network, Duarte, California

¹¹³School of Population and Global Health, The University of Western Australia, Perth, Western Australia, Australia

¹¹⁴Family Cancer Clinic, The Netherlands Cancer Institute—Antoni van Leeuwenhoek Hospital, Amsterdam, The Netherlands

¹¹⁵Department of Medical Oncology, Family Cancer Clinic, Erasmus MC Cancer Institute, Rotterdam, The Netherlands

116Center for Medical Genetics, NorthShore University HealthSystem, Evanston, Illinois

¹¹⁷The University of Chicago Pritzker School of Medicine, Chicago, Illinois

118NN Petrov Institute of Oncology, St. Petersburg, Russia

¹¹⁹Australian Breast Cancer Tissue Bank, Westmead Institute for Medical Research, University of Sydney, Sydney, New South Wales, Australia

¹²⁰The Hereditary Breast and Ovarian Cancer Research Group Netherlands (HEBON), Coordinating Center, The Netherlands Cancer Institute, Amsterdam, The Netherlands

¹²¹Department of Medicine, Division of Oncology, Stanford Cancer Institute, Stanford University School of Medicine, Stanford, California

 $\overline{6}$ $\overline{1471}$ $\overline{151}$

¹²²Ontario Cancer Genetics Network, Lunenfeld-Tanenbaum Research Institute of Mount Sinai Hospital, Toronto, Ontario, Canada

¹²³Lombardi Comprehensive Cancer Center, Georgetown University, Washington, District of Columbia

¹²⁴Research Centre for Genetic Engineering and Biotechnology 'Georgi D, Efremov', Macedonian Academy of Sciences and Arts, Skopje Republic of North Macedonia, North Macedonia

¹²⁵Independent Laboratory of Molecular Biology and Genetic Diagnostics, Pomeranian Medical University, Szczecin, Poland

¹²⁶Parkville Familial Cancer Centre, Peter MacCallum Cancer Center, Melbourne, Victoria, Australia

¹²⁷State Research Institute Innovative Medicine Center, Vilnius, Lithuania

¹²⁸Department of Medicine, Division of Hematology/Oncology, UPMC Hillman Cancer Center, University of Pittsburgh School of Medicine, Pittsburgh, Pennsylvania

¹²⁹Clinical Genetics Research Lab, Department of Cancer Biology and Genetics, Memorial Sloan-Kettering Cancer Center, New York, New York

¹³⁰David Geffen School of Medicine, Department of Obstetrics and Gynecology, University of California, Los Angeles, California

¹³¹Department of Genetics and Fundamental Medicine, Bashkir State Medical University, Ufa, Russia

¹³²Department of Obstetrics and Gynecology, Helsinki University Hospital, University of Helsinki, Helsinki, Finland

133 Department of Cancer Genetics, Institute for Cancer Research, Oslo University Hospital-Radiumhospitalet, Oslo, Norway

¹³⁴Institute of Clinical Medicine, Faculty of Medicine, University of Oslo, Oslo, Norway

¹³⁵Molecular Diagnostic Unit, Hereditary Cancer Program, ICO-IDIBELL (Bellvitge Biomedical Research Institute, Catalan Institute of Oncology), CIBERONC, Barcelona, Spain

136Departement de Ge netique, CHU de Grenoble, Grenoble, France

¹³⁷Genetic Epidemiology of Cancer Team, Inserm U900, Paris, France

¹³⁸Department of Health Sciences Research, Mayo Clinic, Scottsdale, Arizona

¹³⁹Department of Epidemiology, University of Washington School of Public Health, Seattle, Washington

¹⁴⁰Public Health Sciences Division, Fred Hutchinson Cancer Research Center, Seattle, Washington

¹⁴¹Translational Cancer Research Area, University of Eastern Finland, Kuopio, Finland

¹⁴²Institute of Clinical Medicine, Pathology and Forensic Medicine, University of Eastern Finland, Kuopio, Finland

¹⁴³Imaging Center, Department of Clinical Pathology, Kuopio University Hospital, Kuopio, Finland

¹⁴⁴Department of Clinical Science and Education, Sšdersjukhuset, Karolinska Institutet, Stockholm, Sweden

145Moores Cancer Center, University of California San Diego, La Jolla, California

¹⁴⁶Department of Family Medicine and Public Health, University of California San Diego, La Jolla, California

¹⁴⁷Immunology and Molecular Oncology Unit, Veneto Institute of Oncology ÊIOV—IRCCS, Padua, Italy

¹⁴⁸Department of Medical Oncology, University Hospital of Heraklion, Heraklion, Greece

¹⁴⁹Department of Gynecology and Obstetrics, Ludwig Maximilian University of Munich, Munich, Germany

150MRC Clinical Trials Unit at UCL, Institute of Clinical Trials & Methodology, University College London, London, UK

¹⁵¹Department of Electron Microscopy/Molecular Pathology and The Cyprus School of Molecular Medicine, The Cyprus Institute of Neurology & Genetics, Nicosia, Cyprus

¹⁵²Faculty of Medicine, University of Heidelberg, Heidelberg, Germany

¹⁵³NRG Oncology, Statistics and Data Management Center, Roswell Park Cancer Institute, Buffalo, New York

¹⁵⁴Department of Laboratory Medicine and Pathobiology, University of Toronto, Toronto, Ontario, Canada

155Laboratory Medicine Program, University Health Network, Toronto, Ontario, Canada

156 Department of Population Sciences, Beckman Research Institute of City of Hope, Duarte, California

¹⁵⁷Leuven Multidisciplinary Breast Center, Department of Oncology, Leuven Cancer Institute, University Hospitals Leuven, Leuven, Belgium

158Center for Genomic Medicine, Rigshospitalet, Copenhagen University Hospital, Copenhagen, Denmark

¹⁵⁹Latvian Biomedical Research and Study Centre, Riga, Latvia

¹⁶⁰Clinical Genetics Service, Department of Medicine, Memorial Sloan-Kettering Cancer Center, New York, New York

¹⁶¹Department of Molecular Genetics, National Institute of Oncology, Budapest, Hungary

¹⁶²Center for Clinical Cancer Genetics, The University of Chicago, Chicago, Illinois

¹⁶³Department of Genetics and Computational Biology, QIMR Berghofer Medical Research Institute, Brisbane, Queensland, Australia

¹⁶⁴Section of Molecular Diagnostics, Clinical Biochemistry, Aalborg University Hospital, Aalborg, Denmark

¹⁶⁵Department of Genetics, Portuguese Oncology Institute, Porto, Portugal

¹⁶⁶Genome Diagnostics Program, IFOM—The FIRC (Italian Foundation for Cancer Research) Institute of Molecular Oncology, Milan, Italy

¹⁶⁷Department of Medicine Oncology, Peter MacCallum Cancer Centre, Melbourne, Victoria, Australia

¹⁶⁸Laboratory of Cancer Genetics and Tumor Biology, Cancer and Translational Medicine Research Unit, Biocenter Oulu, University of Oulu, Oulu, Finland ¹⁶⁹Laboratory of Cancer Genetics and Tumor Biology, Northern Finland Laboratory Centre Oulu, Oulu, Finland ¹⁷⁰Unit of Molecular Bases of Genetic Risk and Genetic Testing, Department of Research, Fondazione IRCCS Istituto Nazionale dei Tumori (INT), Milan, Italy ¹⁷¹Clinical Genetics, Karolinska Institutet, Stockholm, Sweden 172 Department of Basic Sciences, Shaukat Khanum Memorial Cancer Hospital and Research Centre (SKMCH & RC), Lahore, Pakistan ¹⁷³Department of Chronic Disease Epidemiology, Yale School of Public Health, New Haven, Connecticut ¹⁷⁴Medical Oncology Department, Hospital Universitario Puerta de Hierro, Madrid, Spain ¹⁷⁵Department of Oncology, University Hospital of Larissa, Larissa, Greece ¹⁷⁶Epidemiology Branch, National Institute of Environmental Health Sciences, NIH, Research Triangle Park, North Carolina ¹⁷⁷Research Oncology, GuyÕs Hospital, King's College London, London, UK ¹⁷⁸Division of Psychosocial Research and Epidemiology, The Netherlands Cancer Institute—Antoni van Leeuwenhoek Hospital, Amsterdam, The Netherlands ¹⁷⁹Faculty of Information Technology, Monash University, Melbourne, Victoria, Australia ¹⁸⁰Center for Molecular Medicine Cologne (CMMC), Faculty of Medicine and University Hospital Cologne, University of Cologne, Cologne, Germany ¹⁸¹Division of Molecular Medicine, Pathology North, John Hunter Hospital, Newcastle, New South Wales, Australia ¹⁸²Discipline of Medical Genetics, School of Biomedical Sciences and Pharmacy, Faculty of Health, University of Newcastle, Callaghan, New South Wales, Australia

¹⁸³Hunter Medical Research Institute, John Hunter Hospital, Newcastle, New South Wales, Australia

¹⁸⁴Department of Internal Medicine, Division of Medical Oncology, University of Kansas Medical Center, Westwood, Kanas

¹⁸⁵Genomics Center, Centre Hospitalier Universitaire de Quebec–Universite Laval, Research Center, Quebec City, Qubec, Canada

¹⁸⁶Department of Clinical Genetics, Aarhus University Hospital, Aarhus N, Denmark

¹⁸⁷Precision Medicine, School of Clinical Sciences at Monash Health, Monash University, Clayton, Victoria, Australia

¹⁸⁸Department of Clinical Pathology, The University of Melbourne, Melbourne, Victoria, Australia

¹⁸⁹Population Oncology, BC Cancer, Vancouver, British of Columbia, Canada

190School of Population and Public Health, University of British Columbia, Vancouver, British of Columbia, Canada

¹⁹¹The Curtin UWA Centre for Genetic Origins of Health and Disease, Curtin University and University of Western Australia, Perth, Western Australia, Australia

¹⁹²Division of Breast Cancer Research, The Institute of Cancer Research, London, UK

¹⁹³Faculty of Medicine, University of Southampton, Southampton, UK

¹⁹⁴Epigenetic and Stem Cell Biology Laboratory, National Institute of Environmental Health Sciences, NIH, Research Triangle Park, North Carolina ¹⁹⁵Biomedical Sciences Institute (ICBAS), University of Porto, Porto, Portugal

¹⁹⁶Department of Epidemiology, Mailman School of Public Health, Columbia University, New York, New York

¹⁹⁷Genetic Counseling Unit, Hereditary Cancer Program, IDIBELL (Bellyitge Biomedical Research Institute), Catalan Institute of Oncology, CIBERONC, Barcelona, Spain

¹⁹⁸Department of Clinical Genetics, Odense University Hospital, Odence C, Denmark

¹⁹⁹Department of Medicine, Magee-Womens Hospital, University of Pittsburgh School of Medicine, Pittsburgh, Pennsylvania

²⁰⁰Program in Cancer Genetics, Departments of Human Genetics and Oncology, McGill University, Montreal, Quebec, Canada

²⁰¹Department of Medical Genetics, University of Cambridge, Cambridge, UK

²⁰²Department of Cancer Biology and Genetics, The Ohio State University, Columbus, Ohio

²⁰³Department of Surgery, Leiden University Medical Center, Leiden, The Netherlands

²⁰⁴Department of Medical Oncology, Beth Israel Deaconess Medical Center, Boston, Massachusetts

²⁰⁵Department of Health Science Research, Division of Epidemiology, Mayo Clinic, Rochester, Minnesota

²⁰⁶Department of Clinical Genetics, Leiden University Medical Center, Leiden, The Netherlands

²⁰⁷Department of Clinical Genetics, Erasmus University Medical Center, Rotterdam, The Netherlands

208Department of Genetics, University of Pretoria, Arcadia, South Africa

²⁰⁹Division of Functional Onco-genomics and Genetics, Centro di Riferimento Oncologico di Aviano (CRO), IRCCS, Aviano, Italy

210Hospital Clínico Universitario (SERGAS), Universidad de Santiago de Compostela, CIMUS, Santiago de Compostela, España

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²¹¹Biostatistics and Computational Biology Branch, National Institute of Environmental Health Sciences, NIH, Research Triangle Park, North Carolina

212Clinical Cancer Genomics, City of Hope, Duarte, California

Correspondence

Xia Jiang, Harvard T. H. Chan School of Public Health, Building 2‐205, 677 Huntington Avenue, Boston, MA 02115. Email: xiajiang@hsph.harvard.edu

Abstract

Previous transcriptome‐wide association studies (TWAS) have identified breast cancer risk genes by integrating data from expression quantitative loci and genome‐wide association studies (GWAS), but analyses of breast cancer subtype‐specific associations have been limited. In this study, we conducted a TWAS using gene expression data from GTEx and summary statistics from the hitherto largest GWAS meta‐analysis conducted for breast cancer overall, and by estrogen receptor subtypes (ER+ and ER−). We further compared associations with ER+ and ER− subtypes, using a case‐only TWAS approach. We also conducted multigene conditional analyses in regions with multiple TWAS associations. Two genes, STXBP4 and HIST2H2BA, were specifically associated with ER+ but not with ER– breast cancer. We further identified 30 TWAS‐ significant genes associated with overall breast cancer risk, including four that were not identified in previous studies. Conditional analyses identified single independent breast-cancer gene in three of six regions harboring multiple TWAS‐significant genes. Our study provides new information on breast cancer genetics and biology, particularly about genomic differences between ER+ and ER− breast cancer.

KEYWORDS

breast cancer subtype, causal gene, GWAS, TWAS

1 | INTRODUCTION

Breast cancer is the most common malignancy among women worldwide (Bray et al., [2018](#page-24-0)). The disease has a strong inherited component (Beggs & Hodgson, [2009](#page-24-1)); linkage studies have identified infrequent mutations in BRCA1/2 (Easton et al., [2007;](#page-24-2) Seal et al., [2006](#page-25-0); Turnbull et al., [2010](#page-25-1)) and genome‐wide association studies (GWAS) have identified 177 susceptibility loci to date (Michailidou et al., [2017](#page-25-2)). However, these GWAS‐ discovered variants explain only 18% of the familial relative risk of breast cancer. Moreover, the causal mechanism driving GWAS associations remains largely unknown, as many variants are located in noncoding or intergenic regions, and are not in strong linkage disequilibrium (LD) with known protein‐coding variants (Beggs & Hodgson, [2009;](#page-24-1) Michailidou et al., [2015\)](#page-25-3).

Breast cancer is a heterogeneous disease consisting of several well-established subtypes. One of the most important markers of breast cancer subtypes is estrogen receptor (ER) status. ER+ and ER− tumors differ in etiology (X. R. Yang, Chang‐Claude, et al., [2011](#page-25-4)), genetic

predisposition (Mavaddat, Antoniou, Easton, & Garcia‐ Closas, [2010\)](#page-25-5), and clinical behavior (Blows et al., [2010\)](#page-24-3). ER− tumor occurs more often among younger women, and patients are more likely to carry BRCA1 pathogenic variants (Atchley et al., [2008;](#page-24-4) Garcia‐Closas et al., [2013\)](#page-24-5). ER− tumor also has worse short‐term prognosis. Among the 177 GWAS‐identified breast cancer‐associated single nucleotide polymorphisms (SNPs), around 50 are more strongly associated with ER+ disease and 20 are more strongly associated with ER− disease (Michailidou et al., [2017](#page-25-2); Milne et al., [2017](#page-25-6)).

SNPs associated with complex traits are more likely to be in regulatory regions than in protein‐coding regions, and many of these SNPs are also associated with expression levels of nearby genes (Nicolae et al., [2010\)](#page-25-7). For example, breast cancer GWAS‐identified variants at 6q25.1 regulate ESR1, but also coregulate other local genes such as RMND1, ARMT1, and CCDC170 (Dunning et al., [2016,](#page-24-6) p. 1). These results suggest that by integrating genotype, phenotype, and gene expression, we can identify novel trait‐associated genes and understand biological mechanisms. However, due to costs and tissue

availability, acquiring GWAS and gene expression data for the same set of individuals remains challenging.

A recently published approach, referred to as transcriptome‐wide association study (TWAS; Gamazon et al., [2015](#page-24-7); Gusev et al., [2016](#page-24-8)), overcomes these difficulties by using a relatively small set of reference individuals for whom both gene expression and SNPs have been measured to impute the cis‐genetic component of expression for a much larger set of individuals from their GWAS summary statistics. The association between the predicted gene expression and traits can then be tested. This method has been shown to have greater power relative to GWAS; and has identified 1,196 trait-associated genes across 30 complex traits in a recently performed multitissue TWAS (Mancuso et al., [2017](#page-25-8)).

To date, three TWAS of breast cancer have been conducted (Gao, Pierce, Olopade, Im, & Huo, [2017;](#page-24-9) Hoffman et al., [2017;](#page-25-9) Wu et al., [2018](#page-25-10)). A fourth study linked expression quantitative loci (eQTL) data across multiple tissues and breast cancer GWAS results using EUGENE, a statistical approach that sums evidence for association with disease across eQTLs regardless of directionality. That study then tested EUGENE‐significant genes using a TWAS statistic, which does take directionality into account (Ferreira et al., [2019](#page-24-10)). The two earliest TWAS used GWAS data from the National Cancer Institute's "Up for a Challenge" competition, which included data from 12,100 breast cancer cases (of which 3,900 had ER− disease) and 11,400 controls, as well as eQTL data from breast tissue and whole blood from the GTEx and DGN projects (Gao et al., [2017](#page-24-9); Hoffman et al., [2017\)](#page-25-9). The subsequent TWAS by Wu et al. ([2018](#page-25-10)) and the EUGENE analysis by Ferreira et al. [\(2019\)](#page-24-10) used results from a much larger GWAS conducted by the Breast Cancer Association Consortium (BCAC), which included 122,977 cases (of which 21,468 had ER− disease) and 105,974 controls. Together, these four studies have identified 59 genes whose predicted expression levels are associated with risk of overall breast cancer, and five associated with risk of ER− disease. Of these 64 genes, 30 are at loci not previously identified by breast cancer GWAS.

These previous TWAS largely focused on overall breast cancer risk. Analyses of ER− disease either were conducted using a small sample size (Gao et al., [2017](#page-24-9)) or did not scan all genes using a directional TWAS approach (Ferreira et al., [2019](#page-24-10)). Moreover, none of the previous analyses considered ER+ disease specifically or examined differences in association between predicted gene expression and ER+ versus ER− disease.

The interpretation of TWAS results is not straightforward (Wainberg et al., [2019](#page-25-11)). Specifically, TWAS statistic by itself cannot distinguish between a mediated effect (SNPs influence breast cancer risk by changing the expression of the tested gene), pleiotropy (SNPs associated with gene expression also influence breast cancer risk through another mechanism), or colocalization (SNPs associated with gene expression are in LD with other SNPs that influence breast cancer risk through another mechanism). Previous studies have conducted limited sensitivity analyses (e.g., Wu et al., [2018](#page-25-10) and Ferreira et al., [2019](#page-24-10) conditioned the TWAS tests on lead GWAS SNPs), but the genetic architecture at TWAS‐ identified loci remains largely unclear.

In the current analysis, we complement previous work by conducting a TWAS for overall breast cancer and for ER+ and ER− subtypes. We also applied a case‐only TWAS test to identify predicted transcript levels that were differentially associated with ER+ and ER− disease. We conducted expanded sensitivity analyses, conditioning on multiple TWAS‐ significant genes in a region to account for possible confounding due to LD (colocalization). We chose to focus on the expression of normal breast tissue of European ancestry women to maximize specificity and identify good targets for near‐term follow‐up experiments in mammary cells. One advantage of using a biologically relevant tissue is that it both increases the a priori plausibility of observed associations and increases the likelihood that genes with observed associations will be expressed and influence tumor development in cells from the target tissue. We have reproduced previous results (Ferreira et al., [2019;](#page-24-10) Wu et al., [2018\)](#page-25-10) and provided evidence regarding the independent associations of multiple genes in regions containing one or more TWAS‐significant genes. We also identified genes with subtype‐specific associations, highlighting different biological mechanisms likely underlying the disease subtypes.

2 | MATERIAL AND METHODS

2.1 | Gene expression reference panel

The transcriptome and high-density genotyping data used to build the gene expression model (reference panel) were retrieved from GTEx (GTEx Consortium, [2015\)](#page-24-11), a consortium collected high‐quality gene expression RNA‐seq data across 44 body sites from 449 donors, and genome‐wide genetic information. For the current study, we included 67 women of European ancestry who provided normal breast mammary tissues. RNA samples extracted from tissues were sequenced to generate data on 12,696 transcripts. Genomic DNA samples were genotyped using Illumina OMNI 5 M or 2.5 M arrays, processed with a standard GTEx protocol. Briefly, SNPs with call rates <98%, with differential missingness between the two array experiments (5 M or 2.5 M), with Hardy-Weinberg equilibrium $p < 10^{-6}$ or showing batch effects were excluded. The genotypes were then imputed to the Haplotype Reference Consortium reference panel (McCarthy et al., [2016](#page-25-12)) using Minimac3 for imputation and SHAPEIT for pre‐phasing (Delaneau, Marchini, & Zagury, [2011;](#page-24-12) Howie, Donnelly, & Marchini, [2009\)](#page-25-13). Only SNPs with high imputation quality ($r^2 \geq .8$), minor allele frequency (MAF) \geq 0.05, and were included in the Hap-Map Phase 2 version were used to build the expression prediction models.

2.2 | Breast cancer meta‐GWAS data

The GWAS breast cancer summary‐level data were mainly provided by the Breast Cancer Association Consortium (BCAC; Michailidou et al., [2017\)](#page-25-2), as well as the Consortium of Investigators of Modifiers of BRCA1/2 (CIMBA). BCAC conducted the largest breast cancer meta‐GWAS to date (referred as the overall breast cancer GWAS analysis). The BCAC included 122,977 cases and 105,974 controls of European ancestry. Among these, 46,785 cases and 42,892 controls were genotyped using the Illumina iSelect genotyping array (iCOGS) on 211,155 SNPs; and 61,282 cases and 45,494 controls were genotyped using the Illumina OncoArray on 570,000 SNPs (Yovel, Franz, Stilz, & Schnitzler, [2008\)](#page-26-0). The study also included data from 11 other GWAS on 14,910 cases and 17,588 controls. Genetic data for all individual participating studies were imputed to the 1000 Genomes Project Phase 3 v5 EUR reference panel Logistic regression was fitted to estimate per‐allele odds ratios (ORs), adjusting for country and top principal components (PCs). Inverse variance fixed‐effect meta‐analysis was used to combine the genetic association for breast cancer risk across studies (Milne et al., [2017](#page-25-6)). In CIMBA, genotypes were generated by the Illumina OncoArray and imputed to the 1000 Genomes Project Phase 3 v5 EUR reference panel (Amos et al., [2016\)](#page-23-0). A retrospective cohort analysis framework was adopted to estimate per‐allele hazard ratios (HRs), modelling time‐to‐breast‐cancer and stratified by country, Ashkenazi Jewish origin and birth cohort (Antoniou et al., [2005](#page-24-13); Barnes et al., [2012\)](#page-24-14). Fixed‐effect meta‐analysis (Willer, Li, & Abecasis, [2010\)](#page-25-14) was performed to combine results across genotyping initiatives within the two consortia, assuming that the OR and HR estimates had roughly the same underlying relative risk. We restricted subsequent analyses to SNPs with an imputation r^2 > .3, and an MAF > 0.005 across all platforms were included in the analysis (approximately 11.5 M).

For the ER+ subtype, meta‐GWAS summary data based on 69,501 ER+ cases and 105,974 controls (part of the overall breast cancer samples) were included and analyzed (Mavaddat et al., [2015\)](#page-25-15). For the ER− subtype,

meta‐GWAS summary data based on 21,468 ER− cases and 105,974 controls from the BCAC were combined with 9,414 additional BRCA1 mutation-positive cases and 9,494 BRCA1 mutation-positive controls from CIMBA (Milne et al., [2017](#page-25-6)).

To distinguish different genetic signals between ER+ and ER− subtypes, we further retrieved GWAS summary‐ level data on a case‐only GWAS, which compared ER+ patients (sample size: 23,330 in iCOGs and 44,746 in OncoArray) to ER− patients (sample size: 5,479 and 11,856; Milne et al., [2017\)](#page-25-6). Logistic regression was performed to test the association between genetic variants with known ER status in the two studies separately, adjusting for substudy and top PCs for iCOGs, and patients' countries and top PCs for OncoArray. Results were then combined using a fixed‐effect meta‐analysis.

2.3 | Constructing expression weights

Before constructing the expression model (using GTEx data, regress gene expression on SNPs), we set several criteria to select eligible candidate genes for inclusion in the model (from the total 12,696 transcripts). We used a REML algorithm implemented in GCTA to estimate the *cis* (500 basepair window surrounding transcription start site) SNP‐ heritability (cis- h_g^2) for each transcript expression (Cai et al., [2014](#page-24-15); J. Yang et al., [2010](#page-25-16)). Only genes with significant heritability (nominal $p \leq 0.01$) were included in the subsequent model construction (J. Yang, Lee, Goddard, & Visscher, [2011](#page-25-17)). The p values for null hypotheses $cis-h_g^2 = 0$ were computed using a likelihood ratio test. To account for population stratification, 20 PCs were always included as fixed effects. Consistent with previous research (ENCODE Project Consortium, [2012](#page-24-16); J. Yang et al., [2010\)](#page-25-16), we observed strong evidence for $cis-h_g^2$ on many genes (significantly non‐zero for 1,355 genes).

We then constructed linear genetic predictors of gene expression for these genes. We performed five models: Bayesian Sparse Linear Mixed model, Best Linear Unbiased Predictor model, Elastic‐net regression (with mixing parameter of 0.5), LASSO regression and Single best eQTL model. We used a fivefold cross-validation strategy to validate each model internally. Only genes with good model performance, corresponding to a prediction r^2 value (the square of the correlation between predicted and observed expression) of at least 1% (0.10 correlation) in at least one of the five models, were included in subsequent TWAS analyses. The weights were chosen from the best performed model out of the five models. We adopted this additional filter to improve the interpretability and specificity of results: significant TWAS results based on models with little or no predictive ability likely result from pleiotropy or

(Continues)

Abbreviations: ER, estrogen receptor; SNP, single nucleotide polymorphisms; TWAS, transcriptome‐wide association studies.

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> colocalization, not the effect of modeled gene's expression levels. This additional filter narrowed the number of candidate genes to 901.

2.4 | Transcriptome ‐wide association study (TWAS) analyses

Using the functional weights of those 901 genes and summary level GWAS data, we assessed the association between predicted gene expression and breast cancer risk. We performed summary ‐based imputation using the ImpG ‐ Summary algorithm (Pasaniuc et al., [2014](#page-25-18)). Briefly, let Z be a vector of standardized association statistics (z scores) of SNPs for a trait at a given *cis* locus, $\Sigma_{s,s}$ be the LD matrix from reference genotype data and let $W = (w_1 w_2 w_3 ... w_j)$ be the weights from the expression prediction model precompiled using the reference panel. Under the null hypothesis that none of the SNPs with $w_i \neq 0$ is associated with disease, the test statistic $wz / (w \Sigma_{s,s} w')^{1/2}$ follows a normal distribution with mean $= 0$ and variance $= 1$. To account for finite sample size and instances where $\Sigma_{s,s}$ was not invertible, we adjusted the diagonal of the matrix using a technique similar to ridge regression with $\lambda = 0.1$.

2.5 | Case ‐only TWAS

To assess whether genetically predicted expression was differentially associated with ER+ and ER- breast cancer, we applied the TWAS procedure described above to the Z statistics from the BCAC case ‐only analysis. Following arguments in Barfield et al. [\(2018\)](#page-24-17) the standard TWAS statistic applied to a case ‐only GWAS results tests hypothesis *H*₀: $\beta_2 - \beta_1 = 0$. This is similar to a conventional multinomial logistic model for subtype‐specific breast cancer risk, with expression log odds ratio β_2 for ER- disease and β_1 for ER+ disease, under which scenario, the expression log odds ratio comparing ER– to ER+ cases is $\beta_2 - \beta_1$.

2.6 | Conditional analyses

Colocalization makes the interpretation of TWAS hits challenging (Mancuso et al., [2019](#page-25-19); Wainberg et al., [2019](#page-25-11)). In addition to the main TWAS analysis, we also performed conditional and joint (COJO) multiple ‐SNP analysis at each TWAS significant gene location to distinguish colocalization, and to identify gene(s) independently responsible for the statistical association at each locus. COJO approximates the results of a joint conditional analysis including predicted expression levels from multiple proximal genes. The original COJO approach was designed to assess the association of

individual SNPs with a phenotype; we used an extension that jointly models the associations between multiple linear combinations of individual SNPs (Gusev et al., [2016](#page-24-8)). We conducted two types of COJO: (a) For regions in which multiple associated features were identified (within 500 kb of each other, i.e., colocalization), we jointly modeled these significant TWAS genes to determine the strongest associated gene (or infer independent signals); (b) To provide information on whether the TWAS gene was responsible for the observed SNP‐trait association, we also evaluated whether the GWAS‐identified index SNPs remained significant after conditioning on the genes within the same region.

3 | RESULTS

3.1 | Breast cancer TWAS

We selected 12,696 transcripts from the 67 GTEx breast tissue samples of European‐ancestry women that passed quality control. Based on GCTA‐REML analysis, breast‐tissue expression levels for 1,355 of these genes were heritable (*p* value for *cis-h*_g² < .01). We then built linear predictors for these heritable genes and estimated prediction r^2 using fivefold cross‐validation. A total of 454 genes failed our cross-validation r^2 requirement $(r^2 > .01)$, and we performed TWAS on the remaining 901 genes. We defined statistical significance for TWAS results as a marginal $p < 5.5 \times 10^{-5}$ (Bonferroni correction controlling the familywise error rate at ≤ 0.05 for the 901 genes).

First, to compare with previous GWAS findings and to demonstrate the validity of our results, we performed TWAS analysis in overall breast cancer. We identified 30 genes in 18 cytoband regions associated with breast cancer risk (Table [1](#page-10-0)). Of these regions, 11 (containing 21 genes) were previously reported breast cancer susceptibility loci (harboring one or more GWAS‐significant SNP). Five genes in the remaining seven regions were previously reported in TWAS or EUGENE analyses (LINC00886, CTD‐2323K18.1, MAN2C1, NUP107, and CPNE1), while the remaining four genes in these regions were novel (MAEA, GDI2, ULK3, and HSD17B1P1). NUP107 and CPNE1 did not pass a stringent Bonferroni significance threshold in Wu et al. ([2018\)](#page-25-10) but passed a lessstringent false discovery rate threshold.

We also carried out analyses focusing on breast cancer subtypes. We found 20 genes associated with ER+ breast cancer, and six genes associated with ER– breast cancer $(p < .05/901 = 5.5 \times 10^{-5})$; Table [1\)](#page-10-0). In our results, all genes associated with ER− disease were also associated with ER+ disease, as well as with overall breast cancer risk. Using a more stringent threshold on the strength of the genetic predictor for expression (cross-validation $r^2 > 0.1$; 383 genes passed this threshold), we found four TWAS significant $(p < .05/383 = 1.3 \times 10^{-4})$ genes for ER- disease, 14 genes for ER+ disease, and 19 genes for overall breast cancer (18 out of 19 genes are included in Table [1](#page-10-0) except for one gene, CTD/3110H11.1). As before, these gene sets were nested within each other.

3.2 | Difference of TWAS signal across breast cancer subtypes

We tested whether the imputed gene expression-breast cancer associations differed by subtype using GWAS summary statistics from a case-only analysis, which specifically compared ER+ with ER− breast cancer patients (see Section 2 for details), scanning through

FIGURE 1 Scatter plot comparing the transcriptome‐wide association study z scores in ER+ and ER− patients

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901 eligible genes. Two genes, HIST2H2BA and STXBP4, showed significant associations ($p < .05/901$) with ER status among cases (Figure [1](#page-12-0)). These two genes were associated with ER+ breast cancer but not associated with ER− breast cancer.

3.3 | GWAS signal conditioning on TWAS gene expression

As shown in Table [2,](#page-13-0) 21 (of 30) TWAS‐significant genes were located near GWAS signals. To examine whether the observed GWAS signal within the gene region could be explained by the expression of that gene, we performed additional analyses conditioning SNP‐cancer associations on the predicted expression of that particular significant TWAS gene (See Section 2 and Figure S1, for details). We found that for most regions, GWAS SNPs were no longer associated with the risk of breast cancer once conditioned on the expression of TWAS gene in the region: 15 of 21 genes had no SNPs with a conditional GWAS p value smaller than the genome‐wide significant threshold (5×10^{-8}) . Thus, there were six genes for which the

TABLE 2 Summary of conditional analysis at known breast cancer risk region

	Before conditional analysis			After conditional analysis				
Gene	Number of SNPs	Number of significant SNPs	Index GWAS SNP p value	Number of significant SNPs	Index SNP values	Smallest conditional p	Ratio ^a	Magnitude of change in the minimum p value before and after COJO
ALS2CR12	480	12	$8.2E - 17$	$\mathbf{1}$	rs3769823	$6.40E - 09$	0.92	$1.28E - 08$
ATG10	619	24	$6.9E - 13$	$\boldsymbol{0}$	rs891159	$1.20E - 07$	1.00	$5.75E - 06$
ATP6AP1L	581	24	$6.9E - 13$	$\mathbf{0}$	rs891159	$1.70E - 07$	1.00	$4.06E - 06$
CASP8	493	12	$8.2E - 17$	6	rs3769823	$3.90E - 12$	0.50	$2.10E - 05$
CRHR1	229	13	$1.5E - 10$	$\mathbf{0}$	rs17763086	$1.40E - 05$	1.00	$1.07E - 05$
CRHR1-IT1	230	13	$1.5E - 10$	$\boldsymbol{0}$	rs17763086	$9.90E - 05$	1.00	$1.52E - 06$
HIST2H2BA	202	19	$3.5E - 52$	$\mathbf{1}$	rs11249433	$7.40E - 24$	0.95	$4.73E - 29$
KANSL1-AS1	34	13	$1.5E - 10$	$\overline{0}$	rs17763086	$1.60E - 01$	1.00	$9.38E - 10$
L3MBTL3	724	13	$1.7E - 12$	$\mathbf{0}$	rs6569648	$1.40E - 03$	1.00	$1.21E - 09$
LRRC37A	285	13	$1.5E - 10$	$\mathbf{0}$	rs17763086	$4.60E - 05$	1.00	$3.26E - 06$
LRRC37A4P	285	13	$1.5E - 10$	$\mathbf{0}$	rs17763086	$4.60E - 05$	1.00	$3.26E - 06$
MRPL23-AS1	557	36	$2.4E - 33$	18	rs569550	$1.20E - 29$	0.50	$2.00E - 04$
NUDT17	112	17	$1.5E - 10$	$\boldsymbol{0}$	rs36107432	$3.90E - 05$	1.00	$3.85E - 06$
RP11-15A1.7	594	32	$1E-16$	$\overline{0}$	rs10426528	$4.60E - 07$	1.00	$2.17E - 10$
RP11-250B2.5	503	8	$2.7E - 09$	$\boldsymbol{0}$	rs9343989	$1.00E - 03$	1.00	$2.70E - 06$
RP11-554A11.9 532		36	$2.8E - 44$	33	rs680618	$2.80E - 44$	0.08	$1.00E + 00$
RP11-7306.3	665	13	$1.7E - 12$	$\mathbf{0}$	rs6569648	$1.70E - 03$	1.00	$1.00E - 09$
STXBP4	687	46	$2E - 28$	$\boldsymbol{0}$	rs244353	$2.10E - 04$	1.00	$9.52E - 25$
ZNF155	597	32	$1E-16$	5	rs10426528	$2.00E - 09$	0.84	$5.00E - 08$
ZNF404	551	32	$1E-16$	$\mathbf{0}$	rs10426528	$5.30E - 07$	1.00	$1.89E - 10$
LRRC37A2	152	$\overline{2}$	$2E - 08$	$\mathbf{0}$	rs199498	$1.80E - 02$	1.00	$1.11E - 06$

Abbreviations: COJO, conditional and joint; GWAS, genome‐wide association studies; SNP, single nucleotide polymorphisms.

a Proportion of marginally significant SNPs that are not significant in conditional analyses. Analysis was performed using GWAS summary statistics of ER+ subtypes. The difference between marginal SNP tests for association (GWAS p values) and the SNP p values conditional on significant TWAS genes provides some evidence regarding the independence of the TWAS and single-SNP association signals. The number and proportion of SNPs that are genome-wide significant before and after conditioning on a TWAS‐significant gene summarizes the degree single‐SNP associations are dependent on (or independent of) the TWAS association.

GWAS SNP remained significantly associated with breast cancer risk at the genome‐wide threshold (5×10^{-8}) after conditioning on TWAS gene expression. The region containing HIST2H2BA had only one genome‐wide significant SNP remaining, and the region containing ZNF155 and ZNF404 had five genome‐wide significant SNPs remaining, indicating that the expression of identified genes might explain some but not all of the SNP-breast cancer associations in these regions. For CASP8 and MRPL23‐AS1 regions, half of the GWAS hits remained genome-wide significant, and for the RP11‐554A11.9 region, 33 out of 36 GWAS SNPs remained (Figures [2](#page-14-0) and S1). These results suggest that the genetic association between breast cancer risk and those regions may not be mediated by transcriptional regulation of the genes on which we conditioned.

3.4 | Mutually adjusting for TWAS‐significant genes in the same region

As shown in Table [3,](#page-15-0) we identified six regions with more than one TWAS-significant gene: 2q33 (CASP8, ALS2CR12), 5q14 (ATG10, ATP6AP1L), 6q22 (RP11‐73O6.3, L3MBTL3), 15q24 (ULK3, MAN2C1, CTD‐2323K18.1), 17q21 (LRRC37A4p, CRHR1‐IT1, CRHR1, KANSL1‐AS1, LRRC37A, LRRC37A2), and 19q13 (ZNF404, ZNF155, RP11‐15A1.7). After mutually conditioning on the predicted expression of all significant genes in the same regions, ten genes remained nominally significant ($p < 0.05$). For some regions, only one gene remained, that is ATG10 for 5q14, L3MBTL3 for 6q22 and CRHR1-IT1 for 17q21 (Figures [3a](#page-16-0) and S2); while for other regions, multiple genes remained significant, including CASP8 and ALS2CR12 for 2q33,

FIGURE 2 Conditional and joint analysis (COJO) for genes near a strong breast cancer GWAS hit. (a) COJO results adjusting for predicted expression of ALS2CR12. After conditioning on ALS2CR12, almost all original significant GWAS signals (grey dots) disappear (blue dots). (b) COJO results adjusting for the predicted expression of CASP8. After conditioning on CASP8, some of the original GWAS significant signals (grey dots) remains (blue dots)

TABLE 3 Conditional and joint analysis of gene region with multiple TWAS significant genes

Abbreviations: COJO, conditional and joint; TWAS, transcriptome‐wide association studies.

^aBolded genes remain significant in conditional analyses. Analysis was performed using GWAS summary statistics of ER+ subtypes. Our primary goal in these analyses is to establish whether any of the marginally significant TWAS genes remains significant after conditioning for the most significant gene in the region; sincesince all of the regions with multiple significant genes contain 2–3 significant genes, using a conditional p value threshold of .05 is a reasonable threshold for identifying independent signals.

ULK3 and MAN2C1 for 15q24, and ZNF404, ZNF155, and RP11‐15A1.7 for 19q13 (Figures [3b](#page-16-0) and S2).

4 | DISCUSSION

We conducted a TWAS analysis using GTEx mammary tissue gene expression data and GWAS summary data from the largest meta‐analysis for breast cancer risk. We assessed associations between overall breast cancer risk and ER+ versus ER− disease. We found 30 genes significantly associated with overall breast cancer risk, 20 genes associated with the ER+ subtype, and six genes with the ER− subtype.

These results are consistent with previous reports from TWAS or similar gene‐based approaches, which used various algorithms to build gene expression models. For example, of the 30 genes that we found significantly related to overall breast cancer risk, 23 were also significant in Wu et al. ([2018](#page-25-10)) with very similar test statistics (correlation = 0.96 for the z scores between our and Wu's results), and six were significant in Ferreira et al. [\(2019\)](#page-24-10). One of the six genes we classified as significantly associated with ER− breast cancer was also found significantly associated with ER− breast cancer in Ferreira et al. ([2019](#page-24-10)). Among these studies, the approach taken by Wu et al. was the most similar to ours. Only seven of the 30 genes that we identified were not identified by Wu et al. [\(2018](#page-25-10)), probably due to different cis‐SNP selection criteria and different candidate genes selected for testing. We defined *cis*-SNPs using a 500 KB window around the gene boundary and included only candidate genes with a significant heritability, while Wu et al. used a 2 MB cis‐SNP window and included genes with a prediction performance of at least 0.01 without heritability filtering. For genes whose expression could not be predicted well, Wu et al. built models using only SNPs located in promoter or enhancer regions. Despite these methodological differences, the two TWAS results were highly concordant. However, we did not replicate any of the findings in Hoffman et al. [\(2017](#page-25-9)) and Gao et al. ([2017\)](#page-24-9), which may reflect the smaller sample size of the breast cancer GWAS used in their analyses (3,370 cases and 19,717 controls in Hoffman et al.; 10,597 overall breast cancer cases, 3,879 ER− cases and 11,358 controls in Gao et al.). Specifically, three of the previously reported genes were excluded by our stringent

FIGURE 3 COJO for regions with multiple TWAS associations. For each plot, the top panel shows all genes in the locus. After COJO analysis, the marginally associated genes are highlighted in blue, while those that remain jointly significant are highlighted in green (in this case, L3MBTL3, CASP8, and ALS2C12). The bottom panel shows a Manhattan plot of the GWAS signals before (gray) and after (blue) conditioning on the significant (green) genes. (a) COJO results for 6q22 (only one gene remains significant after COJO). (b) COJO results for 2q33 (an example of multiple genes remaining jointly remain significant after COJO). COJO, conditional and joint analysis; GWAS, genome‐ wide association studies; TWAS, transcriptome-wide association studies

QC procedure (DHODH, ANKLE1 from Hoffman et al. and TP53INP2 from Gao et al. were not heritable in our analysis) and one was not significant in our analysis (RCCD1 from Hoffman et al. $p = .0032$ for overall breast cancer). Both Hoffman et al. and Gao et al. used GWAS results based on a mixed population of European, African, and Asian ancestry (which shared a small set of European samples with our GWAS: $N < 5,700$ individuals from CGEMS and the BPC3, less than 2% of our GWAS sample). They also used different tissues to build their prediction weights: overall breast tissue (men and women combined, all ethnicities) and whole blood tissue (men and women combined, European ancestry).

Of the 30 genes associated with breast cancer risk in our study, 21 fell into known GWAS regions whereas nine were not close to any known GWAS hit and were, therefore, considered novel. Of these nine genes, five were identified and discussed in Wu et al. [\(2018\)](#page-25-10) or Ferreira et al. ([2019\)](#page-24-10). The four genes uniquely identified in the present study were GDI2, HSD17B1P1, MAEA, and ULK3, several of which have been reported to play a role in breast tumorigenesis or related biological processes. For example, the expression of GDI2 has been linked with breast cancer through its contribution to enhanced epidermal growth factor receptor endocytosis (EGFR; de Graauw et al., [2014](#page-24-18)). HSD17B1P1 is a pseudo‐gene related to HSD17, which participates in steroid hormone biosynthesis, metabolism, and signaling pathways potentially related to breast cancer risk (Jakubowska et al., [2010](#page-25-20)). These findings lend support to our results and suggested that further investigation into the roles of the novel genes identified for breast cancer is required.

We performed several conditional analyses not reported in previous TWAS. We examined the local GWAS signals conditioning on the expression of TWAS genes, to provide a measure of how well the expression level of identified TWAS genes explained the local GWAS signals. For many loci, these genes explained a large proportion of the local GWAS signals and were thus candidates for downstream experimental validation. We also identified candidate genes driving the statistical associations in regions with more than one TWAS gene (usually also regions with known GWAS risk loci) by jointly modeling multiple nominally significant genes. For example, previous studies have suggested that polymorphisms in CASP8 are associated with breast cancer risk (Cox et al., [2007](#page-24-19)), whereas a recent paper has shown that the most significant signal in this region is for the imputed intronic SNP rs1830298 in ALS2CR12 (telomeric to CASP8; Lin et al., [2015](#page-25-21)). Our results provide clarification on whether CASP8 or ALS2CR12 expression were more strongly associated with breast cancer risk, since both genes remained significantly associated with breast cancer risk after conditioning on the expression of the other (the conditional p value for ALS2CR12 was 3.70×10^{-6} , whereas the conditional *p* value for CASP8 was .05). Eleven of the 12 GWAS hits disappeared after adjusting for the expression of ALS2CR12, while half of the GWAS hits remained after adjusting for the expression of CASP8. Therefore, we believe that ALS2CR12 SNPs have a stronger effect and are associated with breast cancer through ALS2CR12 expression, while CASP8 remains an additional independent hit, consistent with the latest fine‐ mapping results (Lin et al., [2015](#page-25-21)).

Because the genes found to be associated with ER− disease were also associated with ER+ disease, and these, in turn, were associated with overall breast cancer risk, it is difficult to conclude whether the differences in gene sets are due to distinct mechanisms underlying breast cancer subtypes or due to a lack of statistical power because of the smaller disease subtype sample sizes. To address this question, we further incorporated a case‐only TWAS comparing ER+ versus ER− breast cancer. We identified two genes, STXBP4 and HIST2H2BA, associated with ER status, which were significantly associated only with ER+ but not ER− breast cancer. Previous studies supported the link between rs6504950 (a SNP in STXBP4) and overall breast cancer risk (Antoniou et al., [2010](#page-23-1); Warren Andersen et al., [2013\)](#page-25-22). It has also been hypothesized that the risk allele for the two top breast cancer candidate SNPs, rs2787486 and rs244353, affected gene expression of STXBP4 (Darabi et al., [2016\)](#page-24-20) and CD4 memory cells (Hnisz et al., [2013\)](#page-25-23). One potential explanation for the association between STXBP4 and breast cancer risk is that it encodes syntaxin binding protein 4, a scaffold protein. In addition, STXBP4 functions to stabilize and degrade TP63 isoform (a member of the TP53 tumor suppressor protein family), a biologically plausible candidate cancer susceptibility gene. Similarly, SNPs rs2580520 and rs11249433 upstream of HIST2H2BA have been identified as breast cancer susceptibility alleles

in a previous GWAS (Bogdanova, Helbig, & Dörk, [2013\)](#page-24-21). Our results suggest that functional and pathway analyses targeting these two genes are likely to shed new light on the differences in tumorigenesis and progression mechanisms between ER+ and ER− patients.

By building gene expression linear predictors in GTEx breast tissue, our analysis offers a tissue‐specific model of gene expression. The gene regulatory mechanisms in female breast tissue are arguably the most suitable for studying breast cancer. Moreover, by restricting our reference population to women of European ancestry, rather than mixing genders and ancestries, the resulting gene expression model was a better match to our breast cancer GWAS summary statistics. By using the largest GWAS meta‐analysis currently available, we greatly improved the power compared with previous work by Hoffman et al. [\(2017](#page-25-9)) and Gao et al. [\(2017\)](#page-24-9). Finally, by using case‐only GWAS summary statistics, we provided insights into genes associated with breast cancer subtype specific risk compared with Wu et al. ([2018\)](#page-25-10) and Ferreira et al. ([2019](#page-24-10)).

Similar to previous work by Wu et al. [\(2018](#page-25-10)) and Hoffman et al. ([2017\)](#page-25-9), our analyses focused on genetic tools trained using expression from breast tissue, chosen because of its direct relevance to breast carcinogenesis. However, given the relatively small sample size in the breast tissue eQTL panel, this choice limited both our power to detect genes with *cis*-heritable expression and the precision of estimated genetic predictors for heritable transcripts. The genetic regulation of expression is constant across tissues for many genes, suggesting that considering other tissues with larger eQTL sample sizes or combining eQTL evidence across tissues may improve power. In addition, other tissues may be relevant for breast cancer development. For example, considering that obesity and hormonal signaling have been linked to breast cancer risk (Bertolini, [2013\)](#page-24-22), gene expression in adipose tissue and brain tissue may have parallel involvement with breast cancer etiology. We are currently developing methods for cross‐tissue TWAS, using sCCA (sparse canonical correlation analysis) to build features that combine gene expression values across tissues that share similar genetic regulation mechanisms, while allowing tissues with different regulation patterns to contribute to different features (Feng, Pasaniuc, Major, & Kraft, [2018](#page-24-23)).

In conclusion, we have identified new breast cancer target genes both for functional experiments and as causal gene candidates in the significant TWAS gene regions. We have also identified associations between gene expression and breast cancer risk specific to disease subtypes, where two novel genes have been found specifically associated with ER+ breast cancer risk.

This analytic strategy warrants application in studies aimed at defining the genomic architecture of cancers other than breast cancer.

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CONFLICT OF INTERESTS

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DATA AVAILABILITY STATEMENT

The GWAS summary data that support the findings of overall, ER–, ER+ breast cancer results are openly available in the Breast Cancer Association Consortium (BCAC). The GWAS summary data that support the findings of case‐ only breast cancer results will also be made available at the BCAC website. [http://bcac.ccge.medschl.cam.ac.uk/](http://bcac.ccge.medschl.cam.ac.uk/bcacdata/oncoarray/gwas-icogs-and-oncoarray-summary-results/) [bcacdata/oncoarray/gwas](http://bcac.ccge.medschl.cam.ac.uk/bcacdata/oncoarray/gwas-icogs-and-oncoarray-summary-results/)‐icogs‐and‐oncoarray‐summary‐ [results/](http://bcac.ccge.medschl.cam.ac.uk/bcacdata/oncoarray/gwas-icogs-and-oncoarray-summary-results/) (Michailidou et al., [2017](#page-25-2), pp. 92–94).

REFERENCES

- Amos, C. I., Dennis, J., Wang, Z., Byun, J., Schumacher, F. R., Gayther, S. A., … Easton, D. F. (2016). The OncoArray Consortium: A network for understanding the genetic architecture of common cancers. Cancer Epidemiology, Biomarkers & Prevention, [https://doi.org/10.1158/1055](https://doi.org/10.1158/1055-9965.EPI-16-0106)‐9965. EPI‐16‐[0106](https://doi.org/10.1158/1055-9965.EPI-16-0106)
- Antoniou, A. C., Beesley, J., McGuffog, L., Sinilnikova, O. M., Healey, S., Neuhausen, S. L., … Cimba (2010). Common breast cancer susceptibility alleles and the risk of breast cancer for BRCA1 and BRCA2 mutation carriers: Implications for risk prediction. Cancer Research, 70(23), 9742–9754. [https://doi.org/](https://doi.org/10.1158/0008-5472.CAN-10-1907) [10.1158/0008](https://doi.org/10.1158/0008-5472.CAN-10-1907)‐5472.CAN‐10‐1907
- Antoniou, A. C., Goldgar, D. E., Andrieu, N., Chang‐Claude, J., Brohet, R., Rookus, M. A., & Easton, D. F. (2005). A weighted cohort approach for analysing factors modifying disease risks in carriers of high-risk susceptibility genes. Genetic Epidemiology, 29(1), 1–11. <https://doi.org/10.1002/gepi.20074>
- Atchley, D. P., Albarracin, C. T., Lopez, A., Valero, V., Amos, C. I., Gonzalez‐Angulo, A. M., … Arun, B. K. (2008). Clinical and pathologic characteristics of patients with BRCA‐positive and BRCA‐negative breast cancer. Journal of Clinical Oncology, 26(26), 4282–4288. <https://doi.org/10.1200/JCO.2008.16.6231>
- Barfield, R., Feng, H., Gusev, A., Wu, L., Zheng, W., Pasaniuc, B., & Kraft, P. (2018). Transcriptome-wide association studies accounting for colocalization using Egger regression. Genetic Epidemiology, 42(5), 418–433. <https://doi.org/10.1002/gepi.22131>
- Barnes, D. R., Lee, A., EMBRACE Investigators, kConFab Investigators, Easton, D. F., & Antoniou, A. C. (2012). Evaluation of association methods for analysing modifiers of disease risk in carriers of highrisk mutations. Genetic Epidemiology, 36(3), 274–291. [https://doi.](https://doi.org/10.1002/gepi.21620) [org/10.1002/gepi.21620](https://doi.org/10.1002/gepi.21620)
- Beggs, A. D., & Hodgson, S. V. (2009). Genomics and breast cancer: The different levels of inherited susceptibility. European Journal of Human Genetics, 17(7), 855–856. [https://doi.org/10.](https://doi.org/10.1038/ejhg.2008.235) [1038/ejhg.2008.235](https://doi.org/10.1038/ejhg.2008.235)
- Bertolini, F. (2013). Adipose tissue and breast cancer progression: A link between metabolism and cancer. Breast, 22(Suppl 2), S48–S49. <https://doi.org/10.1016/j.breast.2013.07.009>
- Blows, F. M., Driver, K. E., Schmidt, M. K., Broeks, A., van Leeuwen, F. E., Wesseling, J., … Huntsman, D. (2010). Subtyping of breast cancer by immunohistochemistry to investigate a relationship between subtype and short and long term survival: A collaborative analysis of data for 10,159 cases from 12 studies. PLOS Medicine, 7(5):e1000279. [https://doi.org/](https://doi.org/10.1371/journal.pmed.1000279) [10.1371/journal.pmed.1000279](https://doi.org/10.1371/journal.pmed.1000279)
- Bogdanova, N., Helbig, S., & Dörk, T. (2013). Hereditary breast cancer: Ever more pieces to the polygenic puzzle. Hereditary Cancer in Clinical Practice, 11(1), 12. [https://doi.org/10.1186/](https://doi.org/10.1186/1897-4287-11-12) 1897‐[4287](https://doi.org/10.1186/1897-4287-11-12)‐11‐12
- Bray, F., Ferlay, J., Soerjomataram, I., Siegel, R. L., Torre, L. A., & Jemal, A. (2018). Global cancer statistics 2018: GLOBOCAN estimates of incidence and mortality worldwide for 36 cancers in 185 countries. CA: A Cancer Journal for Clinicians, 68(6), 394–424. <https://doi.org/10.3322/caac.21492>
- Cai, Q., Zhang, B., Sung, H., Low, S.‐K., Kweon, S.‐S., Lu, W., … Zheng, W. (2014). Genome‐wide association analysis in East Asians identifies breast cancer susceptibility loci at 1q32.1, 5q14.3 and 15q26.1. Nature Genetics, 46(8), 886–890. [https://](https://doi.org/10.1038/ng.3041) doi.org/10.1038/ng.3041
- Cox, A., Dunning, A. M., Garcia‐Closas, M., Balasubramanian, S., Reed, M. W. R., Pooley, K. A., … Johnson, N., Breast Cancer Association Consortium. (2007). A common coding variant in CASP8 is associated with breast cancer risk. Nature Genetics, 39(3), 352–358. <https://doi.org/10.1038/ng1981>
- Darabi, H., Beesley, J., Droit, A., Kar, S., Nord, S., Moradi Marjaneh, M., … Dunning, A. M. (2016). Fine scale mapping of the 17q22 breast cancer locus using dense SNPs, genotyped within the Collaborative Oncological Gene‐ Environment Study (COGs). Scientific Reports, 6, 32512. <https://doi.org/10.1038/srep32512>
- Delaneau, O., Marchini, J., & Zagury, J.‐F. (2011). A linear complexity phasing method for thousands of genomes. Nature Methods, 9(2), 179–181. [https://doi.org/10.1038/](https://doi.org/10.1038/nmeth.1785) [nmeth.1785](https://doi.org/10.1038/nmeth.1785)
- Dunning, A. M., Michailidou, K., Kuchenbaecker, K. B., Thompson, D., French, J. D., Beesley, J., … Edwards, S. L. (2016). Breast cancer risk variants at 6q25 display different phenotype associations and regulate ESR1, RMND1 and CCDC170. Nature Genetics, 48(4), 374–386. [https://doi.org/10.](https://doi.org/10.1038/ng.3521) [1038/ng.3521](https://doi.org/10.1038/ng.3521)
- Easton, D. F., Pooley, K. A., Dunning, A. M., Pharoah, P. D. P., Thompson, D., Ballinger, D. G., … Ponder, B. A. J. (2007). Genome‐wide association study identifies novel breast cancer susceptibility loci. Nature, 447(7148), 1087–1093. [https://doi.](https://doi.org/10.1038/nature05887) [org/10.1038/nature05887](https://doi.org/10.1038/nature05887)
- ENCODE Project Consortium. (2012). An integrated encyclopedia of DNA elements in the human genome. Nature, 489(7414), 57–74. <https://doi.org/10.1038/nature11247>
- Feng, H., Pasaniuc, B., Major, M., & Kraft, P. (2018). Sparse canonical correlation analysis (sCCA) significantly improves power of cross‐tissue transcriptome‐wide association studies (TWAS). Genetic Epidemiology, 42, 698.
- Ferreira, M. A., Gamazon, E. R., Al‐Ejeh, F., Aittomäki, K., Andrulis, I. L., Anton‐Culver, H., … Chenevix‐Trench, G. (2019). Genome‐wide association and transcriptome studies identify target genes and risk loci for breast cancer. Nature Communications, 10(1), 1741. [https://doi.org/10.1038/s41467](https://doi.org/10.1038/s41467-018-08053-5)‐ 018‐[08053](https://doi.org/10.1038/s41467-018-08053-5)‐5
- Gamazon, E. R., Wheeler, H. E., Shah, K. P., Mozaffari, S. V., Aquino‐Michaels, K., Carroll, R. J., … Im, H. K. (2015). A gene‐ based association method for mapping traits using reference transcriptome data. Nature Genetics, 47(9), 1091–1098. [https://](https://doi.org/10.1038/ng.3367) doi.org/10.1038/ng.3367
- Gao, G., Pierce, B. L., Olopade, O. I., Im, H. K., & Huo, D. (2017). Trans‐ethnic predicted expression genome‐wide association analysis identifies a gene for estrogen receptor‐negative breast cancer. PLOS Genetics, 13(9):e1006727. [https://doi.org/10.1371/](https://doi.org/10.1371/journal.pgen.1006727) [journal.pgen.1006727](https://doi.org/10.1371/journal.pgen.1006727)
- Garcia‐Closas, M., Couch, F. J., Lindstrom, S., Michailidou, K., Schmidt, M. K., Brook, M. N., … Kraft, P. (2013). Genome‐wide association studies identify four ER negative‐specific breast cancer risk loci. Nature Genetics, 45(4), 392–398, 398e1‐2. <https://doi.org/10.1038/ng.2561>
- de Graauw, M., Cao, L., Winkel, L., van Miltenburg, M. H. a M., le Dévédec, S. E., Klop, M., … van de Water, B. (2014). Annexin A2 depletion delays EGFR endocytic trafficking via cofilin activation and enhances EGFR signaling and metastasis formation. Oncogene, 33(20), 2610–2619. [https://doi.org/10.](https://doi.org/10.1038/onc.2013.219) [1038/onc.2013.219](https://doi.org/10.1038/onc.2013.219)
- GTEx Consortium. (2015). Human genomics. The Genotype‐Tissue Expression (GTEx) pilot analysis: Multitissue gene regulation in humans. Science, 348(6235), 648–660. [https://doi.org/10.](https://doi.org/10.1126/science.1262110) [1126/science.1262110](https://doi.org/10.1126/science.1262110)
- Gusev, A., Ko, A., Shi, H., Bhatia, G., Chung, W., Penninx, B. W. J. H., … Pasaniuc, B. (2016). Integrative approaches for large‐scale transcriptome‐wide association studies. Nature Genetics, 48(3), 245–252. [https://doi.org/10.](https://doi.org/10.1038/ng.3506) [1038/ng.3506](https://doi.org/10.1038/ng.3506)

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- Hnisz, D., Abraham, B. J., Lee, T. I., Lau, A., Saint‐André, V., Sigova, A. A., … Young, R. A. (2013). Super‐enhancers in the control of cell identity and disease. Cell, 155(4), 934–947. <https://doi.org/10.1016/j.cell.2013.09.053>
- Hoffman, J. D., Graff, R. E., Emami, N. C., Tai, C. G., Passarelli, M. N., Hu, D., … Witte, J. S. (2017). Cis‐eQTL‐based trans‐ethnic meta‐analysis reveals novel genes associated with breast cancer risk. PLOS Genetics, 13(3):e1006690. [https://doi.](https://doi.org/10.1371/journal.pgen.1006690) [org/10.1371/journal.pgen.1006690](https://doi.org/10.1371/journal.pgen.1006690)
- Howie, B. N., Donnelly, P., & Marchini, J. (2009). A flexible and accurate genotype imputation method for the next generation of genome‐wide association studies. PLOS Genetics, 5(6): e1000529. <https://doi.org/10.1371/journal.pgen.1000529>
- Jakubowska, A., Gronwald, J., Menkiszak, J., Górski, B., Huzarski, T., Byrski, T., … Hamann, U. (2010). BRCA1‐ associated breast and ovarian cancer risks in Poland: No association with commonly studied polymorphisms. Breast Cancer Research and Treatment, 119(1), 201–211. [https://doi.](https://doi.org/10.1007/s10549-009-0390-5) [org/10.1007/s10549](https://doi.org/10.1007/s10549-009-0390-5)‐009‐0390‐5
- Lin, W.‐Y., Camp, N. J., Ghoussaini, M., Beesley, J., Michailidou, K., Hopper, J. L., … Cox, A. (2015). Identification and characterization of novel associations in the CASP8/ ALS2CR12 region on chromosome 2 with breast cancer risk. Human Molecular Genetics, 24(1), 285–298. [https://doi.org/10.](https://doi.org/10.1093/hmg/ddu431) [1093/hmg/ddu431](https://doi.org/10.1093/hmg/ddu431)
- Mancuso, N., Freund, M. K., Johnson, R., Shi, H., Kichaev, G., Gusev, A., & Pasaniuc, B. (2019). Probabilistic fine‐mapping of transcriptome-wide association studies. Nature Genetics, 51(4), 675–682. [https://doi.org/10.1038/s41588](https://doi.org/10.1038/s41588-019-0367-1)‐019‐0367‐1
- Mancuso, N., Shi, H., Goddard, P., Kichaev, G., Gusev, A., & Pasaniuc, B. (2017). Integrating gene expression with summary association statistics to identify genes associated with 30 complex traits. American Journal of Human Genetics, 100(3), 473–487. <https://doi.org/10.1016/j.ajhg.2017.01.031>
- Mavaddat, N., Antoniou, A. C., Easton, D. F., & Garcia‐Closas, M. (2010). Genetic susceptibility to breast cancer. Molecular oncology, 4(3), 174–191. [https://doi.org/10.1016/j.molonc.2010.](https://doi.org/10.1016/j.molonc.2010.04.011) [04.011](https://doi.org/10.1016/j.molonc.2010.04.011)
- Mavaddat, N., Pharoah, P. D. P., Michailidou, K., Tyrer, J., Brook, M. N., Bolla, M. K., … Garcia‐Closas, M. (2015). Prediction of breast cancer risk based on profiling with common genetic variants. Journal of the National Cancer Institute, 107(5), <https://doi.org/10.1093/jnci/djv036>
- McCarthy, S., Das, S., Kretzschmar, W., Delaneau, O., Wood, A. R., Teumer, A., … Mahajan, A., Haplotype Reference Consortium. (2016). A reference panel of 64,976 haplotypes for genotype imputation. Nature Genetics, 48(10), 1279–1283. [https://doi.org/](https://doi.org/10.1038/ng.3643) [10.1038/ng.3643](https://doi.org/10.1038/ng.3643)
- Michailidou, K., Beesley, J., Lindstrom, S., Canisius, S., Dennis, J., Lush, M. J., … Easton, D. F. (2015). Genome‐wide association analysis of more than 120,000 individuals identifies 15 new susceptibility loci for breast cancer. Nature Genetics, 47(4), 373–380. <https://doi.org/10.1038/ng.3242>
- Michailidou, K., Lindström, S., Dennis, J., Beesley, J., Hui, S., Kar, S., … Easton, D. F. (2017). Association analysis identifies 65 new breast cancer risk loci. Nature, 551(7678), 92–94. <https://doi.org/10.1038/nature24284>
- Milne, R. L., Kuchenbaecker, K. B., Michailidou, K., Beesley, J., Kar, S., Lindström, S., … Simard, J. (2017). Identification of ten

variants associated with risk of estrogen‐receptor‐negative breast cancer. Nature Genetics, 49(12), 1767–1778. [https://doi.](https://doi.org/10.1038/ng.3785) [org/10.1038/ng.3785](https://doi.org/10.1038/ng.3785)

- Nicolae, D. L., Gamazon, E., Zhang, W., Duan, S., Dolan, M. E., & Cox, N. J. (2010). Trait‐associated SNPs are more likely to be eQTLs: Annotation to enhance discovery from GWAS. PLOS Genetics, 6(4):e1000888. [https://doi.org/10.1371/journal.pgen.](https://doi.org/10.1371/journal.pgen.1000888) [1000888](https://doi.org/10.1371/journal.pgen.1000888)
- Pasaniuc, B., Zaitlen, N., Shi, H., Bhatia, G., Gusev, A., Pickrell, J., … Price, A. L. (2014). Fast and accurate imputation of summary statistics enhances evidence of functional enrichment. Bioinformatics, 30(20), 2906–2914. [https://doi.org/10.1093/](https://doi.org/10.1093/bioinformatics/btu416) [bioinformatics/btu416](https://doi.org/10.1093/bioinformatics/btu416)
- Seal, S., Thompson, D., Renwick, A., Elliott, A., Kelly, P., Barfoot, R., … Rahman, N. (2006). Truncating mutations in the Fanconi anemia J gene BRIP1 are low‐penetrance breast cancer susceptibility alleles. Nature Genetics, 38(11), 1239-1241. <https://doi.org/10.1038/ng1902>
- Turnbull, C., Ahmed, S., Morrison, J., Pernet, D., Renwick, A., Maranian, M., ... Easton, D. F. (2010). Genome-wide association study identifies five new breast cancer susceptibility loci. Nature Genetics, 42(6), 504–507. <https://doi.org/10.1038/ng.586>
- Wainberg, M., Sinnott‐Armstrong, N., Mancuso, N., Barbeira, A. N., Knowles, D. A., Golan, D., … Kundaje, A. (2019). Opportunities and challenges for transcriptome‐wide association studies. Nature Genetics, 51(4), 592–599. [https://doi.org/10.1038/](https://doi.org/10.1038/s41588-019-0385-z) [s41588](https://doi.org/10.1038/s41588-019-0385-z)‐019‐0385‐z
- Warren Andersen, S., Trentham‐Dietz, A., Gangnon, R. E., Hampton, J. M., Figueroa, J. D., Skinner, H. G., … Newcomb, P. A. (2013). The associations between a polygenic score, reproductive and menstrual risk factors and breast cancer risk. Breast Cancer Research and Treatment, 140(2), 427–434. [https://doi.org/10.1007/](https://doi.org/10.1007/s10549-013-2646-3) [s10549](https://doi.org/10.1007/s10549-013-2646-3)‐013‐2646‐3
- Willer, C. J., Li, Y., & Abecasis, G. R. (2010). METAL: Fast and efficient meta‐analysis of genomewide association scans. Bioinformatics, 26(17), 2190–2191. [https://doi.org/10.1093/](https://doi.org/10.1093/bioinformatics/btq340) [bioinformatics/btq340](https://doi.org/10.1093/bioinformatics/btq340)
- Wu, L., Shi, W., Long, J., Guo, X., Michailidou, K., Beesley, J., … Zheng, W. (2018). A transcriptome‐wide association study of 229,000 women identifies new candidate susceptibility genes for breast cancer. Nature Genetics, 50(7), 968–978. [https://doi.](https://doi.org/10.1038/s41588-018-0132-x) [org/10.1038/s41588](https://doi.org/10.1038/s41588-018-0132-x)‐018‐0132‐x
- Yang, J., Benyamin, B., McEvoy, B. P., Gordon, S., Henders, A. K., Nyholt, D. R., … Visscher, P. M. (2010). Common SNPs explain a large proportion of the heritability for human height. Nature Genetics, 42(7), 565–569. [https://](https://doi.org/10.1038/ng.608) doi.org/10.1038/ng.608
- Yang, J., Lee, S. H., Goddard, M. E., & Visscher, P. M. (2011). GCTA: A tool for genome‐wide complex trait analysis. American Journal of Human Genetics, 88(1), 76-82. [https://](https://doi.org/10.1016/j.ajhg.2010.11.011) doi.org/10.1016/j.ajhg.2010.11.011
- Yang, X. R., Chang‐Claude, J., Goode, E. L., Couch, F. J., Nevanlinna, H., Milne, R. L., … Garcia‐Closas, M. (2011). Associations of breast cancer risk factors with tumor subtypes: A pooled analysis from the Breast Cancer Association Consortium studies. Journal of the National Cancer Institute, 103(3), 250–263. [https://doi.org/10.1093/](https://doi.org/10.1093/jnci/djq526) [jnci/djq526](https://doi.org/10.1093/jnci/djq526)

Yovel, Y., Franz, M. O., Stilz, P., & Schnitzler, H.‐U. (2008). Plant classification from bat-like echolocation signals. PLOS Computational Biology, 4(3):e1000032. [https://doi.org/10.1371/](https://doi.org/10.1371/journal.pcbi.1000032) [journal.pcbi.1000032](https://doi.org/10.1371/journal.pcbi.1000032)

SUPPORTING INFORMATION

Additional supporting information may be found online in the Supporting Information section.

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