FSTest: an efficient tool for cross-population fixation index estimation on variant call

format files

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Abstract

Fixation index (Fst) statistics provide critical insights into the evolutionary processes affecting the structure of genetic variation within and among populations. The Fst statistics have been widely applied in population and evolutionary genetics to identify genomic regions targeted by selection pressures. The FSTest 1.3 software was developed to estimate four Fst statistics of Hudson, Weir and Cockerham, Nei, and Wright using high-throughput genotyping or sequencing data. Here, we introduced FSTest 1.3 and compared its performance with two widely used software of VCFtools 0.1.16 and PLINK 2.0. Chromosome 1 of 1000 Genomes Phase III variant data belonging to South Asian (N = 211) and African (N = 274) populations were included as an example case in this study. Different Fst estimates were calculated for each single nucleotide polymorphism (SNP) in a pairwise comparison of South Asian against African populations, and the results of FSTest 1.3 were confirmed by VCFtools 0.1.16 and PLINK 2.0. Two different sliding window approaches, one based on a fixed number of SNPs and another based on a fixed number of base pair (bp) were conducted using FSTest 1.3 and VCFtools 0.1.16. Our results showed that regions with low coverage genotypic data could lead to an overestimation of Fst in sliding window analysis using a fixed number of bp. FSTest 1.3 could mitigate this challenge by estimating the average of consecutive SNPs along the chromosome. FSTest 1.3 allows direct analysis of VCF files with a small amount of code and can calculate Fst estimates on a desktop computer for more than a million SNPs in a few minutes. FSTest 1.3 is freely available at https://github.com/similab/FSTest.

Keywords: fixation index, population genetics, genetic variation, variant call format, SNP, sliding window

Introduction

Fixation index (Fst) includes a group of statistics that quantify levels of genetic differentiation between populations based on allele frequencies (Biswas and Akey, 2006). Fst can be used to identify loci at which a beneficial allele has recently gone to fixation (Holsinger and Weir, 2009). The Fst has been conceptually defined in many ways (Hudson *et al.*, 1992; Wright, 1949), among which the Weir and Cockerham (1984) and Nei (1973) estimations are the most popular statistics. Two widely used software, including VCFtools 0.1.16 (Danecek *et al.*, 2011) and PLINK 2.0 (Purcell *et al.*, 2007), have provided the functions to compute the Fst values; the first one estimates Weir and Cockerham's and the latter estimates Weir and Cockerham's and Hudson's.

Fst values can be estimated either per single nucleotide polymorphism (SNP) or as an average per genomic windows harboring multiple SNPs, referred to as "sliding window" analysis. In the sliding window analysis, Fst statistics are calculated for a small window of the data; however, windows can be determined based on a fixed number of base pairs (bp) or a fixed number of SNPs. While VCFtools provides sliding window modes only based on bp, PLINK lacks this feature. However, sliding window analysis on a fixed number of bp, particularly when low coverage genotypic data is utilized, leads to overestimation or underestimation of statistics since the number of SNPs inside windows would be inconsistent (Gusnanto *et al.*, 2014). Therefore, a tool that can implement sliding window analysis using a fixed number of SNPs could be an asset. Moreover, except for VCFtools, a limited number of tools are currently available to produce Fst statistics from variant call format (VCF) data. Here, we introduced FSTest 1.3 software, a single package that completely covers well-known estimates of Fst, performs both SNP-based and

sliding windows analyses, and supports VCF files as inputs. The results of the software were

compared with those obtained from VCFtools 0.1.16 and PLINK 2.0.

Materials and methods

FSTest 1.3 is implemented in Python and accepts VCF files as input. The tool automatically

manages missing genotypes and supports phased and unphased genotypic data. FSTest

calculates between-population Fst estimators of Hudson *et al*. (1992), Weir and Cockerham

(1984), Nei (1973), and Wright (1949) as follows:

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$$Fst_{Hudson} = \frac{(\tilde{p}_1 - \tilde{p}_2)^2 - \frac{\tilde{p}_1(1-\tilde{p}_1)}{n_1-1} - \frac{\tilde{p}_2(1-\tilde{p}_2)}{n_2-1}}{\tilde{p}_1(1-\tilde{p}_2) + \tilde{p}_2(1-\tilde{p}_1)},$$

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$$Fst_{Weir\,and\,Cockerham} = 1 - \frac{2\frac{n_1 n_2}{n_1+n_2}(\frac{1}{n_1+n_2-2})[n_1\tilde{p}_1(1-\tilde{p}_1)+n_2\tilde{p}_2(1-\tilde{p}_2)]}{\frac{n_1 n_2}{n_1+n_2}(\tilde{p}_1-\tilde{p}_2)^2 + (2\frac{n_1 n_2}{n_1+n_2}-1)\frac{1}{n_1+n_2-2}[n_1\tilde{p}_1(1-\tilde{p}_1)+n_2\tilde{p}_2(1-\tilde{p}_2)]},$$

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$$Fst_{Nei} = \frac{(\tilde{p}_1 - \tilde{p}_2)^2}{(\tilde{p}_1 + \tilde{p}_2)\left(1 - \frac{\tilde{p}_1 + \tilde{p}_2}{2}\right)},$$

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$$Fst_{Wright} = \frac{(\tilde{p}_1 + \tilde{p}_2)\left(1 - \frac{\tilde{p}_1 + \tilde{p}_2}{2}\right) - [\left(\frac{n_1}{n_1+n_2}\right)2\tilde{p}_1(1-\tilde{p}_1) + \left(\frac{n_2}{n_1+n_2}\right)2\tilde{p}_2(1-\tilde{p}_2)]}{(\tilde{p}_1 + \tilde{p}_2)\left(1 - \frac{\tilde{p}_1 + \tilde{p}_2}{2}\right)},$$

where n_i is the sample size, and $\tilde{p}_i$ is the sample allele frequency in population i for $i \in \{1, 2\}$.

All Fst estimations range between 0 and 1, and values less than zero are denoted zero by the

software as they do not have a biological interpretation.

For sliding window analysis, the average Fst values across SNPs will be taken for each window.

Users can also implement Z-transformation on all Fst estimates for SNP- or window-based

approaches to get unbiased Z_{Fst} values as given:

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$$Z_{FST_i} = \frac{Fst_i - \overline{Fst}}{S},$$

where Fst_i is the Fst value of ith SNP or ith window, $\overline{FST}$ and S represent the average value and

standard deviation of Fst between two populations estimated from all SNPs or windows. We

have also added the functionality of creating publication-quality Manhattan plots for the results

of SNP and sliding window analyses.

Here, we demonstrated an example case by analyzing chromosome 1 of 1000 Genomes Phase III

variant data downloaded from the project's FTP site (https://ftp.1000genomes.ebi.ac.uk/). A total

of 485 samples belonging to two different ancestries, South Asian (N = 211) and African (N =

274), were included in this study. Quality control was performed on 6,468,094 SNPs on

chromosome 1 using PLINK 2.0 software (Purcell *et al.*, 2007). Markers with minor allele

frequency $<$ 0.01, SNP calling rate $<$ 0.90, and extreme departure from Hardy–Weinberg

equilibrium (p-value $<$ 10^{-7}) were removed. Individuals with missing genotype data $>$ 0.1

were also removed from the study. After quality control, a total of 1,206,810 SNPs from 485

individuals remained for further analysis. The average±standard error (SE), minimum, and

maximum distance between SNPs in the final genotypic data were 207±17, 1, and 21,050,296

bp. The longest distance between SNPs belonged to the region of 12,148,5143-142,535,439 bp.

Fst values of Hudson, Weir and Cockerham, Nei, and Wright were estimated using FSTest 1.3,

Hudson's and Weir and Cockerham's using PLINK 2.0 (Purcell *et al.*, 2007), and Weir and

Cockerham's by VCFtools 0.1.16 (Danecek *et al.*, 2011) (Table 1). Spearman's correlations

among the estimates were calculated, and a correlogram was constructed using corrplot 0.92

package (Wei *et al.*, 2017) in R 4.1.0 (R Core Team, 2021).

106   << Table 1 around here >>

Two sliding window approaches, one based on a fixed number of SNPs and another based on a

fixed number of bp, were applied using FSTest 1.3 and VCFtools 0.1.16 software, respectively

(Table 1). In the first one, a 500 SNPs sliding window with a step size of 50 SNPs, and in the

latter, a 100 kb sliding window with a step size of 10 kb was used. PLINK 2.0 software was not

included in the window-based analyses as it lacks this feature.

A desktop computer with AMD Ryzen 7 4800H (8-core) CPU and 32 gigabytes installed RAM

was used for all analyses. The time of analyses was measured using the log files and compared

among the software. CMplot package (Yin *et al.*, 2021) in R 4.1.0 (R Core Team, 2021) was

used to create circos and Manhattan plots. Visualization of chromosomal segments harboring

$Fst_{mean} > 0.15$ was performed using ChromoMap 0.4.1 (Anand and Rodriguez Lopez, 2022) in

R 4.1.0 (R Core Team, 2021).

Results and discussion

Different Fst statistics were estimated for 1,206,810 SNPs in the pairwise comparison of

chromosome 1 between South Asian and African populations (Fig. 1a). Estimates of Fst_{Hudson}

and $Fst_{Weir\ and\ Cockerham}$ were comparable among the results of FSTest 1.3, VCFtools 0.1.16,

and PLINK 2.0 (Table 2). We obtained similar means±SE for Fst_{Hudson} and

$Fst_{Weir\ and\ Cockerham}$ (0.07±0.00), as well as for Fst_{Wright} and Fst_{Nei} (0.04±0.00). The highest

and the lowest range of Fst values belonged to the Nei (0-0.85) and Wright (-0.08-0.39)

estimators, respectively. Spearman's correlations between different Fst estimates were obtained,

and a correlogram was constructed (Fig 1b). We observed 100% correlations between Fst_{Hudson}

and $Fst_{Weir\ and\ Cockerham}$ estimates achieved by FSTest 1.3 and those by VCFtools 0.1.16 and

PLINK 2.0 (Table 2). Descriptive statistics and correlations of FSTest 1.3 results with those from

VCFtools 0.1.16 and PLINK 2.0 confirms the accuracy of FSTest 1.3 performance.

133   << Figure 1 around here >>

134   << Table 2 around here >>

$Fst_{Weir\ and\ Cockerham}$ was estimated using two different sliding window approaches: (i) based

on a fixed number of SNP using FSTest 1.3, and (ii) based on fixed number of bp using

VCFtools 0.1.16, and the results were shown in Fig 1c and Fig 1d, respectively. To compare the

results of two approaches, windows with $Fst_{mean} > 0.15$ were highlighted (Fig 1c and 1d) and

overlapping segments were compared (Fig 1e). We identified 24 and 23 genomic windows with

$Fst_{mean} > 0.15$ using FSTest 1.3 and VCFtools 0.1.16, respectively. As shown in Figure 1e,

among FSTest 1.3 results, the highest Fst values belonged to the region of 100,000-102,500 kb,

containing eight genomic windows with $Fst_{mean} > 0.15$ ($Fst_{mean} = 0.17$). In contrast, this

region was ranked second by VCFtools 0.1.16, after the 102,500-105,000 kb segment, which

only contained one genomic window with $Fst_{mean} > 0.15$. After checking the results of

VCFtools 0.1.16, we realized that this genomic window only harbors one SNP. We also found

that sliding window analysis using a fixed number of bp might be more sensitive to the edges of

genomic segments. Fig 1d demonstrates that one of the windows with $Fst_{mean} > 0.15$ (no.

12235; chr1: 143,540- 143,640 kb) was located after the segment without variants in the middle

of chromosome 1 and only contained 10 SNPs. Our results are in parallel with Gusnanto *et al.*

(2014) findings that regions with low coverage genotypic data could lead to overestimation or

underestimation of Fst as the number of SNPs inside windows would be inconsistent. Therefore,

the interpretation of the VCFtools 0.1.16 results should always be with attention to the number of

SNPs in the genomic windows. Meanwhile, FSTest could tackle this problem by estimating the

average of consecutive SNPs along the chromosome.

Regarding the analysis runtime, FTSest 1.3 performed slightly slower than VCFtools 0.1.16

(Table 3). Reading VCF files can be time-consuming, mainly when dealing with genomic data

from whole-genome sequencing or large-scale studies. VCF files contain structured data with

numerous fields and annotations for each variant, such as variant type, quality scores, and

genotype information, which must be processed. To mitigate the time-consuming nature of

reading VCF files, parallel processing techniques can help improve the speed of reading and

processing VCF files, which is suggested for the next versions. PLINK 2.0 performed

significantly faster than FSTest 1.3 and VCFtools 0.1.16 due to using light PLINK format files as

inputs.

166   << Table 3 around here >>

In conclusion, FSTest 1.3 software contains extra pairwise Fst estimation methods, which are not

present in the VCFtools 0.1.16 and PLINK 2.0 software. The tool allows direct analysis of VCF

files with a small amount of code and is able to conduct sliding window analyses based on a

fixed number of SNPs. FSTest 1.3 can calculate Fst estimates the results on a desktop computer

for more than a million SNPs in a few minutes.

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Data availability

The genotypes used in this study is available at: https://ftp.1000genomes.ebi.ac.uk/

Author contributions

SV and SS developed the original FSTest software. SV wrote the manuscript. All authors read

and approved the final manuscript.

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 Figures

Figure 1. Circos plot (a) shows different Fst values estimated by FSTest 1.3, PLINK 2.0, and

VCFtools 0.1.16 tools. From inside to outside, layers represent the distribution of Fst_{Hudson}

obtained from PLINK 2.0, Fst_{Hudson} from FSTest 1.3, $Fst_{Weir\ and\ Cockerham}$ from PLINK 2.0,

$Fst_{Weir\ and\ Cockerham}$ from VCFtools 0.1.16, $Fst_{Weir\ and\ Cockerham}$ from FSTest 1.3, Fst_{Nei}

from FSTest 1.3, and Fst_{Wright} from FSTest 1.3. Correlogram (b) demonstrates the Spearman's

correlations among different Fst estimations by applied tools. Manhattan plots (c) and (d) show

the distribution of $Fst_{Weir\ and\ Cockerham}$ using two sliding window analyses by FSTest 1.3 and

VCFtools 0.1.16. A sliding window of 500 SNPs with a step size of 50 SNP was applied to

FSTest 1.3, and a sliding window of 100kb with a step size of 10 kb was implemented in

VCFtools 0.1.16. Red points represent windows with $Fst_{mean} > 0.15$ and numbers show the

number of windows. Chromosome visualizations (e) and (f) depicts the overlapping windows

with $Fst_{mean} > 0.15$ obtained by FSTest 1.3 and VCFtools 0.1.16, respectively.

Tables

Table 1. Different features of Fixation index (Fst) estimation using FSTest 1.3, VCFtools 0.1.16,

and PLINK 2.0 software. Different Fst values per SNP along with two different window-based

approaches, in which the window is determined based on a fixed number of base pairs (Win-bp)

or a fixed number of SNP (Win-SNP), were compared among the tools.

Software	Fst estimates	Approach		
		SNP	Win-SNP	Win-bp
FSTest 1.3	$Fst_{Hudson}, Fst_{Weir\ and\ Cockerham},$ Fst_{Wright}, Fst_{Nei}	✓	✓	-
VCFtools 0.1.16	$Fst_{Weir\ and\ Cockerham}$	✓	-	✓
PLINK 2.0	$Fst_{Hudson}, Fst_{Weir\ and\ Cockerham}$	✓	-	-

240 **Table 2.** Mean, median, standard error (SE), and range of different Fst estimates per SNP.

Tool	Statistics	Mean	Median	SE	Range
FSTest1.3	Fst_{Hudson}	0.07	0.04	0	0-0.59
	$Fst_{Weir\ and\ Cockerham}$	0.07	0.04	0	0-0.56
	Fst_{Wright}	0.04	0.02	0	-0.08-0.39
	Fst_{Nei}	0.04	0.04	0	0-0.85
VCFtools 0.1.16	$Fst_{Weir\ and\ Cockerham}$	0.07	0.04	0	0-0.56
PLINK 2.0	Fst_{Hudson}	0.07	0.04	0	0-0.59
	$Fst_{Weir\ and\ Cockerham}$	0.07	0.04	0	0-0.56

Table 3. Comparison of runtime among different tools and Fst estimates included in this study.

Software	SNP-based				Window-based	Input file
	Fst_{Hudson}	$Fst_{Weir\ and\ Cockerham}$	Fst_{Wright}	Fst_{Nei}	$Fst_{Weir\ and\ Cockerham}$	
FSTest 1.3	72.28[a]	69.84	70.2	77.15	69.39	VCF
VCFtools 0.1.16	NE[b]	38	NE	NE	38	VCF
PLINK 2.0	< 1	< 1	NE	NE	NE	PLINK format

[a] In seconds.

[b] Represents the Fst statistics that were not supported by the software and were not estimated.

(a)

(b)

(c)

(d)

(e)