

PAPER

Identification and validation of copy number variants in Italian Brown Swiss dairy cattle using Illumina Bovine SNP50 Beadchip®

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Abstract

The determination of copy number variation (CNV) is very important for the evaluation of genomic traits in several species because they are a major source for the genetic variation, influencing gene expression, phenotypic variation, adaptation and the development of diseases. The aim of this study was to obtain a CNV genome map using the Illumina Bovine SNP50 BeadChip data of 651 bulls of the Italian Brown Swiss breed. PennCNV and SVS7 (Golden Helix) software were used for the detection of the CNVs and Copy Number Variation Regions (CNVRs). A total of 5,099 and 1,289 CNVs were identified with PennCNV and SVS7 software, respectively. These were grouped at the population level into 1101 (220 losses, 774 gains, 107 complex) and 277 (185 losses, 56 gains and 36 complex) CNVR. Ten of the selected CNVR were experimentally validated with a qPCR experiment. The GO and pathway analyses were conducted and they identified genes (false discovery rate corrected) in the CNVR related to biological processes, cellular component, molecular function and metabolic pathways. Among those, we found the FCGR2B, PPARa, KATNAL1, DNAJC15, PTK2, TG, STAT family, NPM1, GATA2, LMF1, ECHS1 genes, already known in literature because of their association with various traits in cattle. Although there is variability in the CNVRs detection across methods and platforms, this study allowed the identification of CNVRs in Italian Brown Swiss, overlapping those already detected in other breeds and finding additional ones, thus producing new knowledge for association studies with traits of interest in cattle.

Introduction

The understanding of the genetic variation in livestock species, such as cattle, is crucial to associate genomic regions to the traits of interest. Copy Number Variations (CNV) is defined as a variable copy number of DNA segments ranging from 50bp to several megabases (Mb) compared with a reference genome (Mills *et al.*, 2011). The CNVs are important sources of genetic diversity and provide structural genomic information comparable to single nucleotide polymorphism (SNP) data; they influence gene expression, phenotypic variation, environmental adaptability and disease susceptibility (Wang *et al.*, 2007).

The development of SNP arrays allowed the identification of CNVs by high-throughput genotyping on different cattle breeds. CNV loci were identified in several indicine and taurine breeds, and CNV maps of the bovine genome, using SNPs, Next Generation Sequencing (NGS) and Comparative genome hybridization (CGH) arrays, were reported (Matukumalli et al., 2009; Bae et al., 2010; Fadista et al., 2010; Hou et al., 2012; Bickhart et al., 2012). In livestock, recent studies underlined the effects of the CNVs in intron 1 of the SOX5 gene causing the pea-comb phenotype in chickens (Wright et al., 2009), in the STX17 gene responsible for premature hair greying and susceptibility to melanoma in horses (Rosengren et al., 2008). Also, the CNVs in the ASIP gene are responsible in the leading of different coat colours in goats (Fontanesi et al., 2009). In cattle, Meyers et al. (2010), identified the association between CNVs in a deletion state in the SLC4A2 gene and osteoporosis in Red Angus cows. Additionally, it has been reported that a Copy Number Variation Region (CNVR) located on BTA18 is associated with the index of total merit and protein production, fat production and herd life in Holstein cattle (Seroussi

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et al., 2010). Several CNV detection algorithms based on SNP array are available (Xu et al., 2013). Winchester et al. (2009), Pinto et al. (2011) and Tsuang et al. (2010) recommended the use of a minimum of two algorithms for the identification of CNVs in order to reduce the false discovery rates as the algorithms differ in performance and impact in CNV calling (Xu et al., 2013).

The Italian Brown Swiss breed represents the Italian strain of the Swiss Brown Alpine Breed, originally native of central Switzerland. The typical rusticity of the breed, together with its good production attitude, have led its spread all over many European and American countries, with the differentiation of different genetic groups in relation to various environmental conditions. The milk of the Italian Brown Swiss breed has a good cheese-making attitude due to the low frequency of the allele A of the K-casein, in respect to other breeds (http://www.anarb.it/).

Nowadays in literature, there is not a wholegenome CNV map for the Italian Brown Swiss in a large population dataset. The aim of this study was to obtain a consensus CNV genome map in the Italian Brown Swiss cattle based on the Illumina Bovine SNP50 BeadChip® and two SNP based CNV calling algorithms.





Materials and methods

Sampling and genotyping

The National Association of Italian Brown Swiss breeder (ANARB) provided commercial semen samples for 1342 bulls. Genomic DNA was extracted from semen using the ZR Genomic DNA TM Tissue MiniPrep (Zymo, Irvine, CA, USA). Sample DNA was quantified using NanoQuant Infinite®m200 (Tecan, Männedorf, Switzerland) and diluted to 50 ng/L as required to apply the Illumina Infinium protocol. DNA samples were genotyped using Illumina Bovine SNP50 BeadChip® (Illumina Inc., San Diego, CA, USA) containing 54,001 polymorphic SNPs with an average probe spacing of 51.5 kb and a median spacing of 37.3 kb. In this study, the UMD3.1 assembly was used as the reference genome.

Editing data

All SNPs were clustered and genotyped using the Illumina BeadStudio software V.2.0 (Illumina Inc.). Samples that showed a call rate below 98% were excluded for the CNV detection on autosomal chromosomes. The signal intensity data of Log R Ratio (LRR) and B allele frequency (BAF) were exported from the Illumina BeadStudio software and the overall distribution of derivative log ratio spread (DLRS) values was used in the SVS7 software (Golden Helix Inc.) to identify and filter outlier samples, as described by Pinto et al., (2011). Principal component analysis (PCA) for LRR was performed using the SVS7 software to detect the presence of batch effects and correct the signal intensity values accordingly. Samples with extreme wave factors were excluded from the analysis through the SVS7 software wave correction algorithm.

Genomic waves occur when even after normalization the log ratio data still have a longrange wave outline when charted in a genomic log ratio graph. Waviness is hypothesized to be correlated with the GC content of the probes themselves in addition to the GC content of the region around the probes (Diskin *et al.*, 2008).

Copy number variations detection

Two software were chosen for the detection of CNVs: PennCNV (http://www.openbioinformatics.org/penncnv/) and Copy Number Analysis Module (CNAM) of SVS7 software. The use of two software based on different algorithms has the final aim to reduce the false discovery calls resulting from the limitations of the identification of CNVs based on the Illumina Bovine SNP50 BeadChip.

PennCNV detection

The open access PennCNV online software is nowadays one of the most utilized CNV calling software in bovine studies; it considers multiple sources of information such as the LRR and BAF for every SNP. Furthermore, the software performs quality control measurements for each single CNV analysis. Individual-based CNV calling was performed by PennCNV for all autosomes, using the default parameters of the Hidden Markov Model (HMM) that integrates multiple sources of information to infer CNV calls for individual genotyped samples. To reduce the false discovery rate in CNVs calling we used high quality samples with a standard deviation (SD) of LRR < 0.30 and with default set of BAF drift as 0.01. In addition, we deleted the CNVs which overlapped at least 10% of telomere length (the first and last 500 kb of each autosome were considered representing the telomeres).

SVS7 detection

SVS7 software has a user-friendly graphical interface, efficient pipelines for analysis and workflow, optimized computational speed as well as a technical support. The univariate analysis was used for the CNVs identification. The univariate approach segments each sample independently. An extra covariate is gener-

ated every time a sample has a change-point. The covariates' value is the mean intensity of the original segment for that sample. This results in a spreadsheet displaying all change-points found within the samples. The following options in SVS7, were utilised as suggested in the software manual:

 i) univariate outlier removal; ii) maximum number of segments: search for up to 10 per 10,000 markers; iii) minimum markers per segment: 1; iv) max pairwise permuted P value: 0.005 (2000 permutations per segment pair).

Copy number variation regions definition

CNVRs were defined as in Redon *et al.* (2006) with the BedTools software (Quinlan and Hall, 2010) within software. In addition, consensus regions were created among those identified within the two software, using the Wain *et al.* (2009)'s approach, which identified only CNVRs that fully overlapped each other.

Copy number variation regions validation by quantitative polymerase chain reaction

Quantitative PCR (qPCR) experiments were performed to validate the CNVRs among those identified. The BTF3 gene was selected as a reference location for all qPCR experiments (Bae et al., 2010). Primers for the selected target regions and for the reference gene were designed with the Primer Express® Software v3.0.1 (Life Technologies™, Carlsbad, CA, USA) using the minor groove binder (MGB) quantification parameters. All the qPCR experiments were run in quadruple using the qPCR protocol described by TaqMan® Copy Number Assays kit (Life TecnologiesTM) on 7500 Fast Real-time PCR System instrument (Applied Biosystems by Life TechnologiesTM). The samples for each qPCR experiment were randomly selected with or without CNVs for each CNVR.

Table 1. Descriptive statistics for copy number variations identified with PennCNV and SVS7 software.

Copy number	Number of events	Mean	Median	Total length	Min length	Max length
PennCNV						
0	97	311,345.2	245,646.2	30,200,500	46,665	1,053,143
1	2086	159,066.4	134,534.5	331,711,379	40,374	1,688,267
3	2915	488,559.5	385,138.7	1,423,739,019	41,449	4,457,756
4	1	511,301.5	511,301.5	511,301	511,301	511,301
Total	5099					
SVS7						
Loss	762	94,830.3	57,612.5	72,260,727	11,315	1,440,751
Gain	527	80,324.2	37,591.4	42,330,968	20,342	770,044
Total	1289					

0, homozygous deletion; 1, heterozygous deletion; 3, heterozygous duplication; 4, homozygous duplication; loss, homozygous or heterozygous deletion; gain, homozygous or heterozygous duplication.





The analysis of the crossing thresholds (Ct) for each samples tested was carried out using CopyCaller $^{\rm TM}$ software (Applied Biosystems). The validated CNVR positions were converted from Bos_taurus_UMD3.1 to Btau_4.6.1 assembly using the Batch Coordinate Conversion option in the UCSC database (https://genome.ucsc.edu/) in order to identify potential candidate CNV genes for complex traits.

Copy number variation regions annotation

The full Ensembl v76 gene set for the autosomal chromosomes was downloaded (http://www.ensembl.org/biomart/martview/76 d1cab099658c68bde77f7daf55117e).

Gene ontology (GO) and pathways analyses, using the DAVID Bioinformatics Resources 6.7 (http://david.abcc.ncifcrf.gov/), were performed [(with the high classification stringency option and the false discovery rate (FDR) correction)]. The analyses allowed the identification of molecular functions, biological processes, cellular components and pathways for the genes included in the consensus CNVRs.

Results and discussion

The application of stringent quality filters described above reduced the number of bull samples to be analysed to 651. In this way we filtered most of the potentially problematic data that would have reduced the reliability of the called CNVs. Anyway, it is clear that the remaining bulls constitute a good representation of the Italian Brown Swiss sire population that was used for artificial insemination during the last decades, including several half sibfamilies.

Copy number variations and copy number variation regions detection

Table 1 shows the descriptive statistics of the identified CNVs length using PennCNV and SVS7 software.

Using PennCNV, a total of 5099 CNVs were detected, located in all 29 autosomes with a mean size of 350 kb (± 165.259) ranging from 40.4 kb to 4.46 Mb (median=230 kb). The highest number of CNVs was detected on BTA7 (8.4%). In detail, the homozygous deletion, heterozygous deletion and the heterozygous duplication CNVs with the highest frequency were observed on BTA5 (12.4%), BTA7 (13.4%), BTA2 (7.9%), respectively. Only one

homozygous duplication CNV was identified on BTA25. A total of 1289 CNVs were identified by SVS7 in all the 29 autosomes. The length of the CNVs ranged from 11.3 kb to 1.4 Mb with median and average values equal to 45 kb and 88.9 kb, respectively. The highest frequency of gain (23.9%) and loss (21%) CNVs were detected on BTA28, which also showed the highest number of CNVs in total (22.2%).

The discrepancies among the number of CNVs detected from the two software packages is ascribed to the lack of identification of shorter CNVs of the SVS7 univariate approach (here used) (http://doc.goldenhelix.com/). A graphical representation of CNVs obtained by PennCNV and SVS7 software for each chromosome was visualized by HD-CNV software (http://daleylab.org/lab/?page_id=125) and reported in Figure 1. The graph files allow the visualization of the regions where CNVs were observed across samples overlapping consistently.

A total of 1101 CNVRs were mapped with PennCNV software (Table 2). The total length of the sequence covered by the CNVRs was 682 Mb, which corresponded to 27.14% of the bovine autosomal genome in the Brown Swiss breed. The percentage of sequence covered by CNVRs by chromosome ranged from 16.59 (BTA 12) to 50.14 (BTA 19). The CNVs identi-

Table 2. Descriptive statistics for copy number variation regions identified with PennCNV and SVS7 software.

CNVRs	Number of events	Mean	Median	Total length	Min length	Max length	
PennCNV							
Loss	220	210,454.3	148,427.5	46,299,963	40,754	977,685	
Gain	774	596,255.2	403,827.3	461,501,583	45,465	3,873,856	
Complex	107	1,625,208.1	1,068,260.7	173,897,266	179,707	6,703,707	
Total	1101	210,454.4	148,427.5	46,299,963	40,754	977,685	
SVS7		,	,	, ,	,	,	
Loss	185	116,378.6	61,523.6	21,530,044	11,314	1,440,750	
Gain	56	115,358.1	83,498.5	6,460,049	20,341	460,833	
Complex	36	158,863.1	127,525.5	5,719,073	21,916	770,043	
Total	277	•	*	. ,	•		

CNVRs, copy number variation regions; Loss, homozygous or heterozygous deletion; gain, homozygous or heterozygous duplication; complex, CNVRs defined both as deletion and duplication across samples.

Table 3. Comparison between results of this study and results from literature.

	Results	Overlapped CNVRs of this study							
Reference	Methods of detection	Total CNVRs	Number of samples	Breeds	Length (Mb)	Count Percentage overlap		Total length (Mb)	
Bae et al. (2010)	SNP-based Studies (54k)	368	265	1	63.1	13	8.7%	4.1	
Hou et al. (2011)	SNP-based Studies (54K)	682	521	21	158.0	57	38.0%	22.4	
Jiang <i>et al.</i> (2013)	SNP-based Studies (HD)	367	96	1	42.7	15	10.0%	0.92	
Liu <i>et al.</i> (2010)	CGH-based Studies	177	90	17	28.1	3	2.0%	1.3	
Fadista et al. (2010)	CGH-based Studies	304	20	4	22.0	4	2.7%	1.3	
Bickhart et al. (2012)	Resequencing-based Studies	1,265	5	3	55.6	12	8.0%	2.3	
This study	SNP-based Studies (54k)	150			17.1				





fied with SVS7 software were summarized at the population level according to Redon *et al.* (2006)'s approach, resulting into 277 CNVRs (Table 2). The total length of the sequence covered by the CNVRs was 33.71 Mb (1.35%) of the bovine autosomes. The percentage by chromosome of sequence covered by CNVRs ranged from 0.12 (BTA 10) to 3.5 (BTA 12). The highest frequency of CNVRs were identified on BTAs 8 and 4 for PennCNV and SVS7 software, respectively. The consensus performed between the two software generated 150 consensus CNVRs with a total length of 17.1 Mb (0.68 % of the autosomes), as shown in Supplementary File 1.

Table 3 shows the comparison between the CNVRs detected and those reported in literature, confirming both the existence of high variability in CNVRs detection across platforms, methods, population size, cattle breeds and species. Only a small number of CNVR of those detected in our study (150 regions) have already been detected in other studies: the range of overlapping varies between 2 and 38%. An additional factor that could explain this feature, except for the reasons listed above, is the fact that the CNVR set of this study is the result of a consensus between the two software which work through different algorithms. This strengthens the solidness of the CNVR discovered in our study, but weakens on the other hand the possibility of these lasts to overlap with CNVR of other studies.

The highest overlapping coverage (38%) was found with the study of Hou et al. (2011), in which CNVs detection was performed using BovineSNP50 assay including animals from taurine dairy and beef breeds, breeds of predominantly indicine back-ground, Taurine × Indicine composite and African groups. The previous mentioned dataset included 24 Brown Swiss individuals in which 22 CNVRs were identified on 13 BTAs. Only one CNVR on BTA9 from Hou et al. (2011) (4305338 -4386831 Mbp) resulted in common with the region identified in our study (4050528-4476378 Mbp). The comparison between CNVRs here identified with PennCNV software and those detected in the study of Hou et al. (2011) in Brown Swiss cattle, using the same software, provided five common CNVRs on BTAs 2, 9, 12, 14, 18. Table 4 shows a list of the OTL included in these regions (http://www.a nimalgenome.org/QTLdb/doc/genomeversion#UMD_3.1). Some QTL regions enclosed the CNVR but only those that were not further than 1Mb from the beginning and/or end of the CNVR were taken into account. Additionally, also the QTL regions that were included within the CNVR were considered.

Copy number variation regions validation by quantitative polymerase chain reaction

Eleven CNVRs were selected for the validation taking into account the possibility to test with a molecular approach (quantitative PCR) the results obtained by the different *in silico* approaches we used to call CNVRs; three of which were in common between PennCNV and SVS7 software (a molecular validation of *in silico* consensus), six and two of which were randomly chosen among the CNVRs identified

with the two software, respectively. The validation of the selected regions for each software was the best that could be done taking into account the combination of both the primer design and an adequate number of samples.

Supplementary File 2 reports the primer list for the eleven regions. Ten CNVRs (91%) were confirmed by qPCR experiments. Additionally, the proportions of confirmed positive CNVs in each sample varied from 50% to 100% in each of the confirmed CNVRs; however, the average of false negative rate was equal to 25%. Jiang

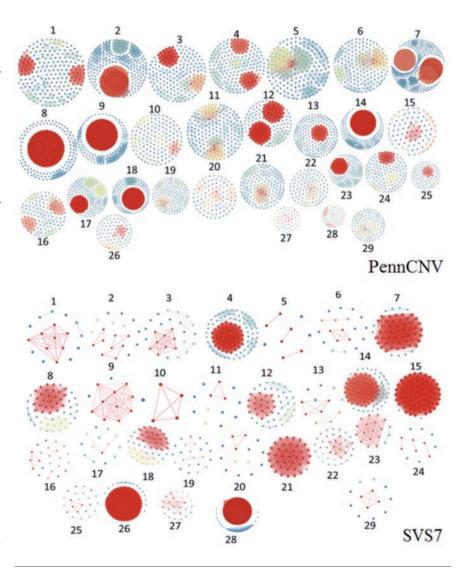


Figure 1. Graphical representation of copy number variations (CNVs) visualized by HD-CNV software. Each graph represents CNVs identified by PennCNV and SVS7 software for each of the 29 bovine autosomes. The size of the graphical representation of each chromosome depends on their total length. Each note represents a CNV, and edges are added between nodes that share 40% (default) overlap. The darker the red spots the more overlapping of CNVs across samples. The blue spots identify the unique events.





et al. (2013) reported similar values rates in the Holstein breed.

Copy number variation regions annotation

Supplementary File 3 shows the 252 Ensembl annotated genes, which correspond to 218 gene symbols in the consensus CNVRs. Supplementary File 4 (spreadsheet genes_clustered by DAVID and spreadsheet genes not clustered by DAVID) report the GO and the pathways analyses for 158 genes among those included in Supplementary File 3. The KEGG pathway analysis revealed that

these genes are mainly represented in the pathway of immune system. This aspect could be interesting considering that disease resistance, that is a complex trait, might be affected, at least in part, by many polymorphisms related to CNVs in addition to many other genetic factors. However, GO and pathway enrichment should be considered with caution in this context. The number of CNVRs could be biased by the detection strategy that was based to define consensus derived CNVRs. That means that we probably obtained a low false positive calling rate but we might have a higher false negative rate (as also mentioned above). For these reasons we focused our discussion on a few spe-

cific genes that might be interesting, according to their potential functional roles and not on the general picture that might be derive from the preliminary GO and pathway enrichment analysis. Among the identified genes, in Table 5 we highlighted those showing differential expression of association with various traits in cattle reported in literature.

In detail, Lewandowska-Sabat *et al.* (2013) in their *in vitro* study, highlighted the role of the *TREM1* (triggering receptor expressed on myeloid cells 1) signaling pathway in which the FCGR2B [(Fc fragment of IgG, low affinity IIb, receptor (CD32)] gene is included. The TREM1, in synergy with the TLR2 (Toll-like

Table 4. Common copy number variation regions in the Brown Swiss breed between Hou et al. (2011) and this study.

Hou et al. (2011) (UMD3.1 assembly) PennCNVThis study (UMD3.1 assembly) PennCNV									QTL	
bta	start CNVR	end CNVR	lenght CNVR	bta	start CNVR	end CNVR	lenght CNVR	start QTL	end QTL	trait (ID)
2	8,788,219	9,113,368	325,150	2	8,788,219	9,040,720	252,501	9,003,563	9,867,063	trans-15-C18:1 fatty acid content (20510)
9	4,305,338	4,386,831	81,494	9	4,050,528	4,476,378	425,850	2,148,415	9,159,784	Interval from first to last insemination (5006)
12	31,368,562	31,679,957	311,396	12	30,099,199	31,555,734	1,456,535	30,751,141	30,912,583	Longissimus muscle area (11733)
14	17,378,950	17,457,836	78,887	14	17,322,658	17,457,836	135,178	9,884,020 12,169,925 16,752,147	19,204,282 20,562,022 18,440,442	Calving ease (maternal) (10958) <i>Longissimus</i> muscle area (4550) Shear force (20791)
18	57,565,406	57,659,303	93,898	18	56,364,657	58,090,087	1,725,430	55,181,080 55,181,080 55,181,080 55,181,080 55,777,394 55,860,765 56,701,305	60,030,732 60,030,732 60,030,732 60,030,732 55,928,978 63,144,054 56,852,890	Stillbirth (direct) (15198), (15199) Birth index (15200) Calving ease (direct) (15201) (15202) Calf size (direct) (15203), (15204) Marbling score (10014) Bilateral convergent strabismus with exophthalmus (10051) Dystocia (direct) (11363)

CNVR, copy number variation region; QTL, quantitative trait locus.

Table 5. List of genes located in identified Brown Swiss copy number variation regions reported in literature.

-				-		
Consensus CNVRs			Genes in (References		
Start	End	Start	End	Ensembl code	Gene symbol	
7957960	7983149	7928113	7944607	ENSBTAG00000021842	FCGR2B	Lewandowska-Sabat <i>et al.</i> (2013)
116895329	117247824	117151549	117233112	ENSBTAG00000008063	$PPAR\alpha$	Bionaz <i>et al</i> . (2013)
30418611	30646042	30519852	30558210	ENSBTAG00000009340	KATNAL1	Zhang <i>et al.</i> (2014)
13179696	13204137	13183734	13266310	ENSBTAG00000034785	DNAJC15	Zhang <i>et al.</i> (2011)
3885798	4017201	3870893	4065010	ENSBTAG00000009578	PTK2	Wang <i>et al.</i> (2013)
9300228	9345140	9262251	9508938	ENSBTAG00000007823	TG	Fernàndez <i>et al.</i> (2014) Bennett <i>et al.</i> (2013)
42976859	43170256	43056660	43132624	ENSBTAG00000021523	STAT3	Zhang <i>et al.</i> (2010)
42976859	43170256	42960226	42996671	ENSBTAG00000010125	STAT5B	3
42976859	43170256	43033597	43054075	ENSBTAG00000009496	STAT5A	
2880532	3189118	3111198	3123860	ENSBTAG00000015316	NPM1	Huang <i>et al.</i> (2010)
59951940	60243916	60016985	60024586	ENSBTAG00000019707	GATA2	Bai <i>et al</i> . (2011)
609241	983759	724446	775899	ENSBTAG00000019745	LMF1	Ren <i>et al.</i> (2011)
25828973	25982293	25856475	25865594	ENSBTAG00000017710	ECHS1	Strillacci et al. (2014)
	Start 7957960 116895329 30418611 13179696 3885798 9300228 42976859 42976859 42976859 2880532 59951940 609241	Start End 7957960 7983149 116895329 117247824 30418611 30646042 13179696 13204137 3885798 4017201 9300228 9345140 42976859 43170256 42976859 43170256 42976859 43170256 2880532 3189118 59951940 60243916 609241 983759	Start End Start 7957960 7983149 7928113 116895329 117247824 117151549 30418611 30646042 30519852 13179696 13204137 13183734 3885798 4017201 3870893 9300228 9345140 9262251 42976859 43170256 43056660 42976859 43170256 42960226 42976859 43170256 43033597 2880532 3189118 3111198 59951940 60243916 60016985 609241 983759 724446	Start End Start End 7957960 7983149 7928113 7944607 116895329 117247824 117151549 117233112 30418611 30646042 30519852 30558210 13179696 13204137 13183734 13266310 3885798 4017201 3870893 4065010 9300228 9345140 9262251 9508938 42976859 43170256 43056660 43132624 42976859 43170256 42960226 42996671 42976859 43170256 43033597 43054075 2880532 3189118 3111198 3123860 59951940 60243916 60016985 60024586 609241 983759 724446 775899	Start End Start End Ensembl code 7957960 7983149 7928113 7944607 ENSBTAG00000021842 116895329 117247824 117151549 117233112 ENSBTAG00000008063 30418611 30646042 30519852 30558210 ENSBTAG00000009340 13179696 13204137 13183734 13266310 ENSBTAG000000034785 3885798 4017201 3870893 4065010 ENSBTAG00000009578 9300228 9345140 9262251 9508938 ENSBTAG00000007823 42976859 43170256 43056660 43132624 ENSBTAG000000021523 42976859 43170256 42960226 42996671 ENSBTAG000000010125 42976859 43170256 43033597 43054075 ENSBTAG00000001936 2880532 3189118 3111198 3123860 ENSBTAG000000015316 59951940 60243916 60016985 60024586 ENSBTAG00000019707 609241 983759 724446 775899 ENSBTAG000000019745	Start End Start End Ensembl code Gene symbol 7957960 7983149 7928113 7944607 ENSBTAG00000021842 FCGR2B 116895329 117247824 117151549 117233112 ENSBTAG00000008063 PPAR α 30418611 30646042 30519852 30558210 ENSBTAG0000009340 KATNAL1 13179696 13204137 13183734 13266310 ENSBTAG00000034785 DNAJC15 3885798 4017201 3870893 4065010 ENSBTAG00000009578 PTK2 9300228 9345140 9262251 9508938 ENSBTAG00000007823 TG 42976859 43170256 43056660 43132624 ENSBTAG000000012523 STAT3 42976859 43170256 42960226 42996671 ENSBTAG00000001125 STAT5B 42976859 43170256 43033597 43054075 ENSBTAG000000019496 STAT5A 2880532 3189118 3111198 3123860 ENSBTAG000000015316 NPM1 59951940 60243916

CNVR, copy number variation region.





receptor 2) pathway, are involved in phagocytosis and production of proinflammatory cytokines, determining optimal host defense during bovine mastitis. The PPAR gene, encoding for the peroxisome proliferator-activated receptor-, maps in the CNVRs found in this study on BTA5. The role of this gene in the fatty acid metabolism is widely described in literature (Bionaz et al., 2013). The CNVRs detected on BTA12 in our study enclose the KATNAL1 (Katanin p60 subunit A-like 1) and DNAJC15 (DnaJ (Hsp40) homolog, subfamily C, member 15) genes. Zhang et al. (2014) characterized the KATNAL1 gene and found the promoter polymorphism associated with semen traits in Chinese bulls. Moreover, SNPs in the DNAJC15 gene were found to be associated with bovine blastocyst rate by Zhang et al. (2011). The polymorphism in the PTK2 (Protein tyrosine kinase 2) gene, annotated in one of the CNVR on BTA14 was found to be associated with milk production traits in Chinese Holstein (Wang et al., 2013). On the same BTA, the gene encoding for the thyroglobulin (TG) was associated with fat distribution, carcass and meat traits in beef cattle (Bennett et al., 2013). Fernàndez et al. (2014) also found the association of SNPs in the TG gene with age of puberty in bulls. The STAT3 [(signal transducer and activator of transcription 3 (acute-phase response factor)], the STAT5B (signal transducer and activator of transcription 5B) and the STAT5A (signal transducer and activator of transcription 5A) genes are annotated in the CNVR on BTA19 detected in our study. The main bovine STATs family members STAT3 and STAT5 are involved in prolactin receptor (PRLR) signaling by JAK/STAT pathway (Janus kinase (JAK) and Signal Transducer and Activator *Transcription*) that activates the expression of milk protein genes (Zhang et al., 2010). The NPM1 gene (nucleophosmin 1) is located in the CNVR found on BTA20. This gene encoding a multifunctional nucleolar phosphoprotein that plays a crucial role in the control of various aspects of cell growth and homeostasis, is a candidate gene for growth traits in cattle (Huang et al., 2010). The GATA2 (GATA binding protein 2) gene, included in the CNVR on BTA22, is involved in the regulation of trophoblast-specific gene transcription in bovine trophoblast CT-1 cells, as described by Bai et al. (2011). The LMF1 (lipase maturation factor 1) gene maps in the CNVR found on BTA25. The mutations of this gene are involved in glyceridemia and hypertriglyceridemia in human and animals, playing an important role in the lipase maturation (Ren et al., 2011). The ECHS1 (enoyl coenzyme A hydratase, short

chain, 1) gene, included in the CNVR on BTA26, was associated with the conjugated linoleic and vaccenic acids in milk in a quantitative trait loci mapping study of Strillacci et al. (2014) in the Italian Brown Swiss cattle breed.

Conclusions

In this study, the first on this breed on a such a large number of individuals, we detected CNVs in the Italian Brown Swiss cattle population based on whole genome SNP genotyping data, using two software packages (PennCNV and SVS7), with the aim to reduce the high error rate commonly recognised in copy number discovery.

The detection from different software for the same CNV call increases confidence in the data giving clearer indications of the dimensions of the CNV identified. CNVRs identified by PennCNV software overlapped in part with the SVS7 data, which emphasised the diversities and the shared features of the two detection methods. The GO and pathway analyses here conducted identified genes that have shown differential expression of association with production traits, carcass and meat traits, reproduction traits and growth traits in cattle. The results enrich the bovine CNVs map providing new information for association studies with economic and health-related traits of interest.

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