

1 **From FAANG to Fork: Application of Highly Annotated Genomes to Improve Farmed Animal**
2 **Production**

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26

27 **Abstract**

28

29 Here we describe a set of research priorities to meet present and future challenges posed to
30 farmed animal production that build on progress, successes and resources from the Functional
31 Annotation of ANimal Genomes (FAANG) project.

32

33 **Keywords: FAANG, genomics, farmed animal, genotype-to-phenotype**

34

35 **Introduction**

36

37 The Food and Agriculture Organisation of the United Nations (FAO) reports that by
38 the year 2050 the global human population is likely to reach 9.1 billion, rising to 11.2 billion
39 by the year 2100. This population growth poses several challenges to the global food system,
40 which will need to produce more healthy food using fewer natural resources, reducing the
41 environmental impact, conserving biodiversity and flexibly adjusting to changing societal
42 expectations. Meeting this demand requires environmentally sustainable improvements to
43 farmed animal health and welfare, and of efficiency and diversification (e.g. to include a
44 broader range of locally adapted species) [1]. The changes in breeding strategies and
45 management practises required to meet these goals will need to build on an improved ability
46 to accurately use genotype to predict phenotype in the world's farmed animal species, both
47 terrestrial and aquatic (Box 1).

48

49 **Box One: Addressing the Challenges of Global Food Production in the 21st Century**

Addressing the Challenges of Global Food Production in the 21st Century

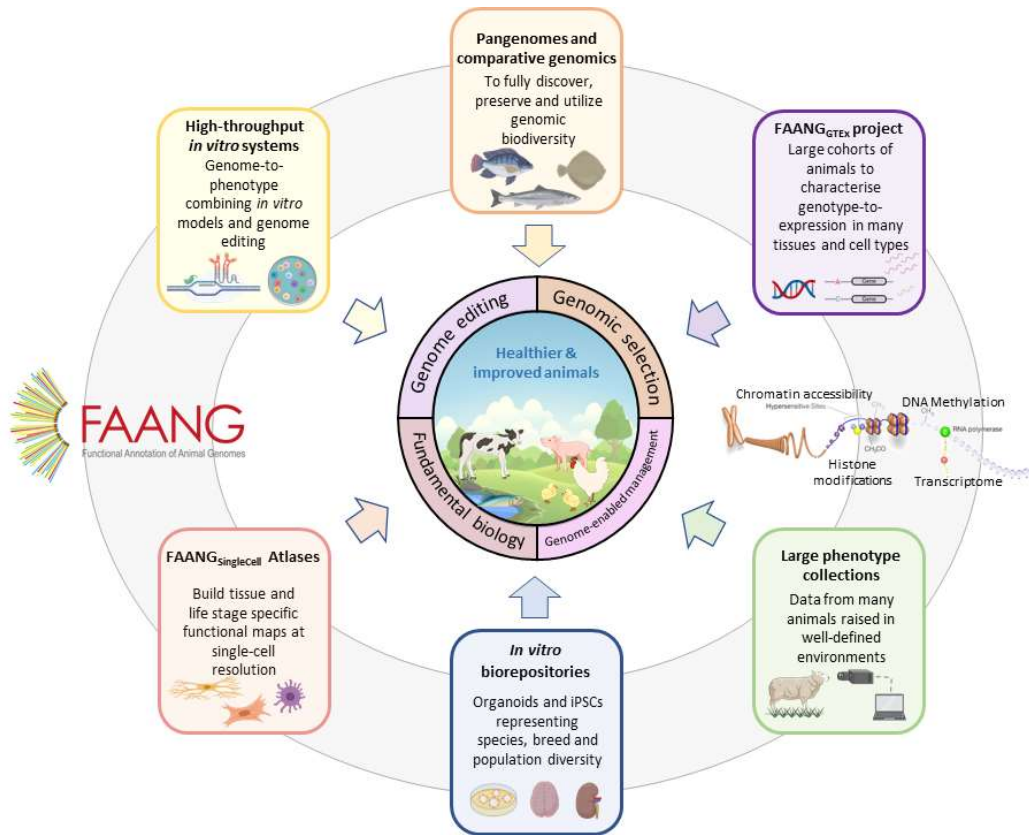
- Sustainable improvements in efficient production of farmed animals, to provide healthy food, are needed to address a growing world population confronted with environmental and social challenges.
- Chief among the improvements required is the ability to more accurately use genotype to predict phenotype (G2P) in specific environments.
- Advances in the analysis of genome function promises to provide tools and knowledge for the genetic improvement of farmed animals and conservation of genomic biodiversity.
- The FAANG consortium is working to discover basic functional genomic knowledge underpinning the G2P question across major terrestrial and aquatic farmed animal species.
- In the coming decade, FAANG will associate large-scale genotypes and phenotypes in pedigreed animal populations to predict G2P relationships.
- The FAANG_{SingleCell} and FAANG_{GTEX} projects will apply exciting new technologies to model cells and tissues and verify which sequence variants affect phenotype.
- Public investment is needed to expand G2P research and training of new scientists to accelerate genetic improvement of diverse farmed animal populations to feed the global consumer.

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51

52 Here we describe a set of research priorities to meet such present and future
53 challenges that build on progress, successes and resources from the Functional Annotation of
54 Animal Genomes (FAANG) project [2]. The first stages of FAANG focused on foundational data
55 generation to characterise expressed and regulatory genomic regions, curation and provision
56 of annotated farmed animal genomes [2,3,4]. These were largely based on individual level,
57 high depth approaches [4]. The primary challenge facing this community now is harnessing
58 these resources to link genotype, phenotype and genetic merit in order to translate this
59 research out of the laboratory and into industry application in the field. To achieve this
60 effectively, we will need to generate functional genomic information for large populations of
61 animals, rather than relying on a small number of deeply annotated individuals. Furthermore,
62 to date, most of the datasets are from tissues consisting of heterogeneous cell populations,
63 hindering the resolution of functional information and limiting our ability to understand the
64 fundamental cellular and subcellular processes underlying phenotypes. Since the original
65 FAANG white paper was published in 2015 [2], exciting new opportunities have arisen to

66 tackle these challenges. We describe a set of research action priorities for FAANG for the next
 67 decade (Figure 1), in each of the sections below.
 68



69
 70 **Figure 1: Research priorities for the next decade of FAANG research.**

71
 72 **Omics empowered genomic selection**

73
 74 In the past twenty years genomic selection has substantially increased genetic gain in
 75 some farmed animal species through the use of large training populations [5]. However,
 76 prediction accuracy in genetically distant populations (i.e. across populations, breeds, and
 77 generations) remains limited due in part to the current reliance on neutral markers in
 78 incomplete linkage disequilibrium with causative genetic variants in the breeding population
 79 of interest [5]. Using variants more tightly linked to causative variants and supported by

80 genomic information in a multi-breed training population can partially alleviate these
81 limitations [6]. Large-scale whole-genome resequencing has produced inventories of many
82 millions of variants for thousands of animals [7]. In such sequence datasets, the causative
83 variants are directly genotyped among millions of neutral markers. This reduces the signal-to-
84 noise ratio when all the data are used for genomic prediction without prior biological
85 information. Efforts to detect causative variants have been successful for variants with large
86 phenotypic effects, often deleterious, using a combination of quantitative, population and
87 molecular genetics [5]. However, economically important traits have a polygenic architecture
88 and causative variants are expected to have small effects, which makes their detection and
89 quantification difficult. Most of these causal variants, with small effects, are likely to be
90 located in regulatory sequences and impact complex traits through changes in gene
91 expression [5]. Thus, it is expected that improvements in prediction accuracy can be achieved
92 by filtering the genetic marker information based upon whether the genetic variants reside in
93 functional sequences and developing robust prediction models that can accommodate the
94 biological priors. As functional (expressed and regulatory) genomic elements are not easy to
95 predict from sequence alone, FAANG will enhance current genome annotation with functional
96 information from a range of relevant tissues, cell types and developmental stages. Recently,
97 novel methods for the integration of biological information (e.g. methylation of regions of
98 predicted functionality) into genomic prediction have been proposed e.g. [6]. These models,
99 which are based on the combination and ranking of many diverse datasets from multiple
100 animals, could facilitate further improvements in predicting genetic merit and consequently
101 on genomic selection, as has been demonstrated in cattle [6]. As many more suitable datasets
102 will become available in the next five years, improving and adapting these methods to
103 enhance genomic prediction accuracy, whilst conserving genetic diversity, across farmed
104 animal species will be a priority for FAANG.

105

106 **FAANG_{GTEX} - Linking genetic variation to genome function**

107

108 The first phase of FAANG is using a specific set of transcriptomic and epigenomic
109 assays to define functional regions of the genome in tissues [2,3]. Due to the significant
110 investment per sample, this phase was limited to only a few individuals and ascribed function
111 was averaged across these replicates [2]. Progress has been made in defining functional
112 regions, and this should be built upon to ascertain the effect of genetic variation on genome
113 function [4]. Collecting functional genomic data across many genetically diverse animals lends
114 itself to the application of statistical genomics to detect quantitative trait loci (QTL) controlling
115 molecular phenotypes. This is particularly powerful when done at sequence level resolution
116 to directly relate molecular phenotypes (e.g. gene expression or methylation information) to
117 variants associated with complex traits. The GTEx consortium has achieved this very
118 effectively across human tissues, enabling expression QTL (eQTL) studies linking gene
119 expression to genetic variation [8], and providing a framework for FAANG to develop a similar
120 project for farmed animals (FAANG_{GTEX}). Large farmed animal cohorts in controlled and well
121 characterised environments with extensive pedigree information and molecular phenotypes
122 would allow researchers, in partnership with industry, to: 1) build better predictive models of
123 genotype-to-phenotype, 2) better understand genotype-by-environment interactions, and 3)
124 prioritise functional variants for inclusion in breeding programmes [5]. Hundreds of thousands
125 of farmed animals currently have imputed genotypes and extended pedigrees with deep
126 phenotypic records [7]. A project analysing the relationship between GWAS SNPs and gene
127 expression for cattle, mining publicly available sequence data, was published earlier this year,
128 demonstrating the feasibility, timeliness and potential of a GTEx approach for farmed animals
129 [9].

130

131

132 **Beyond genomic selection: towards genome-enabled management**

133

134 Beyond its use in genomic prediction [5], the functional data produced by FAANG will
135 provide new perspectives for informed management decisions. Epigenetic and expression
136 information for individual animals could be combined with microbiome data and high-
137 throughput phenotypes from new management technologies (e.g. wearables, GPS, *in-vivo*
138 imaging systems) [10]. These datasets from large cohorts of animals would enhance prediction
139 of adaptive capacity at the individual level through integration of prior environmental data
140 with individual genome information. Thus, providing new opportunities for informed animal
141 management decisions during an animal's lifetime (e.g. to optimize diets or for steering
142 animals into the most appropriate production systems). A genome enabled management
143 approach (providing animals, within a production system, with their specific needs during
144 their lifetime) will be beneficial to improving animal health and welfare, facilitate adaptation
145 to changing environments and contribute to addressing public concerns related to animal
146 production. Achieving this within the next ten years may be possible, but the challenge will be
147 to ensure it is practical and affordable for animal breeders and producers.

148

149 **Understanding and conserving genomic diversity – the power of pangenomes**

150

151 Through large-scale sequencing efforts by the farmed animal genomics community
152 data are now accumulating that characterise the sequence diversity of farmed animals
153 including locally adapted breeds/populations. As a consequence, future genetic management
154 is likely to include the use of pangenomes, that will capture all available population-level
155 genomic information for a given farmed animal species. Using graph-based frameworks, we
156 can more accurately genotype and annotate the genomic diversity present in any given
157 individual [11]. In this way, pangenomes can reveal population- or breed- specific adaptations

158 that could be used to tailor the genotypes chosen in future farming systems in order to
159 conserve biodiversity whilst improving production efficiency and animal health [1].
160 Furthermore, the highly annotated genomes produced by FAANG allows evolutionary
161 conservation across species to be defined for all genomic features [3,12]. Ongoing FAANG
162 projects involve comparative analysis which will reveal the functional basis of phenotypes
163 present in one species that are desirable in others. Such projects contribute to addressing the
164 major opportunity that exists to enhance the sustainable production of a wider diversity of
165 animal species, including numerous and diverse aquaculture species that are poised to exploit
166 functional genomics to expedite genetic improvement, where tailored and cost-efficient
167 approaches will be required [13]. Current FAANG-related projects already extend to several
168 major farmed finfish species in Europe and North America. We envisage an increased
169 representation of aquatic species, including shellfish, and further expansion to include
170 invertebrates, within FAANG projects during the next five to ten years.

171

172 **FAANG_{SingleCell} – deconvoluting transcriptional and regulatory complexity**

173

174 The use of bulk tissue samples in the FAANG studies performed to date captures
175 regulatory elements and expression signals averaged across all represented cell types but fails
176 to reveal the cell-specific basis of the molecular phenotypes of interest. In order to more
177 accurately link genotype to phenotype, data at the level of individual cell types are required.
178 Single-cell sequencing technologies enable the deconvolution of the transcriptional and
179 regulatory complexity in tissues made up of multiple cell types. New technologies to detect
180 gene expression as well as chromatin accessibility, structure, and interactions within single
181 cells provide more comprehensive data to predict function and interaction partners for
182 regulatory elements. As a consequence, one of the main priorities for FAANG within the next
183 five to ten years is to create single-cell atlases for the key tissues of farmed animal species

184 (FAANG_{SingleCell}). The organisational processes, standardisation and data sharing infrastructure
185 established by the community for the first stages of FAANG [4] will provide a strong
186 foundation for FAANG_{SingleCell} to progress quickly and efficiently. The FAANG_{SingleCell} project
187 should build on existing functional tissue maps for other species e.g. [14] and will enable the
188 identification of genomic variants underpinning trait-linked cell types/factors and causal
189 variants [13]. In the FAANG_{GTEX} project described above, single-cell atlases will provide a
190 powerful layer of resolution including cell-specific molecular phenotypes, enabling the fine-
191 scale dissection of complex traits of interest.

192

193 ***In vitro* systems - bridging the gaps between cell, tissue and whole animal scale knowledge**

194

195 Single-cell sequencing technologies can also be used to deeply characterise cell and
196 tissue complexity of *in vitro* systems such as organoids. Over the last five years organoids for
197 many different organ systems and for multiple farmed animal species have been developed
198 [4]. Organoids provide *ex vivo/in vitro* systems for testing candidate causal variants by genome
199 editing technologies, and potentially a system for high-throughput, cost-effective, large-scale
200 *in vitro* phenotyping. Importantly, given the ease of biobanking, organoids have a strong
201 ethical benefit in reducing the number of animals used in experimentation [4]. Multiple
202 organoid models can be derived from very small quantities of tissue or from induced
203 pluripotent stem cells (iPSCs). They provide the potential to generate and test multiple
204 phenotypes to unravel when, and under what conditions, a putative causal variant has an
205 effect. Therefore, farm animal organoids will be valuable over the coming decade, providing
206 information about fundamental biology to model the effects of changing environmental
207 conditions and supporting immunology, vaccinology, physiology, nutritional, and biodiversity
208 conservation studies. The ability to decompose complex phenotypes into key processes will

209 provide a means to robustly relate the deep phenotypes measured in these systems with the
210 traits used for selection, opening to the possibility of using organoids for breeding purposes.

211

212 **Genome editing - a route to application for FAANG data**

213

214 The application of genome editing to farmed animals is advancing rapidly, mainly due
215 to development of CRISPR/Cas technologies [13, 15]. The CRISPR toolbox has expanded to
216 improve precision, allow modulation of gene expression and epigenetic modifications, and
217 now forms an integral part of the future FAANG roadmap [4]. CRISPR-mediated modification
218 of putative genomic elements can confirm their functionality and reveal their roles in cellular
219 (and organoid) function. Genome-wide multiplexed CRISPR approaches now enable the
220 simultaneous interrogation of thousands of genomic features in cell lines, increasing the
221 feasibility of this approach for genome-scale annotation [16]. These high-throughput
222 approaches can also be used in combination with single-cell sequencing technologies to
223 obtain high-resolution molecular phenotypes. In addition, genome editing represents a
224 potential major route for the application of FAANG research in farmed animal breeding
225 programmes via: 1) detection and utilisation of causative variants affecting important traits,
226 2) targeted introgression, or 'introgression-by-editing', of favourable alleles from other strains
227 or species into a closed breeding population, or 3) creation of *de novo* alleles with favourable
228 effects, either predicted from unbiased genome-wide screens or from *a priori* knowledge of
229 the biology of the trait in question. Public perception and regulatory hurdles remain and
230 ongoing discussion through stakeholder engagement must continue and evolve to keep pace
231 with technological advances. While the use of genome editing for the improvement of farmed
232 animals may currently only be possible in some countries, its use in *in vitro* models, such as
233 organoids, is not subject to the same legislation and ethical considerations as the use of whole
234 animals and thus represents a new frontier for FAANG research.

235

236 **Data recording, computation and integration to support the emerging objectives of FAANG**

237

238 As a scientific community FAANG continues to develop a coordinated analysis and
239 data collection infrastructure crucial for its success [4]. The FAANG bioinformatics community,
240 including the centralised Data Coordination Centre (DCC), is focused on open reproducible
241 science, the FAANG data portal [17] as the focal point for this activity. Technological
242 development, coordination and standardisation by the DCC will continue to be crucial for the
243 shift towards population scale studies, single cell datasets, cell atlases and pangenomes,
244 across a growing number of species. This will require new reproducible analysis pipelines and
245 infrastructure, metadata validation services, data portal features such as a centralised atlas
246 browser and online training resources. Single cell atlases and *in vitro* systems for farmed
247 animal species will be accompanied by high quality metadata, archiving and visualisations
248 across species, organ systems, tissues and cell types. As FAANG datasets continue to increase
249 in complexity, there is a growing need for new methods of data visualisation and integration
250 to be made available. These future developments, and the distributed data and analysis
251 infrastructure, will be crucial for the successful application of functional data to farmed animal
252 breeding programmes.

253

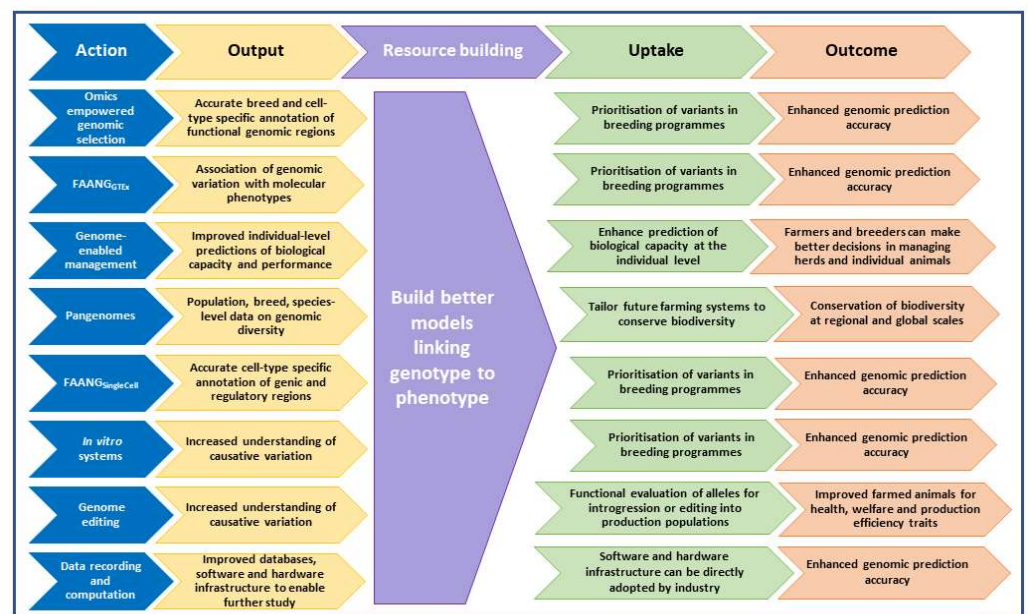
254 **Priorities for the future of FAANG**

255

256 The research priorities we have outlined for FAANG for the coming decade are
257 depicted in Figure 1. The uptake by the farmed animal production industry and the expected
258 outcomes of each prioritised action are summarised in Figure 2. FAANG will improve our
259 ability to more accurately use genotype to predict phenotype. This will directly contribute to
260 addressing the challenges faced for sustainable and responsible global food production in the

next decade (Box 1). However, whilst the molecular assays used to enable functional
 annotation can now be delivered at much lower cost, the costs for the research priorities
 outlined above remain substantial, especially considering the rapid increase in number and
 diversity of target species in the aquaculture sector. As such a strong commitment to invest
 in research is needed. Persuading the European Commission to include FAANG projects in
 Horizon 2020 was a major success for the first stage of FAANG and its leadership. Current
 funding for FAANG supports the research community to improve the functional annotation of
 key farmed animal species and to facilitate more refined genomics-enabled animal breeding
 / genetic improvement. The research priorities outlined here are already strategically aligned
 to the objectives of the European Green Deal and current USDA National Institute for Food
 and Agriculture programmes, and international cooperation will be essential to secure funding
 for their achievement. Given the scale and cost of the research involved it will likely be
 necessary to initially prioritise the development of *in vitro* systems and the enhancement of
 data infrastructure to provide a solid foundation for FAANG_{SingleCell} and FAANG_{GTE}.

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276

277

Figure 2: How implementation of FAANG research priorities over the next decade will benefit farmed animal production.

The timely achievement of all of the research priorities we outline here for the next stages of FAANG will together increase the capacity of the farmed animal production industry to face the challenges of the future, empowering genomic selection, enhancing adaptation to changing environments, conserving biodiversity, and bridging the knowledge gaps between cellular and whole animal scale knowledge.

Authors' Contributions

All co-authors contributed to the writing of the manuscript. ELC assembled and curated the manuscript in collaboration with the co-authors, with joint editing by MW and FAANG co-coordinators (CKT and EG).

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319

320 **Competing Interests**

321 The authors declare that they have no competing interests.

322

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324

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