

1 NeuroChip, an updated version of the NeuroX genotyping platform to
2 rapidly screen for variants associated with neurological diseases

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59

60 Abstract

61 Genetics has proven to be a powerful approach in neurodegenerative diseases research,
62 resulting in the identification of numerous causal and risk variants. Previously, we introduced
63 the NeuroX Illumina genotyping array, a fast and efficient genotyping platform designed for the
64 investigation of genetic variation in neurodegenerative diseases. Here, we present its updated
65 version, named NeuroChip. The NeuroChip is a low cost, custom-designed array containing a
66 tagging variant backbone of about 306,670 variants complemented with a manually curated
67 custom content comprised of 179,467 variants implicated in diverse neurological diseases,
68 including Alzheimer's disease, Parkinson's disease, Lewy body dementia, amyotrophic lateral
69 sclerosis, frontotemporal dementia, progressive supranuclear palsy, corticobasal degeneration
70 and multiple system atrophy. The tagging backbone was chosen because of the low cost and
71 good genome-wide resolution; the custom content can be combined with other backbones, like
72 population or drug development arrays. Using the NeuroChip, we can accurately identify rare
73 variants and impute over 5.3 million common SNPs from the latest release of the Haplotype
74 Reference Consortium. In summary, we describe the design and usage of the NeuroChip array,
75 and show its capability for detecting rare pathogenic variants in numerous neurodegenerative
76 diseases. The NeuroChip has a more comprehensive and improved content, which makes it a
77 reliable, high-throughput, cost-effective screening tool for genetic research and molecular
78 diagnostics in neurodegenerative diseases.

79

80 1. Introduction

81 Neurodegenerative diseases are a major burden to the aging world population and currently
82 these diseases are incurable and irreversible. Common and rare genetic alterations in many
83 genes have been identified as disease-causing or contributing to the development of
84 neurodegeneration (Naj et al., 2017, Singleton and Hardy, 2016). To date, there are four main
85 uses of genetics: 1) to confirm a clinical diagnosis by identifying a causal mutation, 2) to identify
86 risk variants and disease modifiers that influence risk for disease, 3) to increase knowledge of
87 the molecular pathobiology of disease in the hopes of identifying therapeutic targets, and 4) to
88 improve patient selection for pathway-specific clinical trial design. A reliable, high-throughput
89 and cost-effective platform that can rapidly conduct these functions could therefore be
90 immensely valuable to the field.

91 Previously, we presented the NeuroX array, which was a collaborative effort with the objective
92 of designing a genotyping platform that would allow rapid genetic characterization of samples
93 in the context of genetic mutations and risk factors associated with common
94 neurodegenerative diseases (Nalls et al., 2015). This was an exonic array (or exome chip) based
95 on the Infinium HumanExome Beadchip v1.1 containing 242,901 exome-focused variants as
96 well as 24,706 custom variants focusing on neurological diseases. The NeuroX array has already
97 been successfully used in dozens of studies (Barber et al., 2017, Carrasquillo et al., 2016, Ghani
98 et al., 2015, Nalls et al., 2016, Rosenthal et al., 2016). However, due to the backbone's focus on
99 rare exonic variation, common non-exonic variants were largely missed, resulting in a modest
100 genome-wide resolution and only partial capture of the known low frequency exonic variation.

101 Additionally, the number of genotype-phenotype associations and pathogenic variants keeps
102 expanding, so there was a continued need for updating this useful platform.
103 Here, we report on an updated version of NeuroX, named NeuroChip. The NeuroChip backbone
104 is based on a genome-wide genotyping array (Infinium HumanCore-24 v1.0) containing 306,670
105 tagging variants and a custom content that has been updated and extended with
106 neurodegenerative disease-related custom content consisting of 179,467 variants. This
107 backbone was chosen because of the low cost and good genome-wide resolution. This
108 backbone is flexible and other arrays can be used with this custom content, such as population
109 or drug development arrays (Infinium Multi-Ethnic, Infinium DrugDev). The NeuroChip allows to
110 accurately identify rare neurodegenerative candidate variants and impute over 5.3 million
111 common variants. Its approximate cost of ~\$40 per sample is a fraction of the price of next-
112 generation whole exome or whole genome sequencing, and therefore provides a valuable,
113 high-throughput screening tool for loci and variants implicated in neurodegenerative diseases.
114 Further, this array can be used as a tool to prioritize samples for more expensive genome
115 sequencing approaches.

116

117 2. Methods

118 2.1 NeuroChip array design

119 The backbone of the array, the Infinium HumanCore-24 v1.0, contains 306,670 highly
120 informative tagging SNPs which can be used for high-throughput and high-quality imputation of
121 genome-wide variants across diverse populations (Illumina). In addition, the chip contains
122 179,467 custom disease-associated variants (Table 1) covering neurodegenerative diseases

123 including: Alzheimer's disease (AD), Parkinson's disease (PD), Lewy body dementia (LBD),
124 frontotemporal dementia (FTD), amyotrophic lateral sclerosis (ALS), progressive supranuclear
125 palsy (PSP), corticobasal degeneration (CBD) and multiple system atrophy (MSA). The custom-
126 content has been curated by members of the International Parkinson's Disease Genomics
127 Consortium (IPDGC) and the Comprehensive Unbiased Risk factor Assessment for Genetics and
128 Environment in Parkinson disease (COURAGE-PD) consortium to include common variants and
129 rare mutations implicated in neurological diseases as reported in the Human Gene Mutation
130 Database (HGMD Professional 2016.4, QIAGEN), the NHGRI GWAS Catalog
131 (www.ebi.ac.uk/gwas/), the Online Mendelian Inheritance in Man (OMIM) database
132 (www.ncbi.nlm.nih.gov/omim/), the Parkinson's Disease Mutation Database (www.molgen.vib-ua.be/PDMutDB), the Alzheimer's Disease and Frontotemporal Dementia Database
133 (www.molgen.ua.ac.be/admutations/), and based on literature review as well as own data;
134 particularly in the latter case, collaborators submitted variants that were identified in multiple
135 ongoing (or completed) unpublished projects, including variants from genome-wide association
136 (GWA), whole exome, whole genome, targeted sequencing studies and systems biology studies.
137 See Supplementary Table 1 for the complete content of the NeuroChip array.

139

140 2.2 NeuroChip array genotyping

141 We genotyped a cohort of 273 controls as per the manufacturer's instructions (Illumina) to
142 generate pilot NeuroChip data. These samples have been collected by the North American Brain
143 Expression Consortium (NABEC) and described elsewhere (Hernandez et al., 2012). In total, 183
144 males and 90 females were included. All samples were obtained from North American brain

145 banks and subjects reported European ancestry and had no reported neurological disease. To
146 assess the reproducibility of the NeuroChip, we genotyped 15 samples twice in separate
147 experiments.
148 Raw data files were imported into GenomeStudio (version 2.0, Illumina). For initial quality
149 control, we confirmed accurate, high quality genotyping using a call rate threshold of > 95%.
150 We reclustered the samples using a GenCall threshold of 0.15 and recalled all variants. The
151 genotyping cluster file based on ~3,500 individuals of ongoing projects is available in the
152 Supplementary Materials (Supplementary File 1). The mean call rate post-reclustering was
153 0.992 (range: 0.954-0.995). The data were exported from GenomeStudio using the Illumina-to-
154 PLINK module 2.1.4 and imported into PLINK (version 1.90) (Chang et al., 2015). Next, we
155 checked individuals for discrepancies between reported sex and genotypic sex, cryptic
156 relatedness (PIHAT <0.05), and heterogeneity contamination, and found that no samples failed
157 this quality control step.

158

159 2.3 NeuroChip content annotation

160 Annotation of the NeuroChip content was performed using ANNOVAR (Wang et al., 2010). For
161 each variant, a gene-based annotation, *in silico* impact scores, and frequencies from public
162 databases were obtained. To predict the impact scores, the following algorithms were used:
163 SIFT (Kumar et al., 2009), Polyphen-2 (Adzhubei et al., 2010), and CADD (Kircher et al., 2014).
164 Population frequencies were obtained from the Exome Aggregation Consortium (version 0.3.1)
165 (<http://exac.broadinstitute.org/>) containing 60,706 individuals. Additionally, all variants were
166 investigated for their presence in the Human Gene Mutation Database (HGMD, accessed 20

167 December 2016). Variants associated with a common neurodegenerative syndrome (AD, ALS,
168 FTD and PD) were manually curated and are summarized in Supplementary Table 2.

169

170 2.4 NeuroChip content imputation

171 After confirming high-quality genotyping (call rate >95%) and European ancestry in all

172 individuals (based on 1000Genomes clustering) (Genomes Project et al., 2015), we performed

173 imputation using the Michigan imputation server, according to established guidelines

174 (<https://imputationserver.sph.umich.edu>) (Das et al., 2016). In brief, genotypes were prepared

175 for imputation using provided scripts (HRC-1000G-check-bim.pl), which compares variant ID,

176 strand, and allele frequencies to the haplotype reference panel (HRC version r1.1, April 2016)

177 (McCarthy et al., 2016). A total of 332,015 autosomal SNPs were submitted to the Imputation

178 Server using ShapeIT (v2.r790).

179

180 2.5 *APOE* allele genotyping

181 To determine the accuracy of *APOE* allele predictions, we performed Taqman genotyping of

182 two nonsynonymous *APOE* SNPs (rs7412 and rs429358) on an Applied Biosystems ViiA 7 Real-

183 Time PCR System using an established protocol (Federoff et al., 2012). 272 out of 273 control

184 samples had sufficient DNA for genotyping. Allelic discrimination was conducted using

185 QuantStudio software (version 1.3, Thermo Fisher Scientific, Carlsbad, CA, USA). Taqman

186 genotype results were then compared to the corresponding results for the same SNPs

187 generated using the NeuroChip. Given the importance of *APOE*, NeuroChip was designed so

188 that rs7412 is genotyped by four separate probes (three of which performed well: rs7412, seq-

189 rs7412-B1, seq-rs7412-B3). Similarly, rs429358 was genotyped by five separate bead probes
190 (two of which performed well: seq-rs429358-T2, seq-rs429358-T3). This redundancy ensures
191 accurate *APOE* genotyping by the NeuroChip platform.

192

193 3. Results

194 3.1 NeuroChip content overview

195 In total, the NeuroChip array contains 473,442 autosomal variants, 11,840 sex chromosomal
196 variants, and 160 mitochondrial variants. Additionally, 16,274 NeuroChip variants detect small
197 insertions or deletions (Table 1). The overlap between NeuroX and NeuroChip is small (n=
198 19,289 variants) due to the difference in the design of the backbone; the NeuroX array is
199 focused on exonic content, whereas the NeuroChip is focused on genome wide tagging content.

200

201 3.2 NeuroChip pathogenic variant content

202 In total, the NeuroChip harbors 8,086 disease-associated variants that are included in HGMD, a
203 professionally curated database of published genetic variants that have been linked to inherited
204 human diseases (neurological and non-neurological). The NeuroChip HGMD content includes
205 1,233 variants (1,202 SNPs and 31 indels) linked to common neurodegenerative syndromes (see
206 Supplementary Figure 1 for a comparison between NeuroX and NeuroChip). In this content,
207 after manually curation, 601 variants are associated with ALS or FTD, 348 with PD, and 284 with
208 AD. Figure 1 shows the number of disease associated variants per gene covered in common
209 neurodegenerative syndromes based on the HGMD database. Detailed, manually curated and
210 annotated variant lists for the abovementioned neurodegenerative disease categories are

211 documented in Supplementary Table 2. These annotated lists can be used as filters to quickly
212 screen for known mutations and risk variants.

213

214 3.3 NeuroChip genotyping results

215 *Genotyping reproducibility*

216 Of the 15 technical replicates, all samples yielded high quality, reproducible genotyping results.

217 The mean concordance rate per technical replicate was 0.9996 (range=0.9991-0.9999); on

218 average, 190 variants (range=27-435) differed per technical replicate (0.04% of the total

219 included variants on the array). Across the 15 technical replicates, 1,978 unique variants were

220 discordant, of which 749 (37.9%) were from the backbone and 1,229 (62.1%) were from the

221 custom content (Supplementary Table 3).

222

223 *Imputation*

224 Imputation of autosomal variants was performed on a series of 273 European descent

225 individuals using the haplotype reference panel (McCarthy et al., 2016) containing 39,235,157

226 variants, all with an estimated minor allele count of ≥ 5 in 32,488 individuals. Initial pre-

227 imputation filtering of the NeuroChip data (including removing duplicates and non-overlapping

228 variants, switch strands, and updating position) resulted in 332,015 variants. After imputation,

229 11,879,345 variants were obtained with an imputation R^2 of > 0.30 . Filtering based on MAF $>$

230 0.05, Hardy-Weinberg Equilibrium p -value of $> 1e-6$ resulted in 5,316,028 variants. In this

231 imputed dataset, we successfully identified 22 of 26 PD risk alleles and 19 of the 21 AD GWA

232 SNPs (Lambert et al., 2013, Nalls et al., 2014).

233

234 *Genotype accuracy*

235 GenTrain scores were calculated for all NeuroChip variants using GenomeStudio (version 2,
236 Illumina). The GenTrain score is a statistical score based on the shapes of the different allelic
237 clusters and their relative distance to each other (Illumina). Typically, GenTrain scores > 0.7 are
238 considered high quality genotypes. Previously, GenTrain scores of the NeuroX showed that
239 genotyping quality in the custom content was lower compared to the backbone (Nalls et al.,
240 2015). However, preliminary NeuroChip data from several ongoing projects (based on ~3,500
241 individuals) reveals that the backbone and the custom content have a high comparable average
242 score (0.819 and 0.820, respectively), indicating high genotyping accuracy (Supplementary
243 Figures 2 & 3).

244

245 *Validation of APOE genotyping*

246 *APOE* alleles are important genetic risk factors for both AD and LBD, but genotyping of this
247 region is complicated by high GC content (Singleton et al., 2002, Strittmatter and Roses, 1996).
248 For this reason, we chose to validate the accuracy of *APOE* allele genotyping by comparing
249 Taqman results with genotype predictions from the NeuroChip (Supplementary Table 4).
250 Taqman genotyping for rs7412 and rs429358 was successful in all 272 samples. NeuroChip
251 genotyping for both SNPs was successful in 265 out of these 272 controls. Five samples were
252 discordant for *APOE* allele genotyping between Taqman and NeuroChip, likely due to relatively
253 low quality genotype calls in either the Taqman assay or the NeuroChip, representing 1.9% of
254 our test cohort (n = 265 samples). Overall, the performance of the NeuroChip for *APOE*

255 genotyping was significantly better than the original NeuroX platform, which was unable to
256 reliably detect rs7412 and rs429358 genotypes (Ghani et al., 2015, Nalls et al., 2015).

257

258 4. Discussion

259 The main goal was to develop a genotyping array that allows a rapid, high-throughput
260 identification of common and rare single nucleotide variants in the human genome. Affordable
261 screening of large cohorts for disease-associated variants allows for testing of polygenic
262 inheritance that could explain the diversity of clinical and pathological characteristics of
263 neurodegenerative diseases. The NeuroChip is estimated to cost ~ \$40/sample, which is
264 currently less than ~ 10% and ~ 5% of the cost of whole exome sequencing and whole genome
265 sequencing, respectively.

266 We have designed, implemented, and validated the NeuroChip array platform for high
267 throughput genotyping. However, it is important to recognize the limitations of this approach.
268 Like all genotyping arrays, NeuroChip does not detect novel sequence changes. It is also not
269 possible to genotype variants in complex genomic regions (e.g. due to pseudogenes) or to
270 identify repeat expansions due to the difficulty in designing reliable probes. Nevertheless, every
271 effort was made to improve genotyping calling in NeuroChip. For example, it was recognized
272 that the *APOE* locus performed poorly on the original NeuroX platform (Ghani et al., 2015).
273 Given the importance of this genomic region in neurodegeneration, the revised NeuroChip
274 probe design included multiple probes for SNPs in this region. This led to reliable *APOE* allele
275 calling with a concordance rate of 98.1% between NeuroChip and Taqman.

276 In conclusion, we describe the design and implementation of the NeuroChip array, which has a
277 more comprehensive and improved content compared to NeuroX. This versatile genotyping
278 platform provides the community with a novel tool that can be used in both a clinical and
279 research setting. In a clinical setting, it is possible to rapidly screen patients for a large number
280 of known pathogenic variants and in a research setting cost-effective and high throughput
281 detection of both common and rare variants gives the opportunity to perform several analyses
282 including GWAS, burden tests and genetic risk scores calculations.

283

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