The development and application of genomic selection as a new breeding paradigm

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Implications

- A significant gap exists between demand based on population growth and the current trajectory of yield; this is a point of leverage for genomics.
- Although traditional breeding methods have been effective in selecting for animals with easy-to-measure production traits, these methods have essentially “hit an asymptote,” and more difficult-to-measure (and often more important) traits cannot effectively be selected for using traditional methods.
- The race to sequence the first human genome, and, subsequently, the race to enable routine resequencing of tens, if not hundreds of thousands, of additional human genomes, has resulted in a 100 million fold decrease in DNA sequencing prices since 1990. Animal genome sequencing has benefitted from this.
- Sequencing and resequencing of economically important livestock species has resulted in the discovery of millions upon millions of single nucleotide polymorphisms. These single nucleotide polymorphisms are being deployed in massively parallel fashion on DNA microarrays, enabling genome-wide association studies to identify genotype-phenotype correlations for both simple and, more important, complex traits.
- Driven by ever-increasing reductions in the cost of measuring genetic variation, we are entering a new era in which the information from these genome-wide association studies will be utilized effectively in routine testing using genomic selection. Genomic selection holds promise for more widespread adoption than marker-assisted selection because it lacks the requirement of prior knowledge of alleles or marker positions of loci and the requirement that marker-assisted selection must be implemented within families.

Key words: gene mapping, genomic breeding, genomic selection, next-generation sequencing, single nucleotide polymorphism

Introduction

In the past 50 years, the world has experienced an unprecedented increase in population growth. On the basis of current projections, the world population will reach 9 billion people by 2030. Meeting the growing food need using fewer resources is therefore one of the greatest challenges that contemporary agriculture is facing. Recent estimations by the Food and Agriculture Organization of the United Nations (FAO, 2006) indicate that to meet the increasing demand, food production must double in the next 50 years. Expressed another way, agriculture will have to produce more food than in the last 10,000 years combined. When the year 2000 is used as a base, projections indicate an increase in global meat consumption of 68% and in global milk consumption of 57% by 2030 (Steinfeld and Gerber, 2010). The greater demand for food based on animal proteins together with the potential effects of climate change and water, nutrient, and energy scarcity will result in large productivity gaps. It is therefore critical to apply technical and scientific advancements systematically in feeding, nutrition, genetics, reproduction, animal health control, and general improvement of animal husbandry to fill the coming productivity gaps. The largest gains will come from innovations that accelerate agriculture productivity while reducing costs and limiting environmental impacts (Capper, 2011).
Big Expectations of Genomics

Over the past 2 decades, the rapid development of genomics has opened new paths to address the scientific basis of livestock biology and breeding, and has resulted in new production methods to achieve sustained increases in animal feed yields and long-term improvements in the efficiency of livestock production. A new era, the “genomics era,” promises to enable the objective prediction of consequences based on direct access to the full DNA sequence of many individuals, and therefore a renewed and more objective view of the genetic value of animals that is not limited to a few production traits.

One of the triggering factors for the development of this genomic era was the international project to sequence the human genome (the Human Genome Project). The goal of this project was to produce the first (de novo) full DNA sequence of a human being. Along with it came the development and implementation of new genomic tools, particularly improved DNA sequencing technologies and increased availability of high-throughput genotyping platforms. The price to sequence a single nucleotide of DNA has fallen 100 million fold since 1990. This is the equivalent of filling up your car with gas in 1998, waiting until 2011, and being able to drive to Jupiter and back twice. As Figure 1 illustrates, the rate of information coming from current-generation DNA sequencers is increasing exponentially, faster than Moore’s law, and faster than other comparable high-growth scenarios. Technological breakthroughs that have driven this cost decrease have also facilitated the production of whole-genome sequences for several animal species (Table 1).

Genomics: Moving Animal Science to a New Dimension

From a scientific point of view, accelerated genome-scale measurement will have a profound impact. It will fuel comprehension of the basic structure and function of livestock genomes; help unravel the history of life, characterizing the cause of relatively simple phenotypes and genetic diseases; and further explain the control of complex characteristic features. It should accelerate the discovery process and help close the gap between genetics and the traits that are observed, known as the “genotype/phenotype gap.”

With the production of whole-genome sequences for the major livestock species at a fraction of the cost of the Human Genome Project, the comparison of sequences from several individuals of different breeds with a reference sequence resulted in an almost inexhaustible source of genetic markers, primarily polymorphisms in the form of single nucleotide polymorphisms (SNP). With the release of the first draft of the chicken genome sequence (International Chicken Genome Sequencing Consortium, 2004), the chicken community was the first animal community not only to have access to millions of SNP, but also to be organized in a freely accessible database, the Chicken Variation Database (ChickVD; http://chicken.genomics.org.cn/). Similar sequencing and resequencing efforts in several other livestock species resulted in the discovery of hundreds of thousands of SNP covering the entire genome (Ramos et al., 2009). These genetic markers were subsequently deposited in a publicly accessible database maintained by the National Center for Biotechnology Information, the dbSNP (http://www.ncbi.nlm.nih.gov/projects/SNP). This database currently contains SNP information for more than 86 organisms and a total of more than 50 million SNP. Another major technological breakthrough has been the development and constant improvement of DNA array technology, which allows for the inexpensive measurement of SNP within a given sample. The success of these DNA arrays resides in the fact that they present a strong parallel-processing capacity, tremendous miniaturization, and a remarkable ability to be automated. Although first used for gene expression studies, these DNA arrays proved very useful for the development of whole-genome SNP panels for many species, including several agriculturally relevant species (Table 2). With DNA arrays, hundreds of thousands of SNP can be screened in parallel for the cost of a few hundred dollars, allowing scientists to perform genome-wide association studies that simply would have been out of reach with traditional (i.e., microsatellite) markers. Over the past years, several studies have been published dem-

Table 1. Summary of first sequenced genomes for animal species

| Species                      | Genome size (assembly), Gb | Year |
|------------------------------|---------------------------|------|
| Chicken (Gallus gallus)      | 1.05                      | 2004 |
| Dog (Canis familiaris)       | 2.4                       | 2003 |
| Cattle (Bos taurus)          | 2.91                      | 2009 |
| Horse (Equus caballus)       | 2.47                      | 2009 |
| Pig (Sus scrofa)             | 2.2                       | 2009 |
| Sheep (Ovis aries)           | 2.78                      | 2008 |
| Cat (Felis catus)            | 1.64                      | 2006 |
| Rabbit (Oryctolagus cuniculus)| 2.67                     | 2009 |
| Turkey (Meleagris gallopavo) | 1.08                      | 2009 |
| Dromedary (Camelus dromedarius) | 2.2                    | 2011 |
| Medaka (Oryzias latipes)     | 0.7                       | 2011 |
| Honeybee (Apis mellifera)    | 0.236                     | 2011 |

1Modified from Fan et al. (2010).
Demonstrating the effectiveness of using low-cost, whole-genome SNP arrays (e.g., BovineSNP50, OvineSNP50, EquineSNP50, PorcineSNP60) for the fine-scale mapping of inherited defects when testing only a modest number of cases and controls (Becker et al., 2010; Brooks et al., 2010; Meyers et al., 2010).

- For the mapping of quantitative trait loci and DNA regions involved in the genetic mechanism of complex phenotypes, to discover and estimate marker effects (Fortes et al., 2010; Minozzi et al., 2010; Pryce et al., 2010).
- For the characterization of population structure, to better document and understand the history of our livestock populations (Decker et al., 2009; Kijas et al., 2009; Gautier et al., 2010).

The availability of whole-genome SNP panels is therefore bolstering the search for mutations underlying genetic variation in simple and complex traits and revolutionizing the speed at which gene regions and specific genes are being discovered. With the use of whole-genome SNP panels, the traditional approach, namely, positional cloning with a whole-genome scan to map the region of interest, followed by a time-consuming fine-mapping step, is fully replaced with an efficient and cost-effective genotyping step using a whole-genome SNP array containing 50,000 to 70,000 SNP.

### Table 2. Currently available whole-genome single nucleotide polymorphism (SNP) chips developed for important agricultural species

| Species | Identification\(^1\) | Classification | Provider\(^2\) | Consortium | SNP, no. |
|---------|----------------------|----------------|-------------|------------|----------|
| Potato  | Potato               | Public         | Illumina    | SolCAP     | 8,303    |
| Tomato  | Tomato               | Public         | Illumina    | SolCAP     | 7,720    |
| Apple   | Apple                | Public         | Illumina    | RosBREED   | 8,788    |
| Peach   | Peach                | Public         | Illumina    | RosBREED   | 8,144    |
| Cherry  | Cherry               | Public         | Illumina    | RosBREED   | 5,696    |
| Maize   | MaizeSNP50           | Commercial     | Illumina    | Commercial | 56,110   |
| Rice    | Rice 44K             | Commercial     | Affymetrix  | Commercial | 44,100   |
| Chicken | Chicken              | Private: public sale | Illumina | Cobb Vantress-Hendrix-USDA | 57,636 |
| Cat     | Feline               | Private: public sale | Illumina | Morris Animal Foundation | 62,897 |
| Horse   | Equine               | Private: public sale | Illumina | Neogen (GeneSeek) | 65,157 |
| Sheep   | Ovine                | Private: public sale | Illumina | AgResearch | 5,409 |
| Cattle  | BovineHD             | Commercial     | Illumina    | Various    | 777,962  |
| Cattle  | BovineSNP50v2        | Commercial     | Illumina    | Various    | 54,609   |
| Cattle  | BOS 1                | Commercial     | Affymetrix  | Various    | 648,000  |
| Sheep   | OvineSNP50           | Commercial     | Illumina    | Various    | 52,241   |
| Cattle  | BovineLD             | Commercial     | Illumina    | Various    | 6,909    |
| Pig     | PorcineSNP60         | Commercial     | Illumina    | Various    | 62,163   |
| Dog     | CanineHD             | Commercial     | Illumina    | Various    | 173,662  |

\(^1\)HD = high density; LD = low density.

\(^2\)Illumina Inc., San Diego, CA; Affymetrix, Santa Clara, CA.

### Genomics: A Paradigm Shift in Animal Breeding (Toward the Genome-Assisted Barnyard)

The biggest (r)evolution is taking place in the application of genomics to the design and implementation of livestock breeding programs, promising gains across the value chain. For breeders, breeding organizations, and members of the livestock industry, genomics is expected to increase the efficiency and productivity of animal breeding, whereas for consumers and the processing sector, it should enhance security and the quality of animal products. New insights into the growth, nutrition, health, and protection of animals are expected, enabling a better understanding of the molecular mechanisms of traits of interest. Therefore, genomics proposes further opportunities to improve selection accuracy while decreasing the costs, reducing generation intervals, and exploiting new sources of polymorphisms (Dekkers, 2004).

### From Marker-Assisted Selection to Genomic Selection

Beginning in the 1990s, breeders used gene marker technology in the form of marker-assisted selection (MAS) to remove deleterious gene alleles (such as halothane in pigs or fish taint in brown-shelled chicken eggs) or to select favorable conditions based on some marker information (such as the Polled condition in cattle). The limitation of MAS is that it
requires prior knowledge of gene alleles or markers that are associated with the traits of interest together with quantitative estimates of these associations in the specific population. It must therefore be implemented within families. Furthermore, MAS explains only a limited part of the genetic differences between individuals. However, with the availability of cost-effective whole-genome SNP panels for the major livestock species, genomic selection tools are now leading the way to a paradigm shift in animal breeding. With sufficient genetic markers, one can follow the segregation of the entire genome and not merely a set of specific regions of interest, moving from MAS to genomic selection. Parental relationships are no longer required to explain similar performances in animals because, with a dense set of SNP, similar performances can now be explained by the fact that animals are sharing identical chromosome fragments.

**Principle of Genomic Selection**

Genomic selection was first described by Meuwissen et al. (2001) and is based on the fundamental principle that information from a large number of markers could be used to estimate breeding values without having a precise knowledge of where specific genes are located on the genome. With tens of thousands of SNP, well chosen to be representative of the entire genome, it is expected that there will always be an SNP in close proximity to a particular gene or DNA fragment of interest; the existing linkage disequilibrium between one (or several) SNP and a causal mutation will be substantial and can then be used to explain a significant fraction of the variation of the observed trait. The first step in the genomic selection process is therefore access to a large group of animals, either a reference or training population with accurate phenotypes for the trait(s). This population should also be genotyped using a whole-genome SNP array. The resulting data will then serve as a reference to develop a statistical model estimating the effect of each SNP with the trait(s) of interest. The result is a predictive equation to calculate a genomic estimated breeding value (GEBV). After a validation step, the genomic breeding value of new animals can be computed using the prediction equation, based on their genotypes from the SNP array and in the absence of any accurate phenotypes for these animals (Figure 2 and Table 3). The accuracy of the GEBV depends on the size of the population and the heritability of the trait to be considered. In the chicken, for example, González-Recio et al. (2009) referenced a 4-fold increase in GEBV accuracy over the parent average for feed conversion efficiency.

**Implementation of Genomic Selection**

Genomic selection builds on existing breeding programs in which the collection of pedigree information together with phenotypic data is already routine; it provides a new level of information that can be integrated into the decision-making process to identify and select the most promising animals. The principal advantages of genomic selection are that it can be

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**Figure 2.** A reference population of animals is scored for key production traits and genotyped using a commercial or custom single nucleotide polymorphism array. The genotypes are represented by the variable x, with values 0, 1, 2 (homozygous, heterozygous, or alternate homozygous). A prediction equation is generated, combining all the marker genotypes with their effects to compute a genomic estimated breeding value for each animal. This prediction equation can be applied to a group of animals that have not been phenotyped, breeding values can be estimated, and the best animals can be selected for breeding. Adapted from Goddard and Hayes (2009) by permission from Macmillan Publishers Ltd: Nature Reviews Genetics, © 2009.
Next Wave of Discoveries and New Approaches

Higher density SNP arrays with several hundred thousand SNP are already being developed in several livestock, crop, and companion animal species. In cattle, the success of genomic selection is being extended by combining several breeds to increase the size of the reference population and perform cross-breed evaluations. This approach should prove very beneficial for breeds with a limited number of individuals or phenotypic records, or for species for which cross-breeding is an effective tool in the breeding process.

As the cost of sequencing continues to decrease and access to the whole-genome sequence for specific individuals becomes affordable, one of the next steps will be to include whole-genome sequence data in routine genetic evaluations. According to a simulation presented by Meuwissen and Goddard (2010), a 40% gain in accuracy in predicting genetic values could be achieved by using sequencing data instead of data from 30,000 SNP arrays alone. Furthermore, by using whole-genome sequencing data, the prediction of genetic value was able to remain accurate even when the training and evaluation data were 10 generations apart: observed accuracies were similar to those in which the test and training data came from the same generation. According to the authors, “these results suggest that with a combination of genome sequence data, large sample sizes, and a statistical method that detects the polymorphisms that are informative . . ., high accuracy [in genetic/genomic prediction] is attainable” (Meuwissen and Goddard, 2010, p. 630).

However, genomic selection today still treats the genome as a “black box.” It is not necessary to understand what is inside the black box to make effective selection decisions. From this point of view, it is therefore not much different from the broadly used and accepted best linear unbiased prediction method. Genomic selection could be further improved by integrating pertinent biological information and using efficient methodologies to get from the knowledge of a statistically associated marker locus to a functional gene variation. This would move from simply associating sequence variation with distinct phenotypes to a true understanding of the biology of the animal that makes those variations significant, effectively shedding light on the black box.

Opportunities for Developing Countries

In developed countries, phenotypes and pedigrees have been recorded for certain species, such as dairy cattle, for more than 100 years. Progeny testing has been implemented for nearly 50 years. Developing countries are often limited by the absence of programs that record phenotypes on pedigreed animals and the lack of evaluation or national testing programs to assess the genetic value of germplasms. Genomic approaches should help in identifying critical populations for preservation together with some local well-adapted breeds that could be further utilized to breed valuable animals through a combination of selection and cross-breeding. Of course, as with genomics, you can manage only what you can measure, and collecting a minimum number of phenotypes in the field will remain one of the critical and challenging steps to further deployment of genomic selection in developing countries.

Conclusion and Perspectives for the Future

The ability to investigate the genome, the transcriptome, the epigenome, and the metagenome of any species by high-throughput sequencing methods is opening a new world of possibilities. Further reduction in sequencing costs will continue to drive broader acceptance of new approaches and their implementation for the benefit of animal research, the breeding industry, and consumers. All economically impactful agricul-

| Animal | SNP 1 Genotype | SNP 1 Value | SNP 2 Genotype | SNP 2 Value | SNP 3 Genotype | SNP 3 Value | SNP 4 Genotype | SNP 4 Value | Genomic breeding value |
|--------|----------------|-------------|----------------|-------------|----------------|-------------|----------------|-------------|------------------------|
| 1      | AA 8           | BB −4       | AA 2           | AA −6       | AA 0           | BB 6        | BB −2          | AB 0         | 10                     |
| 2      | AA 8           | AA 4        | BB −2          | AA −6       | AB 0           | BB 6        | BB −2          | AB 0         | −8                     |
| 3      | AB 0           | AB 0        | AB 0           | AA −6       | AA 0           | BB 6        | BB −2          | AB 0         | 10                     |
| 4      | BB −8          | AA 4        | AA 2           | AA −6       | AA 0           | BB 6        | BB −2          | AB 0         | −8                     |
| 5      | BB −8          | BB −4       | BB −2          | AB 0        | BB 6           | BB −2       | AB 0           | BB 6         | −14                    |

Table 3. Example of a simplified calculation of the genomic breeding value with 4 single nucleotide polymorphisms (SNP) and estimated effects (allele A vs. B) of +8, +4, +2, and −6 for SNP 1, 2, 3, and 4, respectively.
tural species, subspecies, and their pathogens will no doubt be sequenced in the near future. Thousands of related genomes will also be sequenced to sample genetic diversity within and between germplasm pools, offering critically important information for the implementation of genomic selection programs in developed countries. Genomic selection will surpass conventional methods as the dominant breeding paradigm, and specific haplotypes, detected via high-throughput genotyping or sequencing, will be directly associated with economic values. Breeding programs will be driven primarily by array data because of superior economics and much higher throughput. New expertise in the field of animal pharmacogenomics will help increase vaccine and drug specificity, whereas nutrigenomics will help tailor feeding regimens to genomic profiles.

As genomic information continues to provide hugely valuable biological information, the key for further success of genomic selection and genomic approaches will be to collect the most pertinent phenotypes, identify the causal mutations and the exact mechanisms by which phenotypes are produced, and bring the different superior variants together in breeding lines in as few generations as possible. We are entering a truly exciting era fueled by genomics.

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