Using diverse U.S. beef cattle genomes to identify missense mutations in EPAS1, a gene associated with pulmonary hypertension [version 2; referees: 2 approved]

Previously titled: Using diverse U.S. beef cattle genomes to identify missense mutations in EPAS1, a gene associated with high-altitude pulmonary hypertension

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Abstract
The availability of whole genome sequence (WGS) data has made it possible to discover protein variants \textit{in silico}. However, existing bovine WGS databases do not show data in a form conducive to protein variant analysis, and tend to under represent the breadth of genetic diversity in global beef cattle. Thus, our first aim was to use 96 beef sires, sharing minimal pedigree relationships, to create a searchable and publicly viewable set of mapped genomes relevant for 19 popular breeds of U.S. cattle. Our second aim was to identify protein variants encoded by the bovine endothelial PAS domain-containing protein 1 gene (EPAS1), a gene associated with pulmonary hypertension in Angus cattle. The identity and quality of genomic sequences were verified by comparing WGS genotypes to those derived from other methods. The average read depth, genotype scoring rate, and genotype accuracy exceeded 14, 99%, and 99%, respectively. The 96 genomes were used to discover four amino acid variants encoded by EPAS1 (E270Q, P362L, A671G, and L701F) and confirm two EPAS1 variants previously associated with disease (A606T and G610S). The six EPAS1 missense mutations were verified with matrix-assisted laser desorption/ionization time-of-flight mass spectrometry assays, and their frequencies were estimated in a separate collection of 1154 U.S. cattle representing 46 breeds. A rooted phylogenetic tree of eight polypeptide sequences provided a framework for evaluating the likely order of mutations and potential impact of EPAS1 alleles on the adaptive response to chronic hypoxia in U.S. cattle. This public, whole genome resource facilitates \textit{in silico} identification of protein variants in diverse types of U.S. beef cattle, and provides a means of translating WGS data into a practical biological and evolutionary context for generating and testing hypotheses.
Introduction

The number of disease-causing mutations discovered in bovine genes has increased with the advancement of next-generation sequencing, reference genomes, and high density single nucleotide polymorphisms (SNPs) arrays. There are currently 130 Mendelian traits with known causal mutations in 117 cattle genes. The majority of these mutations cause disease by changing the gene’s protein sequence and thereby altering its normal function. As the list of major genes influencing bovine traits grows longer, there is increasing interest in the protein variants encoded by these genes, either in the source population, or in other uncharacterized populations. For example, in 2006 when a K211 mutation was discovered in the prion gene of a U.S. cow that developed atypical BSE, there was a pressing need to search for this potentially pathogenic allele in other U.S. cattle. In principle, these searches can be accomplished in silico with access to population-scale gene sequence data.

Gene function can be affected by a wide range of genomic sequence differences including: large scale structural polymorphisms like translocations, inversions and copy number variants; and small scale differences like methylation, insertions/deletions (indels), and SNPs. Variants that alter amino acid sequences, such as missense, nonsense, frameshift, and splice site mutations, are among those most likely to affect function. These variants are readily identified by aligning genomic sequences of animals to an annotated reference genome assembly. An essential first step in understanding a gene’s potential influence on a trait, is determining whether any protein variants are encoded in a set of reference individuals.

Access to population-scale gene sequence data, however, has been a limiting step for biomedical veterinary researchers studying U.S. cattle. The reagent costs for a traditional, single-gene approach with PCR-based Sanger exon resequencing of a population set can exceed tens of thousands of dollars, and take more than a year to complete. In addition, exon resequencing is a recurring cost for each gene analyzed. Approaches that use whole exome sequencing are less costly per gene, but incompletely target some bovine genes. Gene coverage with whole genome sequence (WGS) approaches is limited only by the quality of the reference genome, and the amount of data produced. WGS approaches have the advantage of high genotype accuracy and a low cost per gene, as the sequencing only needs to be accomplished once. The primary disadvantages of WGS approaches are the cost of production, the need for computing and informatics systems, and expertise for complex analyses. Regardless of the sequencing approach chosen, selecting the appropriate individuals for study, and verifying their identity and data integrity is essential.

Population-scale WGS data has been reported for a number of major cattle projects and is available at the National Center for Biotechnology Information (NCBI) through BioProjects and the Sequence Read Archive (SRA). Projects include 234 purebred cattle from global Holstein-Friesian, Fleckvieh, Jersey, and Angus breeds; and 379 Canadian cattle from six purebred beef breeds, three crossbred beef breeds, and the Holstein dairy breed. Key influential animals were chosen for these projects with the primary goal of using them to impute genetic variants and improve the accuracy of genomic prediction and genome wide association studies. However, these animals are not ideally suited for protein variant discovery across U.S. beef cattle populations, and the SRA data files are not readily searchable by gene. A third related population-scale WGS BioProject used 154 U.S. bulls from seven popular breeds and a mix of 116 crossbred and purebred animals from eight other breeds. However, the average genome coverage for these influential sires was 2.5 fold, and thus, insufficient for determining genotypes of individual animals.

The present report describes a publicly available and searchable set of mapped genomes for 96 beef sires from 19 breeds of U.S. cattle. These sires were chosen to minimize the relationships shared between pedigrees. The average read depth for these genomes was 14.8 and access is available without restrictions. The genome sequences may be either viewed directly with open source, high performance visualization software via United States Department of Agriculture (USDA), Agricultural Research Service (ARS) internet sites, or downloaded from the SRA at NCBI. Visualization software, in combination with online access, allows users to navigate to a gene, visually identify, and accurately record protein variants occurring in U.S. beef cattle populations. Thus, if a gene mutation is reported in one breed, it can quickly be evaluated in other breeds, and searched for additional variants that may potentially affect protein structure.

The feasibility of this process was demonstrated for the bovine endothelial Per-ARNT-Sim (PAS) domain-containing protein 1 gene (EPAS1) encoding the hypoxia inducible transcription factor 2A (HIF2A). EPAS1 was selected for analysis because two linked missense mutations were reported to be associated with pulmonary hypertension (PH) in Angus cattle. PH causes right-sided heart failure (RHF) and has been recognized as an increasing problem in North American beef cattle in feedyards, and in dairy cattle. Moreover, cattle treated for bovine respiratory disease complex (BRDC) in feedyard environments were three times more likely to die from RHF, and died earlier in the feeding period. The linked missense mutations in EPAS1 encode threonine (T) and serine (S) at amino acid positions 606 and 610, respectively, and were associated with PH when
compared to the more common allele encoding alanine (A) and glycine (G) at these positions. Animals carrying one or two copies of the EPAS1 T606, S610 allele were significantly more likely to develop PH at high altitudes. Although other genes are likely involved, EPAS1 appears to be a major gene affecting PH in cattle.

The protein encoded by EPAS1 is one of three HIF-alpha proteins that plays an important role in transcriptional regulation of the hypoxic response in metazoans, and is highly conserved in mammals (reviewed in 18). HIF2A is an oxygen sensing subunit that forms a heterodimer with a common, constitutively expressed beta subunit, HIF1B. Each HIF heterodimer binds to specific hypoxia responsive elements and transcriptionally activates different sets of genes. HIF2A controls the chronic hypoxia response by binding small molecules, interacting with proteins, and binding to DNA. Amino acid substitutions in HIF2A have the potential to drastically affect those critical interactions.

In humans and mice, EPAS1 missense mutations affecting the HIF2A oxygen-dependent degradation domain (ODDD) cause a gain-of-function, activate the hypoxia pathway, and can cause familial PH19,20. The EPAS1 substitutions at positions 606 and 610 reported by Newman et al. are also in the ODDD of HIF2A, although not at the same positions as those in humans and mice. Although no additional EPAS1 variants were previously observed in exons from the 20 Angus cattle sequenced, a broader reference set of U.S. beef cattle was not evaluated. Here, we report the discovery of four additional EPAS1 missense mutations, a rooted phylogenetic tree of eight distinct HIF2A sequences, a genetic test for typing them, and HIF2A variant frequencies in a separate collection of 1154 U.S. cattle representing 46 breeds. The results illustrate the utility of the approach, and provide a resource for evaluating protein variants in specified genes of interest. Knowledge of cattle protein variants affecting function is critical for transitioning from a descriptive phase of genomics to an applied phase where animal health, welfare, and production may be improved.

Methods
Ethics statement
This article contains no studies performed with animal subjects. Archival DNA was used from extracts of samples that were either: purchased from commercial sources that collected them for artificial insemination of cattle and not for research, purchased from individuals that collected them privately for their purposes (such as food), or donated to the U.S. Meat Animal Research Center (USMARC) by private individuals that collected them privately for their own purposes.

Discovery and validation panels of cattle
The discovery panel consists of 96 unrelated individuals from 19 popular U.S. beef breeds (USMARC Beef Diversity Panel version 2.9 [MBCDPv2.9], Figure 1). The current panel design was based on a previous set of commercially-available sires from 16 breeds with minimal pedigree relationships (MBCDPv2.1). For both panels, pedigrees were obtained from leading suppliers of U.S. beef cattle semen and analyzed to identify unrelated individuals for inclusion. On the basis of the number of registered progeny, the breeds in the MBCDPv2.1 were estimated to represent greater than 99% of the germplasm used in the US beef cattle industry, contain more than 187 unshared haploid genomes, and allow a 95% probability of detecting any allele with a frequency greater than 0.016. As previously described, this “threshold” frequency was defined as the minimum allele frequency at which the probability of observing the allele at least once in an animal group was 0.95.
The probability of observing an allele at least once is $1 - (1 - p)^n$ where “p” is the frequency of the allele and “n” is the number of independent samplings. In this case, the number of unshared haploid genomes for diploid organisms. This assumes that samplings (haploid genomes) are independent and identically distributed (the same p applies to all animals). Setting power or the probability of observing the allele at least once to 0.95 results in the equation: $0.95 = 1 - (1 - p)^n$. Solving this equation for p yields $p = 1 - (0.05)^{1/5}$ for all p between 0 and 1. The panel was updated to increase the number of beef breeds from 16 to 19, and remove the Holstein breed which was well represented in other WGS datasets. To make room for three additional beef breeds (Braunvieh, Corriente, and Tarentaise), the maximum number of sires within a breed was reduced from eight to six (NCBI BioProject PRJNA324822).

A separate set of cattle samples was used to validate results obtained from the above discovery panel. The validation panel consisted of samples from male and female registered purebred cattle with diverse pedigrees. Samples were from semen, blood, or hair follicles, depending on gender and availability as previously described. Where possible, animals within breed were chosen so they did not share parents or grandparents, and none were closely related to the 96 sires in the MBCDPv2.9. The breeds and samples used were: Angus (n = 24), Ankole-Watusi (n = 24), Ayrshire (n = 24), Beefmaster (n = 24), Belgian Blue (n = 24), Blonde d’Aquitaine (n = 24), Brahman (n = 24), Brahmos (n = 24), Braunvieh (n = 24), Brangus (n = 24), Brown Swiss (n = 24), Charolais (n = 24), Chianina (n = 24), Corriente (n = 24), Devon (n = 24), Dexter (n = 24), Gelbvieh (n = 24), Guernsey (n = 25), Hereford (n = 24), Highland (n = 24), Holstein (n = 86), Indu-Brazil (n = 24), Jersey (n = 28), Limousin (n = 24), Maine-Anjou (n = 24), Marchigiana (n = 24), Mini-Hereford (n = 24), Mini-Zebu (n = 24), Montbeliarde (n = 24), Murray Grey (n = 21), Nelore (n = 24), Piedmontese (n = 24), Pinzgauer (n = 24), Red Angus (n = 24), Red Poll (n = 24), Romagnola (n = 24), Salers (n = 24), Santa Gertrudis (n = 24), Senepol (n = 23), Shorthorn (n = 24), Simmental (n = 24), Tarentaise (n = 24), Texas Longhorn (n = 24), Texas Longhorn, Cattlemen’s Texas Longhorn Registry (CTRLR, n = 22), Tuli (n = 24), and Wagyu (n = 24).

**WGS production, alignment, and SNP genotyping**

DNA was extracted from commercial semen with a typical phenol/chloroform method and stored at 4°C in 10 mM TrisCl, 1 mM EDTA (pH 8.0) as previously described. Approximately 5 μg of bovine genomic DNA was fragmented by focused-ultrasonication to generate fragments less than 800 bp long (Covaris, Inc. Woburn, Massachusetts USA). These fragments were made to be indexed, 500 bp paired-end library according to the manufacturer’s instructions (TruSeq DNA PCR-Free LT Library Preparation Kits A and B, Illumina, Inc., San Diego, California USA). After construction, indexed libraries were pooled in groups of four to eight, and sequenced with a massively parallel sequencing machine and high-output kits (NextSeq500, two by 150 paired-end reads, Illumina Inc.). After sequencing, the raw reads were filtered to remove adaptor sequences, contaminating dimer sequences, and low quality reads. Pooled libraries with compatible indexes were repeatedly sequenced until 40 GB of data with greater than Q20 quality, was collected for each sire. In preliminary trials, 40 GB of Q20 data consistently resulted in greater than 10-fold read coverage for each animal. Previous results showed that this level of coverage provided scoring rates and accuracies that exceeded 99%.

The DNA sequence alignment process was similar to that previously reported. Briefly, FASTQ files corresponding to a minimum of 40 GB of Q20 sequence were aggregated for each animal. DNA sequences from FASTQ files were aligned individually to UMD3.1 with the BWA aln algorithm version 0.7.12, then merged and collated with bowtie sampe. The resulting sequence alignment map (SAM) files were converted to binary alignment map (BAM) files, and subsequently sorted via SAMtools version 1.3.1. Potential PCR duplicates were marked in the BAM files using the Genome Analysis Toolkit (GATK) version 3.6. Regions in the mapped dataset that would benefit from realignment due to small indels were identified with the GATK module RealignerTargetCreator, and realigned using the module IndelRealigner. The BAM files produced at each of these steps were indexed using SAMTools. The resulting indexed BAM files were made immediately available via the Intrepid Bioinformatics genome browser http://www.intrepidbio.com/ with groups of animals linked at the USMARC WGS browser http://www.ars.usda.gov/Services/docs.htm?docid=25585. The raw reads were deposited at NCBI BioProject PRJNA324822. Mapped datasets for each animal were individually genotyped with the GATK UnifiedGenotyper with arguments “--alleles” set to the VCF file (File S1), “--genotyping_mode” set to “GENOTYPE_GIVEN_ALLELES”, and “--output_mode” set to “EMIT_ALL_SITES”. Lastly, some SNP variants were identified manually by inspecting the target sequence with IGV software version 2.1.28 (described in the Methods section entitled ‘Identifying protein variants encoded by EPAS1’). In these cases, read depth, allele count, allele position in the read, and quality score were taken into account when the manual genotype determination was made.

**Evaluating WGS data integrity with 121 reference SNPs and 770 k bead array SNPs**

Genotypes from a set of 121 reference SNPs were used as an initial verification of the WGS datasets. Many of these DNA markers have been widely used for parentage determination, animal identification, and disease traceback (Table S1). The 121 reference SNPs were previously genotyped across the MBCDPv2.9 by multiple PCR-Sanger sequencing reactions, two independent designs of multiplexed matrix-assisted laser desorption/ionization time-of-flight mass spectrometry (MALDI-TOF MS) genotyping assays, and multiple bead array platforms, and are tabulated in Table S2. The error rate in the WGS data was estimated by comparing the consensus genotypes for these SNPs to the WGS genotypes. An animal’s WGS dataset passed initial verification when the accuracy of the WGS genotypes exceeded 97%, and the average mapped read depth was proportional to the amount of WGS data collected. Animals’ datasets that failed this initial verification were inspected closely for contaminating and/or missing files. Electronic file transfer errors resulted in contaminated and missing data for approximately one third of the 96 WGS datasets and required systematic testing, correction, and reprocessing. Linear regression analysis was accomplished in Excel version 2016. Access to the sequence via USDA internet site (http://www.ars.usda.gov/Services/docs.htm?docid=25585) and Intrepid Bioinformatics site (http://server1.intrepidbio.com/FeatureBrowser/customlist?record?listid=7686214634) was provided as soon as the .BAM files...
were produced. Because the raw datasets were available online as they were produced, the FASTQ files were deposited in the NCBI SRA only after they were validated as described above. These 96 sets of files can be accessed through BioProject PRJNA324822 in the Project Data table under the Resource Name: SRA Experiments. SNPs from the BovineHD BeadChip (Illumina Inc.) were selected for comparison because they were numerous, uniformly distributed across the bovine genome, and available. Based on the nucleotide sequence of the probes obtained from the manufacturer, the positions of the SNPs were verified via a BLAT process as previously described. A total of 772,990 variant positions were successfully mapped with this process, with 54 positions being discrepant when compared to those in the manufacturer’s most recent release of probe descriptions. The VCF file for these 772,990 variants is provided (File S1). The genotypes from the WGS data were compared to those from the high-density bead array with a custom program written specifically for this operation. Three classes of discordant genotypes were identified. First, those scored as homozygous in the WGS data and heterozygous in bead array data. These could have resulted from low coverage in WGS data at that position, or errors in the bead array caused by probes hybridizing to repeated sequences. The second type of discordance was scored as heterozygous in the WGS data, and homozygous in the bead array data. These could have resulted from allele-specific probe hybridization problems in the bead array platform. The final category consisted of missing genotypes in the bead array data, which were likely caused by errors in the conversion of the manufacturer’s “AB” genotype calls to the nucleotide calls.

Identifying protein variants encoded by bovine EPAS1
Using public internet access to USMARC sites, the nucleotide variation in the exon regions of EPAS1 was visualized with open source software installed on a laptop computer and recorded manually in a spreadsheet. Briefly, a Java Runtime Environment (Oracle Corporation, Redwood Shores, CA) was first installed on the computer. When links to the data were selected from the appropriate web page, IGV software automatically loaded from a third-party site (Intrepid Bioinformatics, Louisville KY) and the mapped reads were loaded in the context of the bovine UMD3.1 reference genome assembly. For viewing EPAS1 gene variants, WGS from a set of eight animals of different breeds was loaded (“mixed groups of 8”, http://www.ars.usda.gov/Research/docs.htm?docid=25586) and the IGV browser was directed to the appropriate genome region by entering “EPAS1” in the search field. The IGV zoom function was used to view the first exon at nucleotide resolution with the “Show translation” option selected in IGV. The exon sequences were visually scanned for polymorphisms that would alter amino acid sequences, such as missense, nonsense, frameshift, and splice site mutations. Once identified, the nucleotide position corresponding to a protein variant was viewed and recorded for all 96 animals. Using IGV, codon tables, and knowledge of the HIF2A protein sequence (NP_777150), the codons affected by nucleotide alleles were translated into their corresponding amino acids and their positions noted. Haplo-type-phased protein variants were assigned unambiguously in individuals that were homozygous, and those individuals with only one variant amino acid. A maximum parsimony phylogenetic tree was constructed manually from the unambiguously phased protein variants and used to infer phase in any remaining variants with simple maximum parsimony assumptions.

WGS datasets from five closely-related Bovinae species were mapped to the cattle reference assembly UMD3.1 with a process similar to that previously reported. These mapped Bovinae samples included two each of yak, gaur, and banteng; and one sample each of plains bison, water buffalo. The mapped genomes were visually inspected across the EPAS1 exons in the same browser environment as the cattle data, and variant codons were recorded. Information about the source and the content of the WGS datasets is provided in Table S3. Because reference SNP genotypes are not readily available for these species, verification of the integrity and quality of the newly sequenced Bovinae WGS datasets was limited. For each dataset, the mapped read density in conserved exons was estimated and compared to the amount of Q20 sequence collected for that animal. No inconsistencies were noted between the expected and observed read depths. In addition, distinctive homozygous “species-specific” nucleotides were observed for each species, and these same nucleotides were not observed in the other species. The genomes for all eight animals were made viewable by IGV at http://www.ars.usda.gov/Services/Docs.htm?docid=25585. They are also available at NCBI BioProjects: PRJNA325061, PRJNA221623, and PRJNA207334.

MALDI-TOF MS genotyping of six EPAS1 missense mutations
A single multiplex assay was designed for the six EPAS1 missense SNPs with the information in Table 1 with software provided by the manufacturer (Agena Biosciences, San Diego, California, USA). The oligonucleotide sequences and assay conditions are provided in Table S4. After design and validation with bovine control DNAs for each SNP, the MBCDPv2.9 DNA was tested in a blinded experiment in which the true genotypes were unknown by those typing the samples. Assay design and genotyping was performed by GeneSeek (Lincoln, Nebraska, USA) with the MassARRAY platform and iPLEX Gold chemistry according to the manufacturer’s instructions (Agena Biosciences). MALDI-TOF MS genotypes for six SNPs are provided for the MBCDPv2.9 and 1154 of 1168 cattle from 46 breeds in Table S5.

Results
Panel design, genome sequencing, and quality control of WGS datasets
A beef cattle diversity panel was designed to broadly sample the genetic diversity of U.S. populations, while fitting within the constraints of a 96-sample format, often used for automated DNA sequencing and genotyping. The composition and design of the panel was updated from a previously reported set as described in the Methods. A minimum of four sires were included for each breed, with the more popular U.S. breeds having five or six animals (Figure 1). There was relatively little power for detecting rare variants within breed, since not more than 12 haploid genomes were sampled (95% probability of detecting any polymorphism with a frequency greater than 0.22, Methods). Despite the modest power within breed, sequencing the entire panel significantly increased the chances of detecting relatively rare variants segregating in U.S. beef cattle. With more than 187 of 192 unshared haploid genomes in the 96 sires, it was estimated there was a
95% probability of observing polymorphisms with a frequency greater than 0.016. Thus, the power for allele detection in this beef diversity panel was derived from having exceedingly few pedigree relationships within breed, and essentially none between breeds.

The WGS was generated by sequencing indexed pools of libraries whose composition was adjusted iteratively across multiple instrument runs to achieve at least 40 GB of FASTQ sequence. The average amount of total sequence per sample was 48.3 GB (±12.0) and varied between 40.2 GB and 109.4 GB. This approach reduced the overall data production cost, however each animal had data files from multiple sequencing runs that required manual collation prior to analysis, and thus increased the labor cost. In addition to the usual challenges of sample contamination, sample switches, missing data, variable quality data, and data transfer errors, the FASTQ files produced by the instrument had identical names across multiple machine runs. This added another layer of complexity to maintaining file provenance. The process of manually aggregating and transferring an average of 42 similarly-named FASTQ files for each animal was inherently prone to error and unavoidable with the instrument and the institutional network security restrictions.

Thus, to verify the WGS data integrity at the end of the process, genotypes from a set of 121 reference SNPs were used as a first test. These SNPs are distributed across the genome, highly-informative in U.S. beef cattle, have been widely used for bovine parentage testing (Methods). The WGS-derived genotypes for these 121 SNPs were obtained by viewing an animal’s mapped reads at the relevant genome coordinates, with public software, a third party database, and web links created for this task (illustrated in Figure 2A, http://www.ars.usda.gov/Research/docs.htm?docid=25586). As described in the Methods, data inconsistencies of multiple types were discovered by comparison with the known reference genotypes and corrected in approximately one third of the file sets. Comparison to the reference SNP genotypes also provided a check for the expected linear relationship between the amount of sequence collected and the depth of reads mapped to the reference assembly. Regression analysis showed that the average read depth at the 121 reference SNPs was directly proportional the amount of sequence collected (Figure 2B). The 48.3 GB of sequence collected for each animal resulted in an average of 14.4-fold depth of mapped read coverage. The overall accuracy of WGS genotypes for the 121 reference SNPs was 99.5%, with 56 sires having 100% concordance (Figure 2C). The few WGS genotype errors observed were typically caused by undetected heterozygous alleles at sites with low read coverage. Thus, the use of 121 reference SNPs was effective for discovering and repairing errors in these WGS datasets, and verifying the coverage.

A broader characterization of the coverage and quality of each dataset was accomplished by comparing an average of 730,410 of SNP genotypes from each sire to those from a high density bead array (Methods). The average distribution of read depths was slightly positively skewed with a mode of 12.5 when combined for all animals (Figure 3A). The average read depth for these 730 k SNPs varied between 40.2 GB and 109.4 GB. This approach reduced the overall data production cost, however each animal had data files from multiple sequencing runs that required manual collation prior to analysis, and thus increased the labor cost. In addition to the usual challenges of sample contamination, sample switches, missing data, variable quality data, and data transfer errors, the FASTQ files produced by the instrument had identical names across multiple machine runs. This added another layer of complexity to maintaining file provenance. The process of manually aggregating and transferring an average of 42 similarly-named FASTQ files for each animal was inherently prone to error and unavoidable with the instrument and the institutional network security restrictions.

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Thus, to verify the WGS data integrity at the end of the process, genotypes from a set of 121 reference SNPs were used as a first test. These SNPs are distributed across the genome, highly-informative in U.S. beef cattle, have been widely used for bovine parentage testing (Methods). The WGS-derived genotypes for these 121 SNPs were obtained by viewing an animal’s mapped reads at the relevant genome coordinates, with public software, a third party database, and web links created for this task (illustrated in Figure 2A, http://www.ars.usda.gov/Research/docs.htm?docid=25586). As described in the Methods, data inconsistencies of multiple types were discovered by comparison with the known reference genotypes and corrected in approximately one third of the file sets. Comparison to the reference SNP genotypes also provided a check for the expected linear relationship between the amount of sequence collected and the depth of reads mapped to the reference assembly. Regression analysis showed that the average read depth at the 121 reference SNPs was directly proportional the amount of sequence collected (Figure 2B). The 48.3 GB of sequence collected for each animal resulted in an average of 14.4-fold depth of mapped read coverage. The overall accuracy of WGS genotypes for the 121 reference SNPs was 99.5%, with 56 sires having 100% concordance (Figure 2C). The few WGS genotype errors observed were typically caused by undetected heterozygous alleles at sites with low read coverage. Thus, the use of 121 reference SNPs was effective for discovering and repairing errors in these WGS datasets, and verifying the coverage.

A broader characterization of the coverage and quality of each dataset was accomplished by comparing an average of 730,410 of SNP genotypes from each sire to those from a high density bead array (Methods). The average distribution of read depths was slightly positively skewed with a mode of 12.5 when combined for all animals (Figure 3A). The average read depth for these 730 k SNPs varied between 40.2 GB and 109.4 GB. This approach reduced the overall data production cost, however each animal had data files from multiple sequencing runs that required manual collation prior to analysis, and thus increased the labor cost. In addition to the usual challenges of sample contamination, sample switches, missing data, variable quality data, and data transfer errors, the FASTQ files produced by the instrument had identical names across multiple machine runs. This added another layer of complexity to maintaining file provenance. The process of manually aggregating and transferring an average of 42 similarly-named FASTQ files for each animal was inherently prone to error and unavoidable with the instrument and the institutional network security restrictions.
which had a concordance of 91.8% (Figure 3B). However, the 121 reference SNP genotypes for this same animal were 98.4% accurate (119/121). This result suggests that the lower genotype concordance in the Corriente sire may have been caused by the quality of the bead array data. For all other animals, the discordant genotypes were infrequent, with “allele dropouts” being the most common type (Methods). Allele dropouts were inferred at a SNP site when one allele of a heterozygote was not detected (i.e., “dropped”). Although rare, there were more dropped alleles observed in the bead array data (1.1%) than for the WGS data (0.7%). Taken together, the analyses indicate that the WGS datasets from these 96 diverse beef sires are of sufficient quality and coverage for use in identifying and decoding gene variants in U.S. beef cattle.

Identification of protein variants encoded by **EPAS1**

The 96 sets of aligned WGS data were visually analyzed in the **EPAS1** coding region to identify potential HIF2A protein variants (Methods). **EPAS1** consists of 16 exons spanning 90 kb of genomic DNA and encodes an 870 amino acid protein with multiple functional domains (Figure 4A and Figure 4B). Viewing the aligned sequences and detecting variants was simple, fast, and accurate with the IGV software and a browser developed for this purpose (Figure S1). Four previously undescribed missense mutations were discovered and predicted to cause the substitution of glutamine (Q) for glutamate (E) at position 270; leucine (L) for proline (P) at position 362; glycine (G) for Alanine (A) at position 671; and phenylalanine (F) for leucine (L) at position 701 (Table 1 and Figure 4B). The two additional amino acid variants previously associated with PH, were also observed (A606T and G610S). No other
Figure 4. Physical maps and unrooted maximum parsimony phylogenetic trees of HIF2A protein variants found in cattle. Panel A, genomic DNA map of EPAS1: blue arrows, exon regions; grey horizontal lines, intron regions. Panel B, map of HIF2A domains in relationship to missense mutations found in cattle: bHLH, basic helix-loop-helix domain; PAS-A and PAS-B, Per-Arnt-Sim domains; ODDD, oxygen-dependent degradation domain; N-TAD, N-terminal transactivation domain; C-TAD, C-terminal transactivation domain. Panels C and D represent results from the 96-member, 19 breed diversity panel and the 1154-member, 46 breed set, respectively. The most frequent HIF2A isoform (“variant 1”) was used as the reference sequence for the trees. For “variants 1” through “8”, each node in the tree represents a different isoform of HIF2A that varies by one amino acid compared to adjacent nodes. The areas of the circles are proportional to the variant frequency in the group of cattle tested. “Variant 3” (pink circle; T606, S610) is identical to that associated with PH in Angus cattle. “Variant 2” (Q270) is identical to the 870 amino acid protein encoded by the bovine reference assembly UMD3.1.

missense, nonsense, frameshift, splice site, or indel variants affecting the coding region were detected. Haplotypes encoding seven predicted HIF2A variants were translated and placed in the context of a phylogenetic tree (Figure 4C). Five of seven predicted HIF2A protein variants (variants “2”, “4”, “5”, “6”, and “7”) were previously unreported, and accounted for 17% of the total in the beef cattle diversity panel.

To verify the accuracy of EPAS1 genotypes and determine the protein variant frequencies in a larger set of U.S. cattle, MALDI-TOF MS assays were developed for the six missense SNPs (Methods). In a blinded test, 575 of 576 (99.8%) EPAS1 MALDI-TOF MS genotypes from the 96 sires were concordant with those from WGS, confirming that the newly discovered SNPs were authentic and the WGS and MALDI-TOF MS genetic tests were accurate. The average HIF2A variant frequencies in a set of 1154 purebred cattle from 46 breeds were similar to those observed in the beef cattle diversity panel (Table 2 and Figure 4D) with a call rate of 98.8%.

The HIF2A isoform associated with an increased risk for PH in Angus cattle (T606, S610; “variant 3”) was observed in 18 of 46 breeds, with four breeds having frequencies higher than Angus (Table 3). The Guernsey dairy breed had the highest

<table>
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<tr>
<th>Breed</th>
<th>Variant 3 Frequency</th>
</tr>
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<tbody>
<tr>
<td>Guernsey</td>
<td>0.63</td>
</tr>
<tr>
<td>Angus</td>
<td>0.23</td>
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...
Table 2. Frequencies of predicted HIF2A protein variants in U.S. cattle

<table>
<thead>
<tr>
<th>Protein variant code&lt;sup&gt;a&lt;/sup&gt;</th>
<th>Variant amino acids&lt;sup&gt;b&lt;/sup&gt;</th>
<th>Protein variant frequency&lt;sup&gt;c&lt;/sup&gt;</th>
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<tbody>
<tr>
<td></td>
<td></td>
<td>Beef cattle diversity panel (n = 96)</td>
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<tr>
<td>1</td>
<td>E270, P362, A606, G610, A671, L701</td>
<td>0.760</td>
</tr>
<tr>
<td>2</td>
<td>Q270, P362, A606, G610, A671, L701</td>
<td>0.042</td>
</tr>
<tr>
<td>3</td>
<td>E270, P362, T606, S610, A671, L701</td>
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</tr>
<tr>
<td>4</td>
<td>E270, L362, A606, G610, A671, L701</td>
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</tr>
<tr>
<td>5</td>
<td>Q270, L362, A606, G610, A671, L701</td>
<td>0.047</td>
</tr>
<tr>
<td>6</td>
<td>E270, P362, A606, G610, G671, L701</td>
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<tr>
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<tr>
<td>8</td>
<td>E270, P362, T606, G610, A671, L701</td>
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</tr>
</tbody>
</table>

<sup>a</sup>HIF2A protein variant allele definitions are shown in Figure 4.

<sup>b</sup>The bolded residues are those differing from “variant 1”.

<sup>c</sup>The coefficient of determination for these frequencies (r<sup>2</sup>) was 99.9

Table 3. The frequencies of predicted HIF2A protein variants in 46 U.S. breeds.

<table>
<thead>
<tr>
<th>Breed group</th>
<th>Animals typed</th>
<th>HIF2A protein variant allele frequency&lt;sup&gt;c&lt;/sup&gt;</th>
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<tr>
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<tr>
<td>Angus</td>
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<td>Maine-Anjou</td>
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<td>Marchigiana</td>
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</tr>
<tr>
<td>Murray Gray</td>
<td>21</td>
<td>0.67</td>
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</table>
Alignment of cattle HIF2A sequences with 70 available species of the Gnathostomata superclass showed that a third of the residues (288 of 870) were perfectly conserved throughout (Table S6). Of the six HIF2A variant sites identified in cattle, the most conserved residue was glutamate at the E270Q site, which was present in all 70 Gnathostomata tested, 37 of which are shown in Figure 6. The leucine residue of the L701F variant site was less conserved, but still present throughout the Amniota, with the phenylalanine variant being present in the Tetrapoda and higher. The glycine residue at the G610S variant site was conserved in the Laurasiatheria, with the notable exception of S610 in swine, a species known for a marked pulmonary vasoconstrictive response to hypoxia. The proline residue of the P362L variant site was conserved through Cetartiodactyla with leucine present in Perissodactyla and higher. The alanine residue of the A671G variant site was conserved in the Bovidae with threonine and other residues present in Cetartiodactyla and higher. Variant A606T was the least conserved of all the variant sites with the alanine residue only conserved in the Bovinae, and the threonine residue present in other ruminants with isoleucine present in Cetartiodactyla and higher. Based exclusively on the degree of conservation across vertebrate species, the predicted ranking for potentially deleterious EPAS1 missense mutations in cattle was: E270Q > L701F > G610S > P362L > A671G > A606T. However, the actual impact of these polymorphisms on cattle is dependent on additional factors, some of which are discussed in the next section.

**Discussion**

Our primary goals were to create a searchable and publicly viewable genomics resource consisting of 96 sires representing a broad cross section of U.S. beef cattle, and demonstrate proportion of the risk allele with 18 of 26 animals (69%) having one or two copies of “variant 3” (Table S5). Notably, all 96 animals from the Bos indicus breeds (Brahman, Nelore, Indu-Brazil, and mini-zebu) were homozygous for the most common HIF2A “variant 1”. An important result of typing the extended 46 breed set of cattle, was the discovery of an unlinked T606 mutation (“variant 8”, Figure 4D) present in Romagnola, Chianina, and Maine-Anjou cattle (Table S5). The discovery of an eighth variant brought the number of possible HIF2A diploid combinations to 36, and underscored the importance of accurate HIF2A typing in animals used to study PH and RHF in beef cattle.

To determine the most likely phylogenetic root of the HIF2A tree, and thus establish a possible order of mutational events, HIF2A sequences were analyzed in eight individuals from closely related species: from the Bos, Bison, and Bubalus genera. HIF2A “variant 1” was the likely ancestral root, based on its similarity to HIF2A from the most closely related species (Figure 5). Thus, the S610 mutation likely occurred on the T606 haplotype and is the more recent mutation of the two. Identifying breeds and individuals that have the HIF2A T606 allele provides the opportunity for future comparisons of the relative effects of T606 alone (“variant 8”), or in combination with S610 (‘variant 3’). The highly conserved HIF2A amino acid residues across vertebrates provides insight into the potential impact of missense mutations in cattle, because invariant residues tend to be critical for protein function. The 870 amino acid sequence of cattle HIF2A is highly similar to those from sheep, whale, human, mouse, and alligator (97, 90, 88, 83, and 73% identity, respectively).
its use for identifying missense mutations in \textit{EPAS1}, a bovine gene associated with PH and RHF\textsuperscript{35}. To that end, we first determined the amount of WGS required to achieve accurate diploid genotyping when aligned to the bovine reference genome. Our findings verified that 40 GB of short read, paired-end DNA sequence data from the NextSeq500 instrument, provided an average of 12-fold genome coverage. Accordingly, we used 40 GB as a minimum for each animal, aligned the sequences, and made them available online for downloading or viewing with a custom visualization tool that supports accurate assessment of genetic variation. The average coverage of the beef cattle diversity panel was 14.8-fold and resulted in an average genotype accuracy of approximately 99\%. These numbers were consistent with results obtained with 379 Canadian beef and dairy cattle, 65 Swiss dairy cattle, and 96 human WGS datasets, sampled at various coverage levels, and compared to bead array data\textsuperscript{10,12,30}.

Surprisingly, the amount of effort required to verify the identity and accuracy of the datasets was nearly equal to that required to produce it. Comparing WGS genotypes from 121, well-distributed, highly-informative reference SNPs was sufficient to identify and correct errors in the datasets. However, larger SNP datasets from commercial DNA microarrays provided an additional level of confidence that was useful. Considering the time and resources invested in collecting the WGS, it seemed prudent to have a significant set of independently obtained reference SNP genotypes available for each animal in the group, and use them to validate each WGS dataset.

An important secondary goal of this project was to provide biomedical veterinary researchers the ability to readily inspect gene mutations reported in one breed, evaluate them in other breeds, and search for any additional mutations that may affect protein structure. A web-based platform was created to achieve this goal, and then tested by analyzing \textit{EPAS1}, a gene where missense mutations had been previously documented. Six missense mutations, including the two that were previously reported to be associated with PH\textsuperscript{35}, were readily identified by viewing the aligned raw sequence. The variants were validated by designing MALDI-TOF MS genetic tests and genotyping a broader population of animals. Determining the haplotype phase of these mutations in a total of 1250 individuals from 46 breeds, resulted in eight predicted HIF2A protein variants, five of which were previously unknown. Comparing HIF2A variant frequencies in the 96-member, 19 breed diversity panel with those of the 1154-member, 46 breed set confirmed that the smaller panel was a good approximation of the larger one ($r^2 = 99.9\%$). Together, the five newly identified HIF2A variants comprised 16.4\% of the total alleles in 46 breeds, and represented a significant proportion of the HIF2A variation in cattle. The MALDI-TOF MS genetic test for \textit{EPAS1} was designed to facilitate future study of these variants, and provide a way to control for \textit{EPAS1} stratification in studies of bovine respiratory diseases. The ability to identify the full range of protein variants in a population is critical for designing studies intended to test the candidate gene’s influence on a trait.

The report by Newman \textit{et al.}\textsuperscript{35}, describing an \textit{EPAS1} T606, S610 gene variant associated with PH, raises intriguing questions about the biological mechanisms leading to disease. Among them is the possibility that only one of the two linked missense mutations is pathogenic. Our findings suggest that the S610 variant would be the more plausible candidate of the two. Five lines of evidence are consistent with this hypothesis. First, the S610 missense mutation appears to have arisen on an existing \textit{EPAS1} T606 haplotype, making it the more recent mutation. Younger mutations in functional genes are more likely to be deleterious that older ones\textsuperscript{33}. Second, the G610 residue is conserved across ruminant species, which is consistent with the removal of deleterious alleles by strong purifying selection. Third, among the Laurasiatheria representatives, swine is the only one besides cattle to have the S610 allele. Swine also develop severe PH when exposed to chronic hypoxia\textsuperscript{34}. Fourth, the T606 residue occurs in sheep and goats which have only mild responses to chronic hypoxia\textsuperscript{35,36}. Fifth, the T606 residue is also present in Tibetan Antelope, a species that evolved

\begin{figure}
\centering
\includegraphics[width=\textwidth]{figure5.png}
\caption{	extbf{Rooted maximum parsimony phylogenetic tree of HIF2A protein variants found in cattle and closely related species.} The cattle HIF2A “variant 1” was used as the reference sequence for comparison with HIF2A from five other species (Methods). Cattle HIF2A residues were highly conserved between these species and only differed at 11 total sites. In “variant 1” the cattle residues at these 11 positions were: V442, K613, T663, L644, M649, R656, M661, L668, F678, V693, H733. For cattle “variants 1” through “8”, the areas of the circles shown are proportional to the variant frequency in the group of 1250 cattle tested. “Variant 3” (pink circle; T606, S610) is identical to that associated with PH in Angus cattle\textsuperscript{10}. The nodes derived from analysis of other species are indicated with a black filled circles and do not represent frequency information.}
\end{figure}
Figure 6. Evolutionary comparison of HIF2A residues from six variant sites. HIF2A sequences from a representative subset of 37 species from the Gnathostomata superclass were deduced from WGS or downloaded from GenBank. Abbreviations and symbols are as follows: TMRCA, estimated time to most recent common ancestor in millions of years; letters, IUPAC/IUBMB codes for amino acids; dot, amino acid residues identical to those in cattle HIF2A “variant 1”; dash, polypeptide region missing in shark HIF2A.

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<th>Taxonomic group</th>
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at high-altitude and does not suffer from PH. These observations are consistent with the hypothesis that the S610 missense mutation is pathogenic and causes PH in cattle. Alternatively, both alleles may be required in combination to cause disease, or they may be in genetic linkage with an undiscovered cause of PH. Regardless, the pathogenic mutation hypothesis is testable with appropriate individuals from Chianina and Maine-Anjou cattle breeds, since examples of HIF2A T606, S610 (“variant 3”) and HIF2A T606 ("variant 8") are present in both breeds.

Of the remaining EPAS1 missense mutations, E270Q stands out as having the greatest potential for affecting the function of HIF2A. This prediction is based on the observation that the E270 residue was invariant across all 70 vertebrate species evaluated. The E270Q variant is located in the PAS-B domain, the second of two tandemly positioned PAS domains. Mutations in murine PAS-B have been shown to affect the ability of HIF2A to sense chemical signals via ligand binding, and thus stabilize the transcriptionally active heterodimer in response to hypoxia, although the E270 residue was not specifically tested\(^\text{3}\). EPAS1 haplotypes encoding Q270 residues accounted for 22% of the total in 80 Holstein sires used in the present study, and could be a potential cause of the PH observed Holstein cattle\(^\text{9}\). These 80 Holstein cattle were also devoid of the EPAS1 T606, S610 haplotype associated with PH in Angus cattle. Overall EPAS1 Q270 alleles are present at a frequency of 8% in U.S. cattle tested (HIF2A variants “2”, “5”, and “7”, Figure 4) and are predicted to have deleterious biological consequences.

Another highly conserved residue was L701, which was invariant through the Amniota. However, the F701 substitution was present only on the Q270 haplotype (“variant 7”) and is located in an interdomain region of HIF2A of unknown function. Moreover, the F701 substitution was only observed in one of 1250 animals tested (Salers sire no. 19999882 in MBCDPv2.9). The WGS for this animal at position chr11:28663897, together with concordant genotypes by MALDI-TOF MS confirmed the authenticity of this SNP. However, its exceedingly low frequency makes it unlikely that this substitution, no matter how disruptive, would have a significant impact on U.S. cattle. The remaining missense mutations P362L and A671G were neither highly conserved in vertebrates, nor located in HIF2A regions of known significance. However, their combined frequency in cattle was not insignificant at nearly 8%. Either of these could potentially affect HIF2A function and thereby influence traits associated with EPAS1. The substitution of proline for leucine is a particularly significant change that has been shown to cause functional disruptions in other proteins such as T4 lysozyme and caspase-9\(^\text{10-14}\). However, the impact of this substitution remains unknown.

After more than 20 years of selection for herd sires with low pulmonary artery pressures (PAP), 50% of calf mortalities were still attributed to PH in some high-altitude ranches\(^\text{15}\). We hypothesize that selection for the most common, ancestral HIF2A sequence will enrich for the most favorable allele for U.S. beef cattle and complement efforts that employ PAP testing. This HIF2A sequence contains residues E270, P362, A606, G610, A671, and L701 (“variant 1", Figure 4) and has the highest amino acid sequence identity when compared to HIF2A from yak, gaur, banteng, and bison. The frequency of “variant 1" was 100% in the Brahman, Brown Swiss, Indu Brazil, Nelore, and mini-zebu breeds; greater than 90% in Ankole-Watusi, Beefmaster, Brahmosine, Devon, Marchigiana, Piedmontese, Romagnola, Simmental, Tuli, Wagyu; and 78% overall. Thus, if employed, selection for HIF2A “variant 1” would not be an exercise in introgression, but rather an effort to remove a minority of potentially deleterious alleles. This may help reduce the overall incidence of PH and the problems associated with it in cattle.

**Conclusion**

In summary, the WGS resources described here are suitable for use in identifying and decoding gene variants in the vast majority of U.S. beef cattle. When applied to EPAS1, the findings suggest that there may be deleterious alleles circulating in U.S. in addition to those previously associated with high altitude PH. These resources, including the web interface, underlying sequence data, genetic tests, and the associated information are available to researchers, companies, veterinarians, and producers for use without restriction.

**Data availability**

Validated cattle FASTQ files are available in the NCBI SRA under accession numbers SRR4001609-SRR4002095; SRR4004613-SRR4004644; SRR4002950-SRR4003067; SRR4003069-SRR4003073; SRR4003075-SRR4003079; SRR4003081-SRR4003085; SRR4003087-SRR4003094; SRR4003096-SRR4003139; SRR4003141-SRR4003146; SRR4003148-SRR4003152; SRR4003154-SRR4003158; SRR4003160-SRR4003164; SRR4003166-SRR4003170; SRR4003172-SRR4003177; SRR4003179-SRR4003182; SRR4003184-SRR4003188; SRR4003190-SRR4003451; SRR4004645-SRR4004679; SRR4004680-SRR4004734; SRR4004736-SRR4004891; SRR4004893-SRR4004920; SRR4004922-SRR4004948; SRR4004950-SRR4004982; SRR4004991-SRR4004992; SRR4004994-SRR4004997; SRR4005006-SRR4005012; SRR4005021-SRR4005026; SRR4005044-SRR4005048; SRR4005057-SRR4005062; SRR4005071-SRR4005195. The data have also been deposited with links to BioProject accession number PRJNA324822 in the NCBI BioProject database (https://www.ncbi.nlm.nih.gov/bioproject/).

FASTQ files produced from closely related species are available in the NCBI SRA under accession numbers SRR4035250-SRR4035309 and are associated with BioProject accession number PRJNA325061.

In addition, access to the aligned sequences is available via USDA internet site: http://www.ars.usda.gov/Services/Docs.htm?docid=25585. Download access to the .BAM files is available at the Intrepid Bioinformatics site: http://server1.intrepidbio.com/FeatureBrowser/customlist/record?listid=7686214634.
Author contributions
Conceived and designed the experiments: MPH, TPLS, TSK.
Performed the experiments: BS, JKC, JQ, MPH, TPLS, TSK, VB.
Analyzed the data: BS, JKC, JQ, MPH, TPLS, TSK, VB.
Contributed reagents/materials/analysis tools: BS, JKC, JQ, MPH, TPLS, TSK, VB.
Wrote the paper: MPH, TPLS, TSK.

Competing interests
The authors declare the following interests: co-authors V.B., J.Q, and B.S. are full-time employees at GeneSeek, a Neogen company that provides agrigenomic and veterinary diagnostic services. T.S.K. is the CEO of Intrepid Bioinformatics, a company that provides web-based systems to privately store, analyze, curate, share, and remotely access genetic data. There are no patents, products in development, or marketed products to declare. These interests do not alter the authors’ adherence to all the journal’s policies on sharing data and materials.

Grant information
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The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Acknowledgments
We thank R. Godtel and K. McClure for outstanding technical assistance, and J. Watts for secretarial support. We thank Drs. D. Grotelueschen and P. Hackett for thoughtful discussions and improvements to the manuscript; Dr. L. Cundiff for advice in designing the MARC beef cattle diversity panel, and Dr. W. Laegreid for supporting it; Dr. D. Coover (SEK Genetics, Inc., Galesburg, KS), D. Frank (ABS Global, Inc., DeForest, WI), B. House (Select Sires, Inc., Plain City, OH), and C. Rugg (Bovine Elite, Inc., College Station, TX.) D. Trimmer (Accelerated Genetics, Baraboo, WI), The Braunvieh Assoc. of America, J. Spawn (North American Tarentaise Assoc., and the American Tarentaise Assoc.) for assistance in sire selection; Drs. C. Chase (Senepol), H. Neibergs (Wagyu), and T. Sonstegard and C. Van Tassell (Holstein) for providing samples from additional breeds; D. Davis (Cattlemen’s Texas Longhorn Registry) for providing an alternate set of Texas Longhorn samples; Drs. D. Armstrong and E. Louis from Omaha’s Henry Doorly Zoo for providing gaur blood samples, Drs. L. Chemnick and O. Ryder from the San Diego Zoo’s Beckman Center for Conservation Research for providing banteng DNA samples; Mr. M. Swartz from the International Yak Association for providing a frozen hide sample of Queen Allante D171, after she died of natural causes; Mr. M. Bredthauer from the Straight Arrow Bison ranch for providing a liver sample from an animal processed for food; and Mr. R. Goode (Goode Cattle Co. Damon, TX) for his generous and dedicated assistance in collecting hair samples from rare U.S. breeds. Mention of trade names or commercial products in this publication is solely for the purpose of providing specific information and does not imply recommendation or endorsement by the USDA. The USDA is an equal opportunity provider and employer.

Supplementary material
Table S1. Features of 121 USMARC bovine parentage SNPs.
Click here to access the data
Table S2. Summary WGS statistics and reference genotypes for 121 parentage SNPs with the USMARC Beef Diversity Panel Version 2.9
Click here to access the data
Table S3. Source information for individuals from Bovinae species mapped to the cattle UMD3.1 reference assembly.
Click here to access the data
Table S4. MALDI-TOF MS assay design for six missense mutations in bovine EPAS1.
Click here to access the data
Table S5. HIF2A protein variant diplotypes in 1154 animals from 46 breeds of U.S. cattle.
Click here to access the data
Table S6. Alignment of HIF2A sequences from Gnathostomata species.
Click here to access the data
References

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Publisher Full Text | Free Full Text

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PubMed Abstract | Free Full Text
Open Peer Review

Current Referee Status: ✔️ ✔️

Version 1

Referee Report 23 September 2016
doi:10.5256/f1000research.9962.r16383

Matthew C. McClure
Irish Cattle Breeding Federation, Bandon, Ireland

Overall a very nice paper on EPAS1 and high-altitude pulmonary hypertension (HAPH). Only time will tell if the more important outcome of this paper is the determination that the G610S alleles likely the causative mutation for HAPH (over A606T), that multiple EPAS1 alleles could together cause/impact some version of HAPH, or that there is finally a public database to compare whole genome sequence data between studies. As cattle producers and breed associations move to use more genetic tests to minimize genetic variants that cause economic loss in the breed the first 2 points above will be very useful, especially with “50% of calf mortalities… attributed to PH in some high-altitude ranches”, as noted in the manuscript. For researchers studying other genetic diseases the latter point will be more important as such a public database is needed to help filter out non-causative alleles. Currently such researchers have had to turn to other scientists that have their own WGS databases, or to semi-private database like the 1000 bull dataset where you must be a member to access the data. In the past I personally have had to use the former option and was grateful that the bovine community is full of gracious scientists that would help.

This is a very well written paper and was a pleasure to review.

Overall I do approve of the manuscript to be indexed and advise only minor edits and some suggestions listed below.

Suggestions:
1. In the abstract you state that the “existing bovine WGS databases… diversity in US beef cattle”. Suggest this be changed to global beef cattle or similar as this issue is not confined to the USA.

2. Page 3, 4th paragraph: While it's a semi-private WGS dataset, given its size, I do recommend including the 1000 Bulls project when talking about population-scale WGS data. You could mention that it’s there but not publicly available which is one of the main advantages your WGS data has as reported in the manuscript

Recommended:
1. Introduction: You state that “114 Mendelian traits with known causative mutations” and reference a 2014 paper. I advise updating that number as OMIA lists (23/9/2016) that 130 Mendelian disorders have their causative mutation known in cattle. I’m fine if you keep the reference you have and include a OMIA reference as your reference is an excellent review paper.
2. Table 1: You present a large amount of data about each variant and this is wonderful as so many publications would only list one or two of your columns. I would like to see the flanking sequence be increased from 5bp per side to 15bp. Ensemble returns 0 hits for all 6 of your flanking sequences with BLASTing against the UMD3.1 assembly. Increasing it to my recommended length allows BLAST to find unique placements of the sequence. For future researchers to be able to test for disease causing variants having minimal information listed in publication makes it a challenge to make sure you are correctly designing diagnostic probes. I speak from experience in designing diagnostic probes for the IDB cattle chip used in Ireland.

3. Table 3 appears to be only a subset of Table 4 with no additional information. Recommend removing Table 3 and only using Table 4. If you have a justifiable reason for keeping Table 3 than advise listing the breeds in alphabetical order and not in variant 3 frequency order.

4. Table 4. In this you show no animals having the variant 7 allele, yet in the 5th paragraph on page 14 you state that "the F701 substitution was only observed in one of the 1250 animals tested (Salers,...)". Please include the frequency of this allele in the Salers row in Table 4.

5. Figure 5: Why is node '3' in grey? It's not stated in the Figure.

6. For multiple Supplementary tables: Advise including this information from Table 1: cIUPAC/IUBMB ambiguity codes used for nucleotides:...... Or stating that IUPAC ambiguity codes are used, right now the reader just have to figure it out that a S means the individual is G/C for that SNP.

7. Table S1: Don't see why columns C, G, or H are needed: Distance between markers (Mb), Nearest gene (UMD3.1), Distance to nearest gene (kbp), respectively. The rest of the data in the table is excellent and wish more papers provided SNP data at this level. Really like that the ambiguous SNP in the flanking sequence are shown in bold and capital. I recommend removing column C, G, and H as they are meaningless as to information on Parentage SNP.

8. Table S2: You need to note that a “A, C, T, or G” shown as the genotype indicates that this is a homozygous SNP. One can figure it out since you also show the IUPAC ambiguous letters for heterozygous genotypes, but it would be better to state.

9. Table S5: Column E is labelled as “EPAS1 protein variant diplotype”. In Table 1, 2, 3 you use the phrase “HIF2A protein variant”. I advise using the same here, otherwise you can easily confuse the reader. Especially as HIF2A is the name of the protein coded by the EPAS1 gene, so if you are referring to protein variants HIF2A is more appropriate than EPAS1.

**Competing Interests:** No competing interests were disclosed.

I have read this submission. I believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

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Author Response 28 Sep 2016

*Michael Heaton*, US Meat Animal Research Center, USA

We have provided point-by-point responses to the reviewer's suggestions and recommendations below:
Reviewer suggestion: “In the abstract you state that the ‘existing bovine WGS databases… diversity in US beef cattle’. Suggest this be changed to global beef cattle or similar as this issue is not confined to the USA.”
Authors response: We agree and have made this change.

Reviewer suggestion: “Page 3, 4th paragraph: While it’s a semi-private WGS dataset, given its size, I do recommend including the 1000 Bulls project when talking about population-scale WGS data. You could mention that it’s there but not publicly available which is one of the main advantages your WGS data has as reported in the manuscript”
Authors response: We agree. Please note that the first reference provided in this paragraph (Daetwyler et al. Nat Genet. 2014; 46(8): 858–65) is the current published reference for the 1000 bulls project. Their sequence is public and available at http://www.ncbi.nlm.nih.gov/bioproject/PRJNA238491. In addition, the WGS from our article is available for inclusion in subsequent rounds of the 1000 bulls project.

Reviewer recommendation: “Introduction: You state that “114 Mendelian traits with known causative mutations” and reference a 2014 paper. I advise updating that number as OMIA lists (23/9/2016) that 130 Mendelian disorders have their causative mutation known in cattle. I’m fine if you keep the reference you have and include a OMIA reference as your reference is an excellent review paper.”
Authors response: The reviewer correctly points out that the number of Mendelian traits is increasing. We have updated the text to reflect the 130 entries currently listed on the OMIA website: http://omia.angis.org.au/home/.

Reviewer recommendation: “Table 1: You present a large amount of data about each variant and this is wonderful as so many publications would only list one or two of your columns. I would like to see the flanking sequence be increased from 5bp per side to 15bp. Ensemble returns 0 hits for all 6 of your flanking sequences with BLASTing against the UMD3.1 assembly. Increasing it to my recommended length allows BLAST to find unique placements of the sequence. For future researchers to be able to test for disease causing variants having minimal information listed in publication makes it a challenge to make sure you are correctly designing diagnostic probes. I speak from experience in designing diagnostic probes for the IDB cattle chip used in Ireland.”
Authors response: We agree. To ensure a robust BLAST result, we have included 50 bp of flanking genomic sequence on each side of the variant SNPs in Table 1 (101 bp total).

Reviewer recommendation: “Table 3 appears to be only a subset of Table 4 with no additional information. Recommend removing Table 3 and only using Table 4. If you have a justifiable reason for keeping Table 3 than advise listing the breeds in alphabetical order and not in variant 3 frequency order.”
Authors response: We have deleted Table 3 and have adjusted the text accordingly.

Reviewer recommendation: “Table 4. In this you show no animals having the variant 7 allele, yet in the 5th paragraph on page 14 you state that “the F701 substitution was only observed in one of the 1250 animals tested (Salers,…”)”. Please include the frequency of this allele in the Salers row in Table 4.”
Authors response: Table 4 contains genotypes for 1154 animals, including 23 Salers. None of these animals have the F701 variant. There are an additional five Salers in the 96-member beef cattle diversity panel. One of those five Salers has the allele. We have reworded the sentence on
Moreover, the F701 substitution was only observed in one of 1250 animals tested (Salers sire no. 19999882 in MBCDPv2.9).

Reviewer recommendation: Figure 5: Why is node ‘3’ in grey? It's not stated in the Figure.
Authors response: “Variant 3” (pink circle; T606, S610) is identical to that associated with PH in Angus cattle. We have added this sentence to the figure legend.

Reviewer recommendation: “For multiple Supplementary tables: Advise including this information from Table 1: cIUPAC/IUBMB ambiguity codes used for nucleotides:….. Or stating that IUPAC ambiguity codes are used, right now the reader just have to figure it out that a S means the individual is G/C for that SNP.
Authors response: We agree and will add this information as footnotes to Supplementary Material Tables S2 and S5.

Reviewer recommendation: “Table S1: Don’t see why columns C, G, or H are needed: Distance between markers (Mb), Nearest gene (UMD3.1), Distance to nearest gene (kbp), respectively. The rest of the data in the table is excellent and wish more papers provided SNP data at this level. Really like that the ambiguous SNP in the flanking sequence are shown in bold and capital. I recommend removing column C, G, and H as they are meaningless as to information on Parentage SNP.”
Authors response: The information in Column C (genomic spacing) is an important feature of parentage markers since close proximity between two markers reduces the likelihood of meiotic recombination and thus affects the power for parentage assignment. However, the SNPs distance can be derived from UMD3.1 position so we have removed this redundant column. Regarding the “Distance to nearest gene” column, many parentage SNPs are in genes and thus the gene name is sometimes used as an identifier for parentage SNP. Because the parentage SNPs are heavily used on bead arrays and in stand-alone assays, their proximity to genes and their location in a gene is information that gets used quite a lot. For these reasons, we prefer not to delete these columns because this information is tedious to extract from other sources.

Reviewer recommendation: “Table S2: You need to note that a “A, C, T, or G” shown as the genotype indicates that this is a homozygous SNP. One can figure it out since you also show the IUPAC ambiguous letters for heterozygous genotypes, but it would be better to state.”
Authors response: We agree and have added this sentence as a footnote to Table S2: “Homozygous genotypes are listed as a single letter corresponding to the nucleotide at those sites, whereas, IUB ambiguity codes denote heterozygous genotypes those sites: Y = C/ T; R = A/G; M = A/C; K = G/T; W = A/T; S = C/G.”

Reviewer recommendation: “Table S5: Column E is labelled as “EPAS1 protein variant diplotype”. In Table 1, 2, 3 you use the phrase “HIF2A protein variant”. I advise using the same here, otherwise you can easily confuse the reader. Especially as HIF2A is the name of the protein coded by the EPAS1 gene, so if you are referring to protein variants HIF2A is more appropriate than EPAS1”
Authors response: We agree and have change the heading for Column E to “HIF2A diplotype.”. We have also added this footnote to cell G6: “Homozygous genotypes are listed as a single letter corresponding to the nucleotide at those sites, whereas, IUB ambiguity codes denote heterozygous genotypes those sites: Y = C/ T; R = A/G; M = A/C; K = G/T; W = A/T; S = C/G. Genotypes listed as “O” indicate missing information.”

No competing interests were disclosed.

Competing Interests:
Heaton et al. are to be commended for a thorough description of a novel approach to discovering bovine protein variants \textit{in silico}. Further, the authors demonstrate the application of their newly created and publicly accessible mapped genomes of 19 beef breeds by identifying protein variants encoded by \textit{EPAS1}, a gene recently associated with susceptibility to pulmonary hypertension (PH) in cattle at altitudes of 5,200-7,850 ft. (Newman et al., 2015).

The authors also provide an excellent discussion of the plausible \textit{in vivo} effects of the amino acid variants encoded by \textit{EPAS1}. Of particular interest, is the different variant frequencies reported among the various breeds (Figure 4; Table 4). Interestingly, Holstein cattle are susceptible to hypoxia-induced PH and serve as a favorable animal model of the equivalent human condition; yet, as discussed by the authors, PH in Holstein cattle may be due to an amino acid variant (Q270) that differs from S610 variant that contributes to PH susceptibility in Angus cattle. This demonstrates the importance of a having publicly searchable set of mapped genomes for popular breeds of U.S. cattle.

Minor revisions:

1. Third paragraph, first line of Introduction: remove conjunction ("However") from the start of the sentence and place mid-sentence (suggestion only). This may improve sentence flow.

2. I am uncomfortable with the term “high-altitude pulmonary hypertension” because it implies that this physiological phenomenon is only problematic in high altitude regions when, in fact, it occurs at even moderate altitudes in both dairy and beef breeds of cattle (Malherbe et al., 2012; Neary et al., 2016). The Newman study was performed on measurements collected at 5,200-7,850 ft., which in my opinion, is more consistent with moderate than high altitude. “High-altitude” also implies that hypobaric hypoxia is the sole etiology; however, as you discuss, other causes of hypoxia and inflammation, such as respiratory disease, may also be involved. The term pulmonary hypertension (PH) ought to be sufficient.

References

**Competing Interests:** No competing interests were disclosed.

I have read this submission. I believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Author Response 28 Sep 2016

**Michael Heaton,** US Meat Animal Research Center, USA

We agree with the reviewer’s suggestions and have made these changes. Thus, we have removed “high altitude” as a modifier of “pulmonary hypertension” in the title and throughout the article.

**Competing Interests:** No competing interests were disclosed.