Type: Letter/Methods

`selscan 2.0`: scanning for sweeps in unphased data

Zachary A. Szpiech[1,2,*]

[1] *Department of Biology, Pennsylvania State University, University Park, PA 16802, USA*
[2] *Institute for Computational and Data Sciences, Pennsylvania State University, University Park, PA 16802, USA*
[*] *Correspondence: szpiech@psu.edu*

Abstract

Haplotype-based scans to identify recent and ongoing positive selection have become commonplace in evolutionary genomics studies of numerous species across the tree of life. However, the most widely adopted approaches require phased haplotypes to compute the key statistics. Here we release a major update to the selscan software that re-defines popular haplotype-based statistics for use with unphased "multi-locus genotype" data. We provide unphased implementations of iHS, nSL, XP-EHH, and XP-nSL and evaluate their performance across a range of important parameters in a generic demographic history. Source code and executables are available at https://www.github.com/szpiech/selscan.

1 Introduction

Haplotype-based summary statistics—such as iHS (Voight, et al. 2006), nSL (Ferrer-Admetlla, et al. 2014), XP-EHH (Sabeti, et al. 2007), and XP-nSL (Szpiech, et al. 2021)—have become commonplace in evolutionary genomics studies to identify recent and ongoing positive selection in populations (e.g.,Colonna, et al. 2014; Zoledziewska, et al. 2015; Nedelec, et al. 2016; Crawford, et al. 2017; Meier, et al. 2018; Lu, et al. 2019; Zhang, et al. 2020; Salmon, et al. 2021). When an adaptive allele sweeps through a population, it leaves a characteristic pattern of long high-frequency haplotypes and low genetic diversity in the vicinity of the allele. These statistics aim to capture these signals by summarizing the decay of haplotype homozygosity as a function of distance from a putatively selected region, either within a single population (iHS

and nSL) or between two populations (XP-EHH and XP-nSL). However, each of these statistics presumes that haplotype phase is known.

Recent work has shown that converting haplotype data into multi-locus genotype data is an effective approach for using haplotype-based selection statistics such as G12, LASSI, and saltiLASSI (Harris, et al. 2018; Harris and DeGiorgio 2020; DeGiorgio and Szpiech 2021) in unphased data. Recognizing this, we have reformulated the iHS, nSL, XP-EHH, and XP-nSL statistics to use multi-locus genotypes and provided an easy-to-use implementation in selscan 2.0 (Szpiech and Hernandez 2014). We also evaluate the performance of these unphased statistics under various generic demographic models.

2 New Approaches

When the --unphased flag is set in selscan v2.0+, biallelic genotype data is collapsed into multi-locus genotype data by representing the genotype as either 0, 1, or 2—the number of derived alleles observed. In this case, selscan v2.0+ will then compute iHS, nSL, XP-EHH, and XP-nSL as described below. We follow the notation conventions of Szpiech and Hernandez (2014).

2.1 Extended Haplotype Homozygosity

In a sample of n diploid individuals, let $\mathcal{C}$ denote the set of all possible genotypes at locus x_0. For multi-locus genotypes, $\mathcal{C} \coloneqq \{0,1,2\}$, representing the total counts of a derived allele. Let $\mathcal{C}(x_i)$ be the set of all unique haplotypes extending from site x_0 to site x_i either upstream or downstream of x_0. If x_1 is a site immediately adjacent to x_0, then $\mathcal{C}(x_1) \coloneqq \{00,01,02,10,11,12,20,21,22\}$, representing all possible two-site multi-locus genotypes. We can then compute the extended haplotype homozygosity (EHH) of a set of multi-locus genotypes as

$$EHH(x_i) \sum_{h \in \mathcal{C}(x_i)} \frac{\binom{n_h}{2}}{\binom{n}{2}},$$

where n_h is the number of observed haplotypes of type h.

If we wish to compute the EHH of a subset of observed haplotypes that all contain the same 'core' multi-locus genotype, let $\mathcal{H}_c(x_i)$ be the partition of $\mathcal{C}(x_i)$ containing genotype $c \in \mathcal{C}$ at x_0. For example, choosing a homozygous derived genotype ($c = 2$) as the core, $\mathcal{H}_2 \coloneqq \{20,21,22\}$. Thus, we can compute the EHH of all individuals carrying a given genotype at site x_0 extending out to site x_i as

$$EHH_c(x_i) = \sum_{h \in \mathcal{H}_c(x_i)} \frac{\binom{n_h}{2}}{\binom{n_c}{2}},$$

where n_h is the number of observed haplotypes of type h and n_c is the number of observed multi-locus genotypes with core genotype of c. Finally, we can compute the complement EHH of a sample of multi-locus genotypes as

$$cEHH_c(x_i) = \sum_{h \in \mathcal{C}(x_i) \setminus \mathcal{H}_c(x_i)} \frac{\binom{n_h}{2}}{\binom{n_{c'}}{2}},$$

where $n_{c'}$ is the number of observed multi-locus genotypes with a core genotype of not c.

2.2 iHS and nSL

Unphased iHS and nSL are calculated using the equations above. First, we compute the integrated haplotype homozygosity (iHH) for the homozygous ancestral ($c = 0$) and derived ($c = 2$) core genotypes as

$$iHH_c = \sum_{i=1}^{|\mathcal{D}|} \frac{1}{2}\big(EHH_c(x_{i-1}) + EHH_c(x_i)\big)\mathrm{d}(x_{i-1}, x_i) + \sum_{i=1}^{|\mathcal{U}|} \frac{1}{2}\big(EHH_c(x_{i-1}) + EHH_c(x_i)\big)\mathrm{d}(x_{i-1}, x_i),$$

where $\mathcal{D}$ is the set of downstream sites from the core locus and $\mathcal{U}$ is the set of upstream sites. $d(x_{i-1}, x_i)$ is a measure of genomic distance between to markers and is the genetic distance in centimorgans or physical distance in basepairs for iHS (Voight, et al. 2006) or the number of sites observed for nSL (Ferrer-Admetlla, et al. 2014). We similarly compute the complement integrated haplotype homozygosity (ciHH) for both homozygous core genotypes as

$$ciHH_c = \sum_{i=1}^{|\mathcal{D}|} \frac{1}{2}\left(cEHH_c(x_{i-1}) + cEHH_c(x_i)\right)\mathrm{d}(x_{i-1}, x_i)$$

$$+ \sum_{i=1}^{|\mathcal{U}|} \frac{1}{2}\left(cEHH_c(x_{i-1}) + cEHH_c(x_i)\right)\mathrm{d}(x_{i-1}, x_i).$$

The (unstandardized) unphased iHS is then calculated as

$$iHS = \begin{cases} \mathrm{iHS}_2, & \text{if } \mathrm{iHS}_2 > \mathrm{iHS}_0 \\ \text{-}\mathrm{iHS}_0, & \text{otherwise} \end{cases},$$

where $iHS_2 = \log_{10}\left(\frac{iHH_2}{ciHH_2}\right)$ and $iHS_0 = \log_{10}\left(\frac{iHH_0}{ciHH_0}\right)$. Unstandardized iHS scores are then normalized in frequency bins, as previously described (Voight, et al. 2006; Ferrer-Admetlla, et al. 2014). Unstandardized unphased nSL is computed similarly with the appropriate distance measure. Large positive scores indicate long high-frequency haplotypes with a homozygous derived core genotype, and large negative scores indicate long high-frequency haplotypes with a homozygous ancestral core genotype. Clusters of extreme scores in both directions indicate evidence for a sweep.

2.3 XP-EHH and XP-nSL

Unphased XP-EHH and XP-nSL are calculated by comparing the iHH between populations A and B, using the entire sample in each population. iHH in a population P is computed as

$$iHH_{\mathrm{P}} = \sum_{i=1}^{|\mathcal{D}|} \frac{1}{2}\left(\mathrm{EHH}(x_{i-1}) + \mathrm{EHH}(x_i)\right)\mathrm{d}(x_{i-1}, x_i) + \sum_{i=1}^{|\mathcal{U}|} \frac{1}{2}\left(\mathrm{EHH}(x_{i-1}) + \mathrm{EHH}(x_i)\right)\mathrm{d}(x_{i-1}, x_i),$$

where the distance measure is given as centimorgans or basepairs for XP-EHH (Sabeti, et al. 2007) and number of sites observed for XP-nSL (Szpiech, et al. 2021). The XP statistics between population A and B are then computed as $XP = \log_{10}\left(\frac{iHH_A}{iHH_B}\right)$ and are normalized genome wide in a single bin. Large positive scores indicate long high-frequency haplotypes in

population A, and large negative scores indicate long high-frequency haplotypes in population B. Clusters of extreme scores in one direction indicate evidence for a sweep in that population.

3 Methods

3.1 Simulations

We evaluate the performance of the unphased versions of iHS, nSL, XP-EHH, and XP-nSL under a generic two-population divergence model using the coalescent simulation program discoal (Kern and Schrider 2016). We explore five versions of this generic model and name them Demo 1 through Demo 5 (Table 1). Let N_0 and N_1 be the effective population sizes of Population 0 and Population 1 after the split from their ancestral population (of size N_A). For Demo 1, we keep a constant population size post-split and let $N_0 = N_1 = 10{,}000$. For Demo 2, we keep a constant population size post-split and let $N_0 = 2N_1 = 10{,}000$. For Demo 3, we keep a constant population size post-split and let $2N_0 = N_1 = 10{,}000$. For Demo 4, we initially set $N_0 = N_1 = 10{,}000$ and let N_0 grow stepwise exponentially every 50 generations starting at 2,000 generations ago until $N_0 = 5N_1 = 50{,}000$. For Demo 5, we initially set $N_0 = N_1 = 10{,}000$ and let N_1 grow stepwise exponentially every 50 generations starting at 2,000 generations ago until $5N_0 = N_1 = 50{,}000$.

For each demographic history we vary the population divergence time $t_d \in \{2000,\ 4000,\ 8000\}$ generations ago. For non-neutral simulations, we simulate a sweep in Population 0 in the middle of the simulated region across a range of selection coefficients $s \in \{0.005,\ 0.01,\ 0.02\}$. We vary the frequency at which the adaptive allele starts sweeping as $e \in \{0,\ 0.01,\ 0.02,\ 0.05,\ 0.10\}$, where $e = 0$ indicates a hard sweep and $e > 0$ indicates a soft sweep, and we also vary the frequency of the selected allele at time of sampling $f \in \{0.7,\ 0.8,\ 0.9,\ 1.0\}$ as well as $g \in \{50,\ 100\}$ representing fixation of the sweeping allele g generations ago. For all simulations we set the genome length to be $L = 500{,}000$ basepairs, the ancestral effective population size to be $N_A = 10{,}000$, the per site per generation mutation rate at $\mu = 2.35 \times 10^{-8}$,

and the per site per generation recombination rate at $r = 1.2 \times 10^{-8}$. For neutral simulations, we simulate 1,000 replicates for each parameter set, and for non-neutral simulations we simulate 100 replicates for each parameter set. As iHS and nSL are single population statistics, we only analyze Demo 1, Demo 3, and Demo 4 with these statistics, as Demo 2 and Demo 5 have a constant size history identical to Demo 1 for Population 0, where the sweeps are simulated.

For all simulations, we compute the relevant statistics (--ihs, --nsl, --xpehh, or --xpnsl) with selscan v2.0, using the --unphased and --trunc-ok flags. For iHS and XP-EHH, we also use the --pmap flag in order to use physical distance instead of a recombination map.

3.2 Power and False Positive Rate

To compute power for iHS and nSL, we follow the approach of Voight et al. (2006). For these statistics, each non-neutral replicate is individually normalized jointly with all matching neutral replicates in 1% allele frequency bins. Because extreme values of the statistic are likely to be clustered along the genome (Voight, et al. 2006), we then compute the proportion of extreme scores ($|iHS| > 2$ or $|nSL| > 2$) within 100kbp non-overlapping windows. We then bin these windows into 10 quantile bins based on the number of scores observed in each window and call the top 1% of these windows as putatively under selection. We calculate the proportion of non-neutral replicates that fall in this top 1% as the power. To compute the false positive rate, we compute the proportion of neutral simulations that fall within the top 1%.

To compute power for XP-EHH and XP-nSL, we follow the approach of Szpiech et al. (2021). For these statistics, each non-neutral replicate is individually normalized jointly with all matching neutral replicates. Because extreme values of the statistic are likely to be clustered along the genome (Szpiech, et al. 2021), we then compute the proportion of extreme scores (XP-EHH > 2 or XP-nSL > 2) within 100kbp non-overlapping windows. We then bin these windows into 10 quantile bins based on the number of scores observed in each window and call the top 1% of these windows as putatively under selection. We calculate the proportion of non-

neutral replicates that fall in this top 1% as the power. To compute the false positive rate, we compute the proportion of neutral simulations that fall within the top 1%.

4 Results

We find that the unphased versions of iHS and nSL have good power (Figures 1, S1-S4, S13-16, and S21-24) to detect selection prior to fixation of the allele, with nSL generally outperforming iHS. In smaller populations (Figure 1C and 1D), power does suffer relative to larger populations (Figure 1A, 1B, 1E, 1F). Each of these statistics also have low false positive rates hovering around 1% (Table S1).

Similarly, we find that the unphased versions of XP-EHH and XP-nSL have good power as well (Figures 2, 3, S5-S12, S17-S20, and S25-S32). When the sweep takes place in the smaller of the two populations (Figure 2C and 2D), we see a similar decrease in power. When one population is undergoing exponential growth (Figure 3) performance is generally quite good, likely the result of a larger effective selection coefficient in large populations. These two-population statistics generally outperform their single-population counterparts, especially for sweeps that have reached fixation recently. Each of these statistics also have low false positive rates hovering around 1% (Table S1).

5 Discussion

We introduce multi-locus genotype versions of four popular haplotype-based selection statistics—iHS (Voight, et al. 2006), nSL (Ferrer-Admetlla, et al. 2014), XP-EHH (Sabeti, et al. 2007), and XP-nSL (Szpiech, et al. 2021)—that can be used when the phase of genotypes is unknown. We implement these updates in the latest v2.0 update of the program selscan (Szpiech and Hernandez 2014), with source code and pre-compiled binaries available at https://www.github.com/szpiech/selscan.

6 Acknowledgements

This work was supported by start-up funds from the Pennsylvania State University's Department of Biology. Computations for this research were performed using the Pennsylvania State University's Institute for Computational Data Sciences' Roar supercomputer.

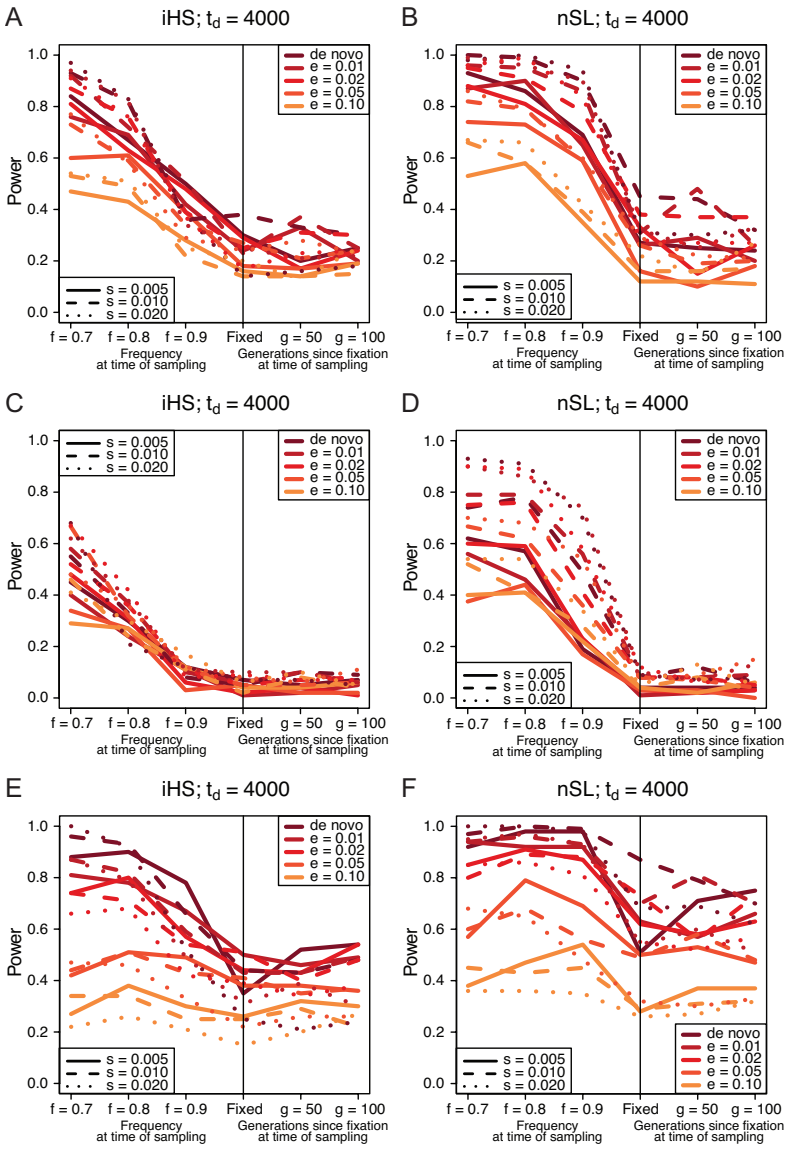

Figure 1. Power curves for unphased implementations of iHS (A, C, and E) and nSL (B, D, and F) under demographic histories Demo 1 (A and B), Demo 3 (C and D), and Demo 4 (E and F). s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

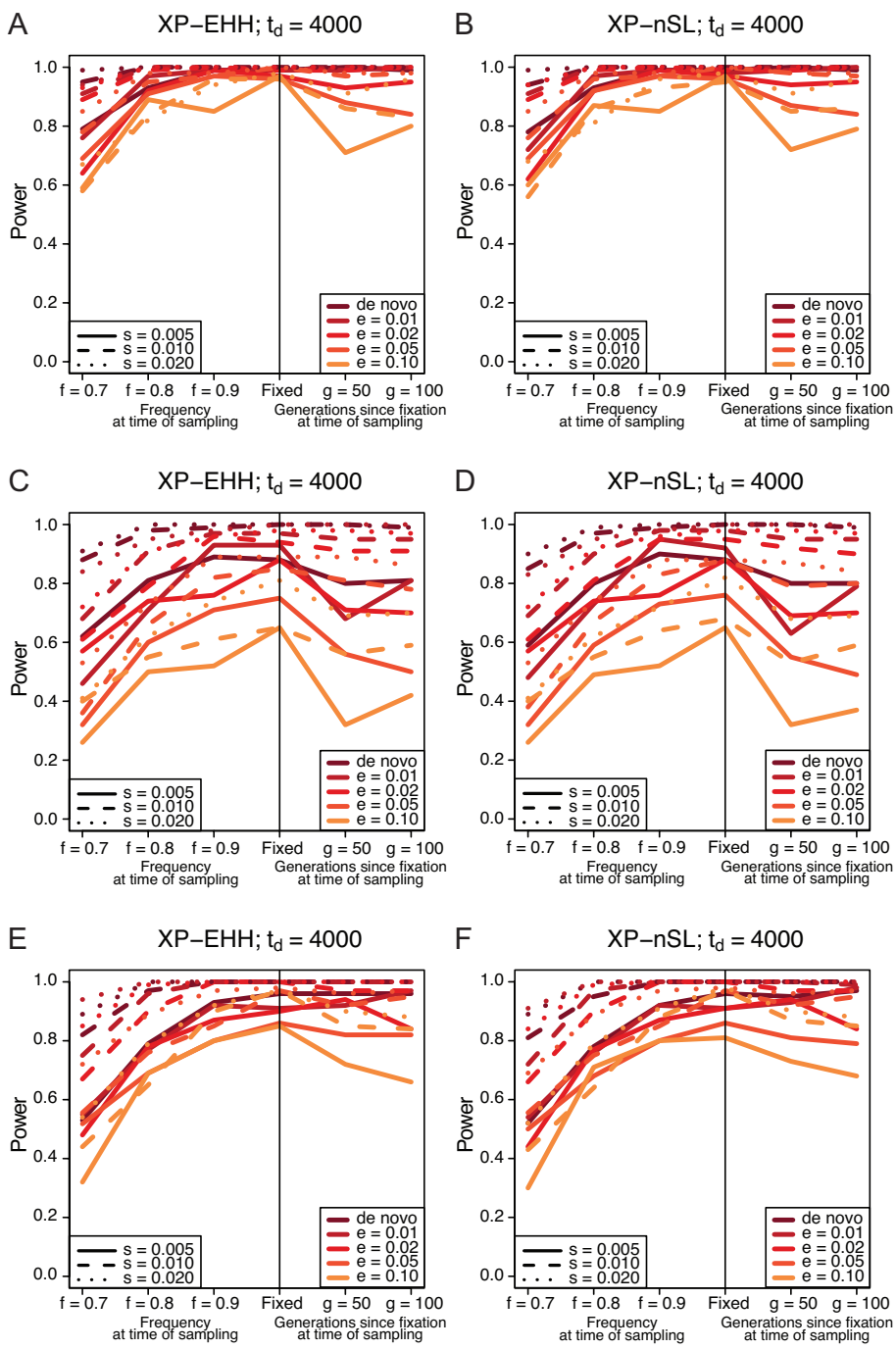

Figure 2. Power curves for unphased implementations of XP-EHH (A, C, and E) and XP-nSL (B, D, and F) under demographic histories Demo 1 (A and B), Demo 2 (C and D), and Demo 3 (E and F). s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

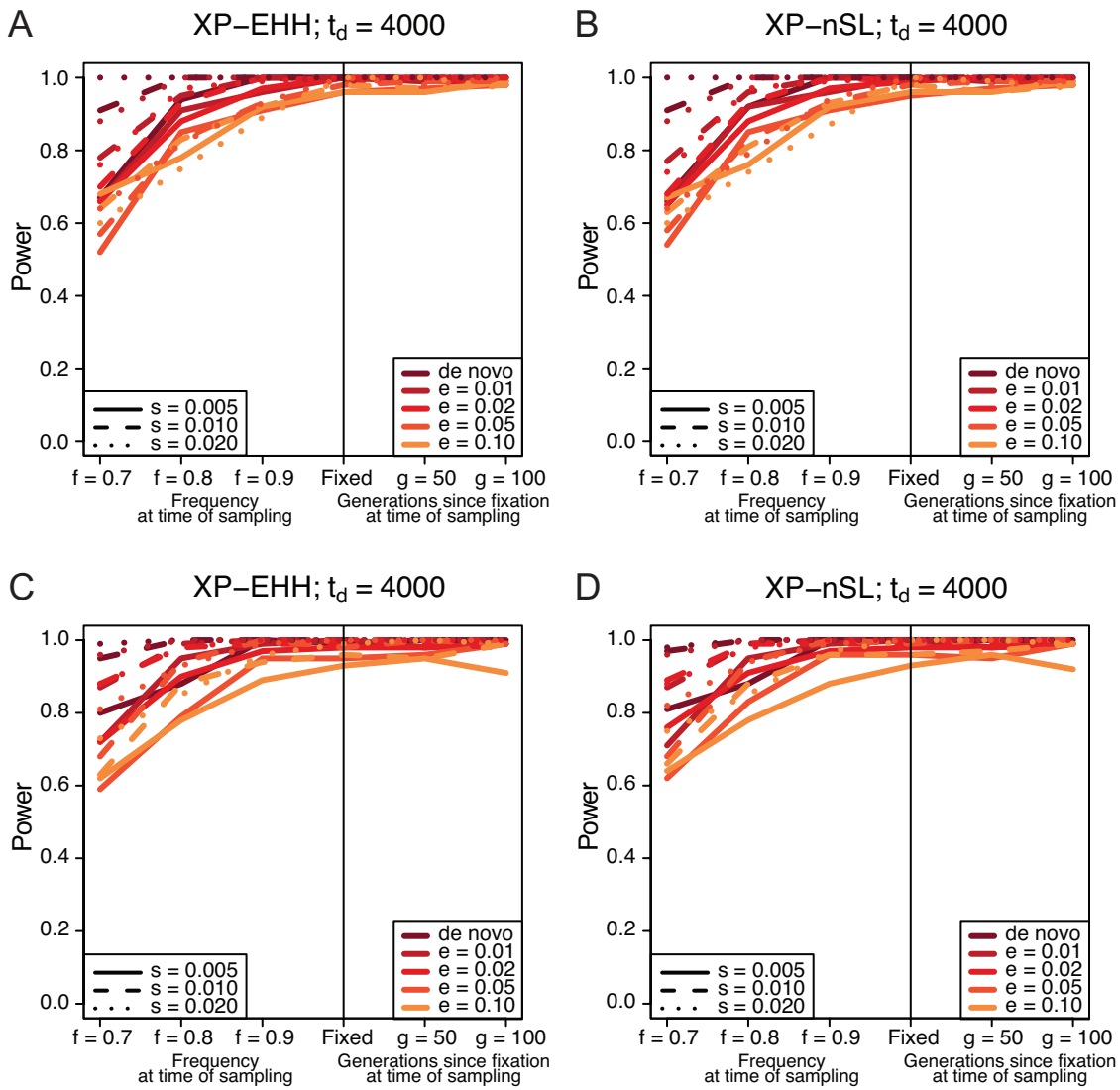

Figure 3. Power curves for unphased implementations of XP-EHH (A and C) and XP-nSL (B and D) under demographic histories Demo 4 (A and B), and Demo 5 (C and D). s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

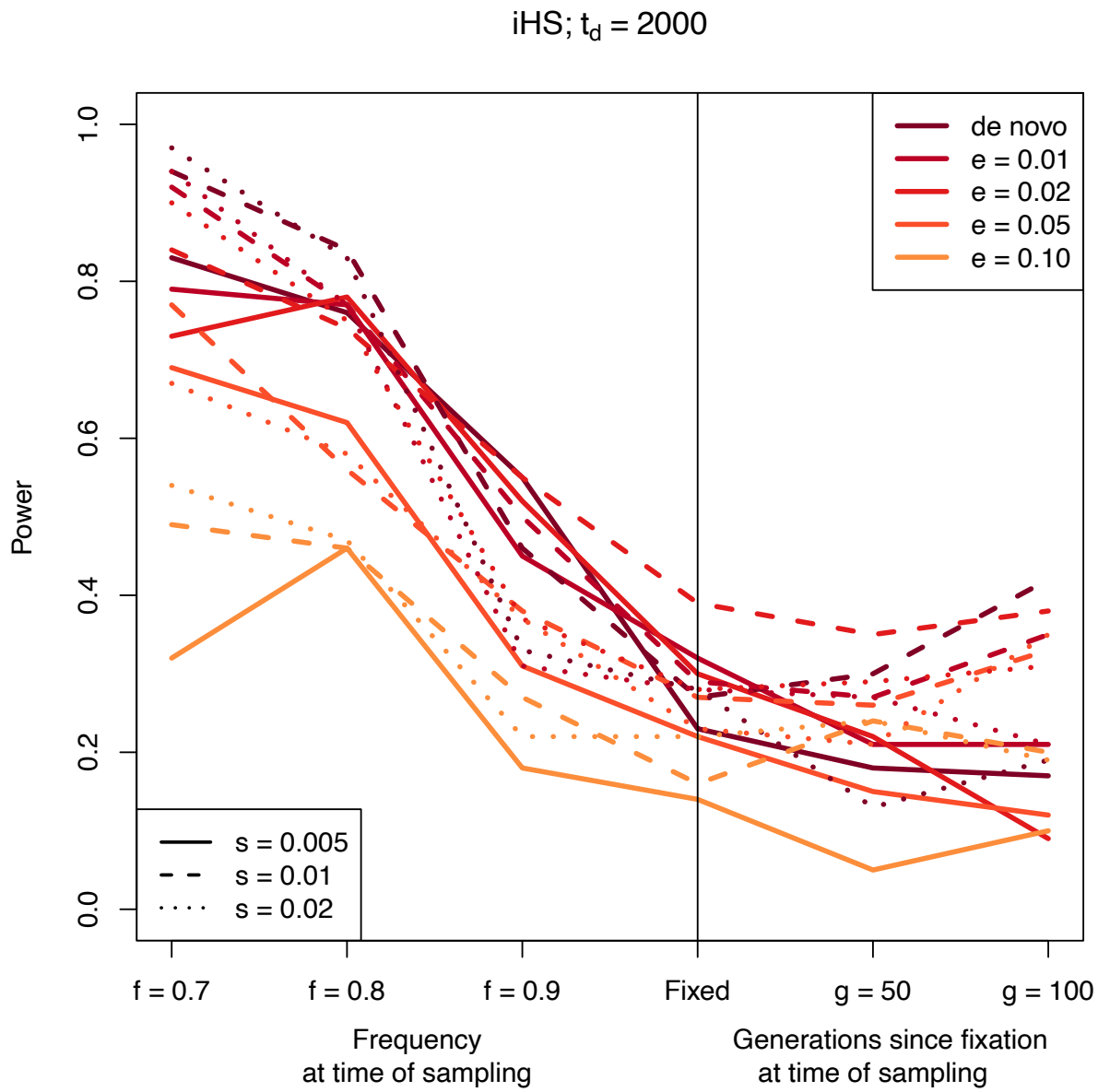

Figure S1. Demo 1 iHS $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

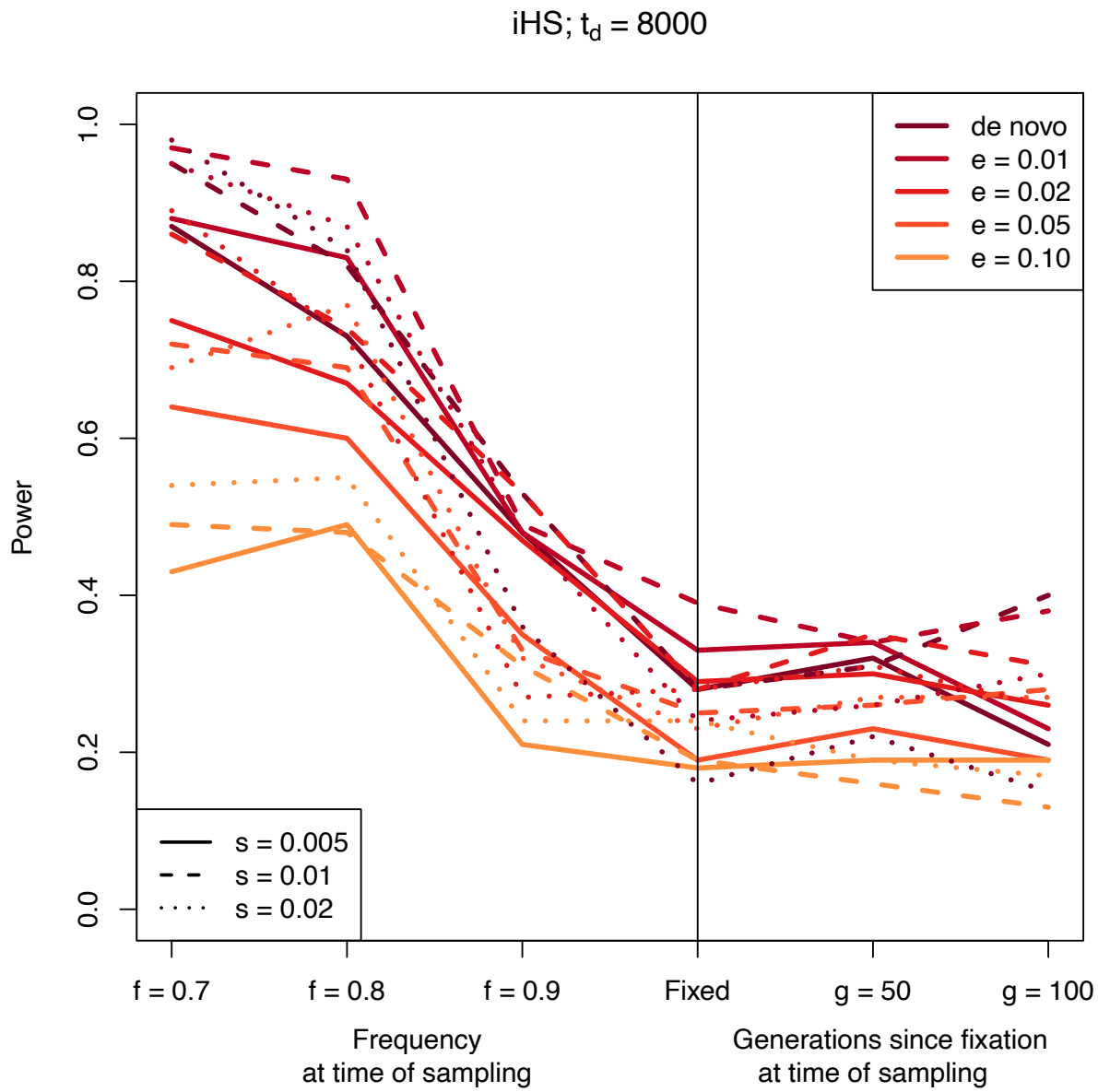

Figure S2. Demo 1 iHS $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

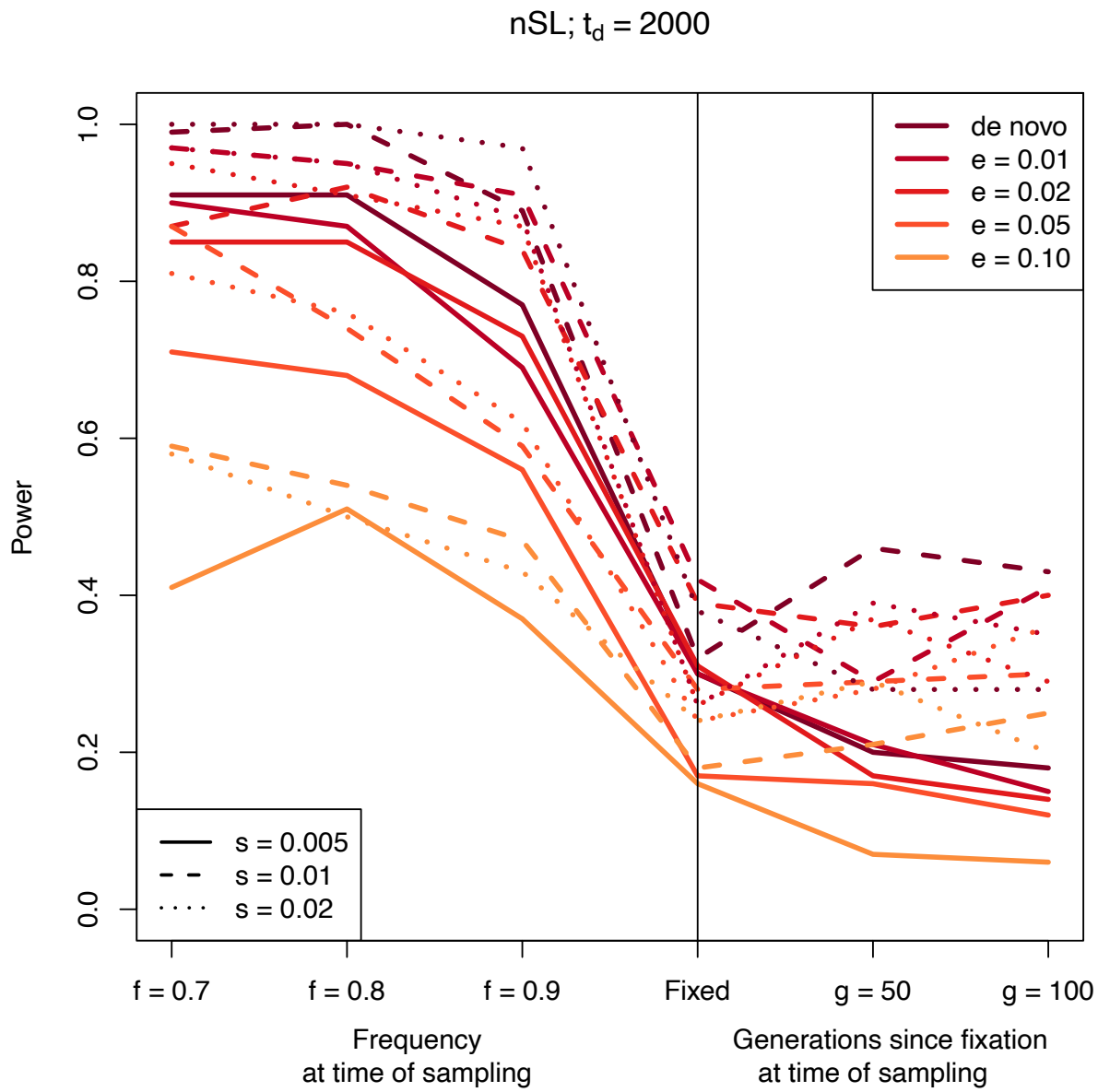

Figure S3. Demo 1 nSL $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

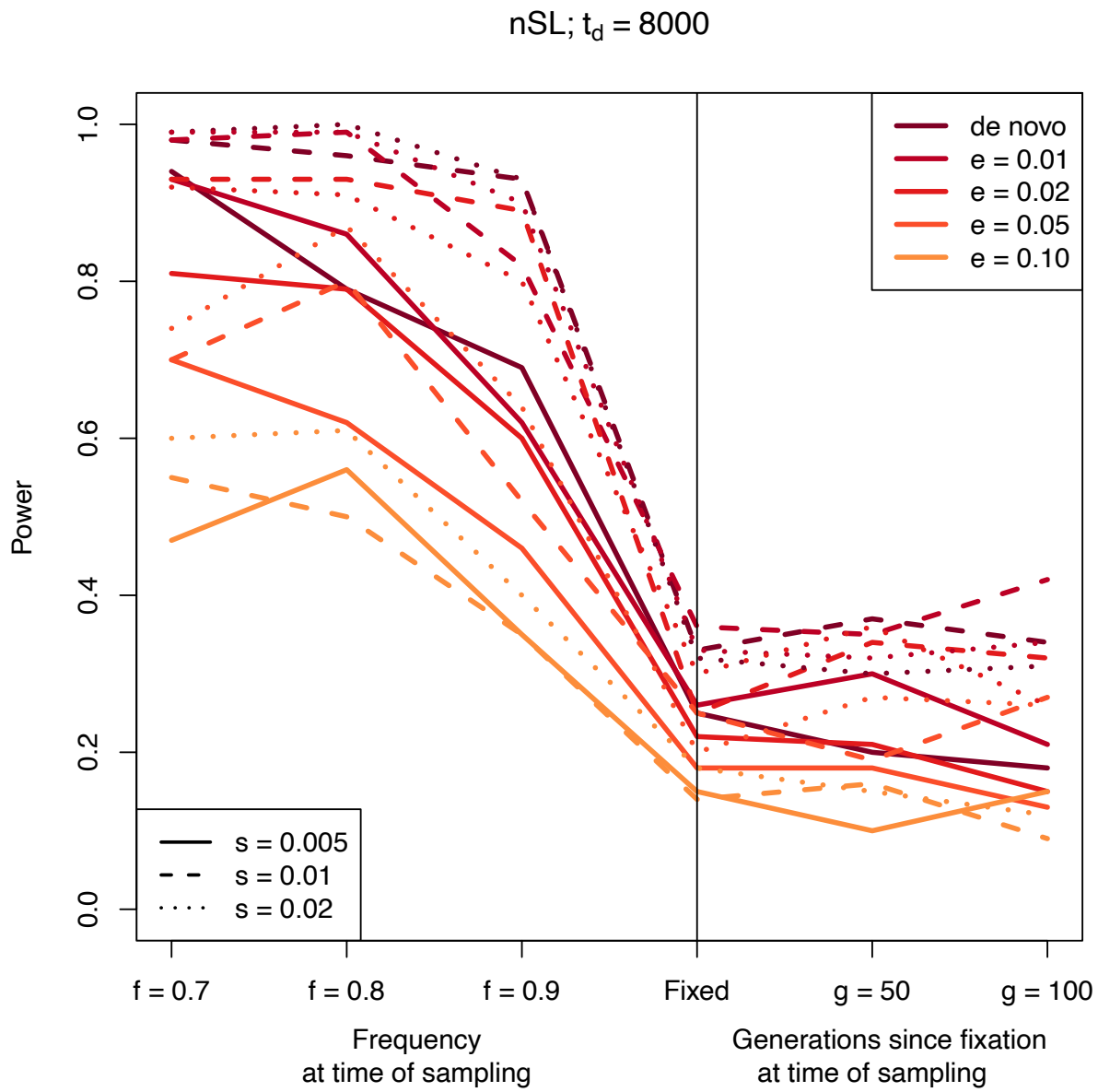

Figure S4. Demo 1 nSL $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

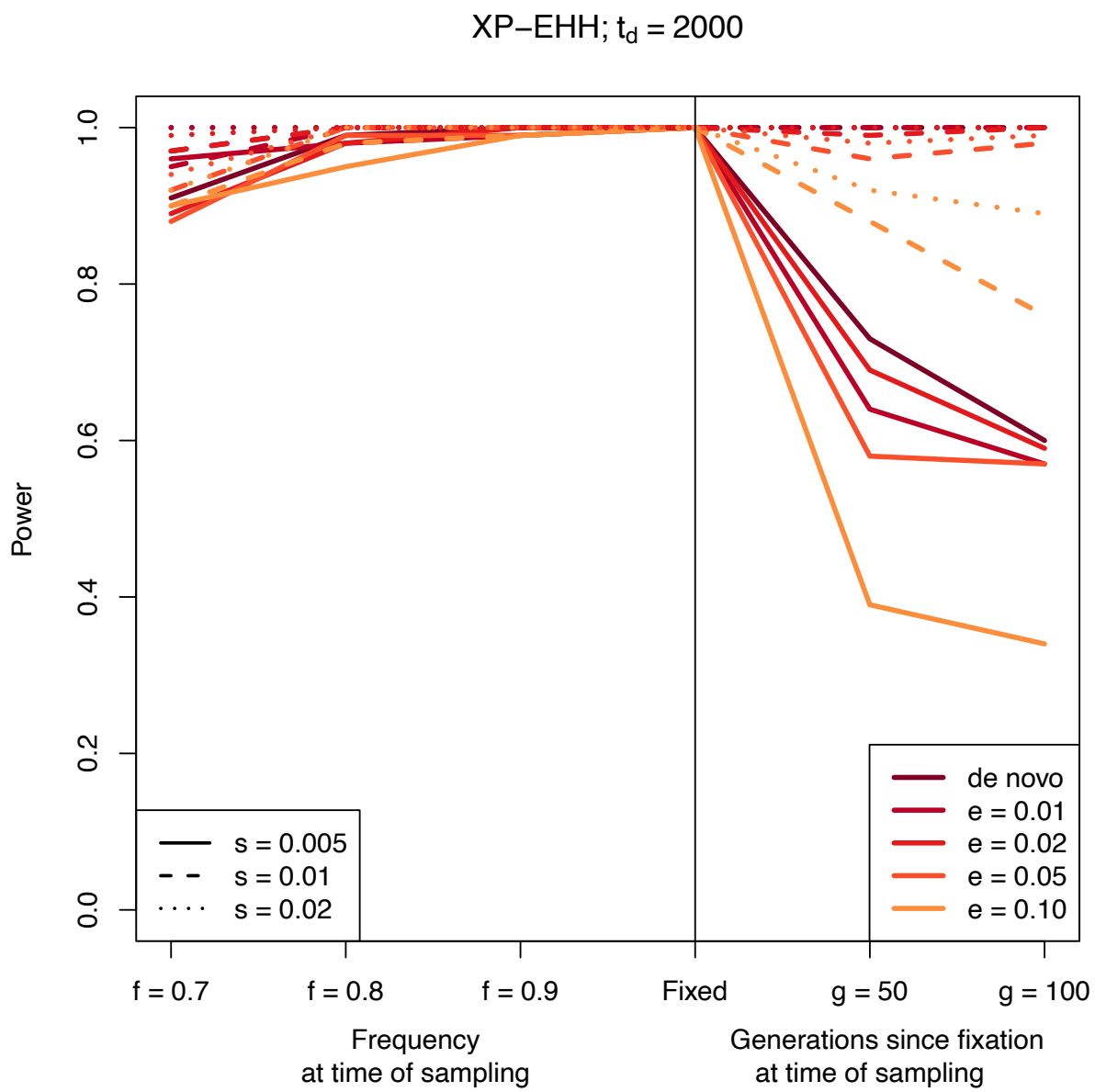

Figure S5. Demo 1 XP-EHH $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

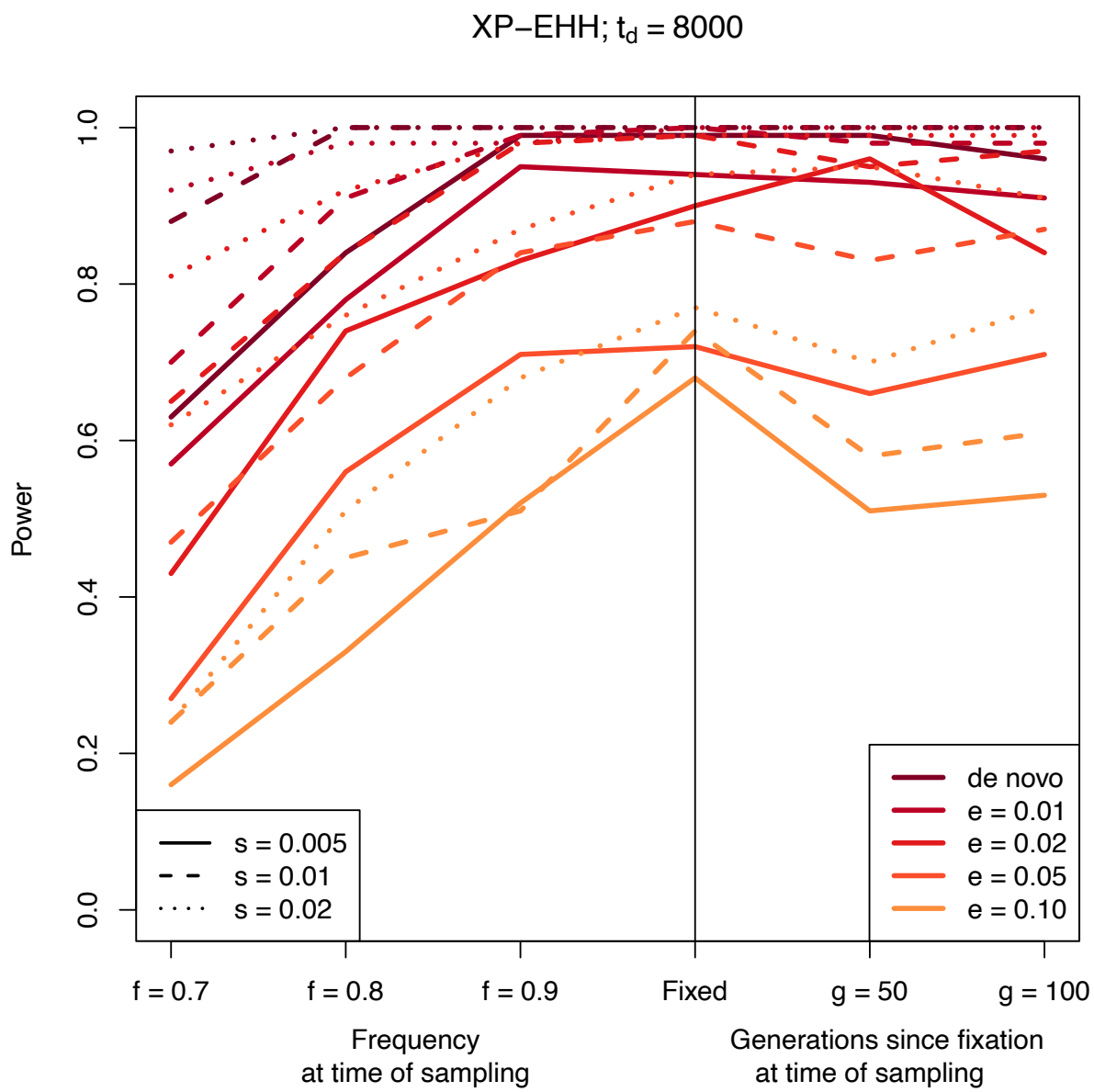

Figure S6. Demo 1 XP-EHH $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

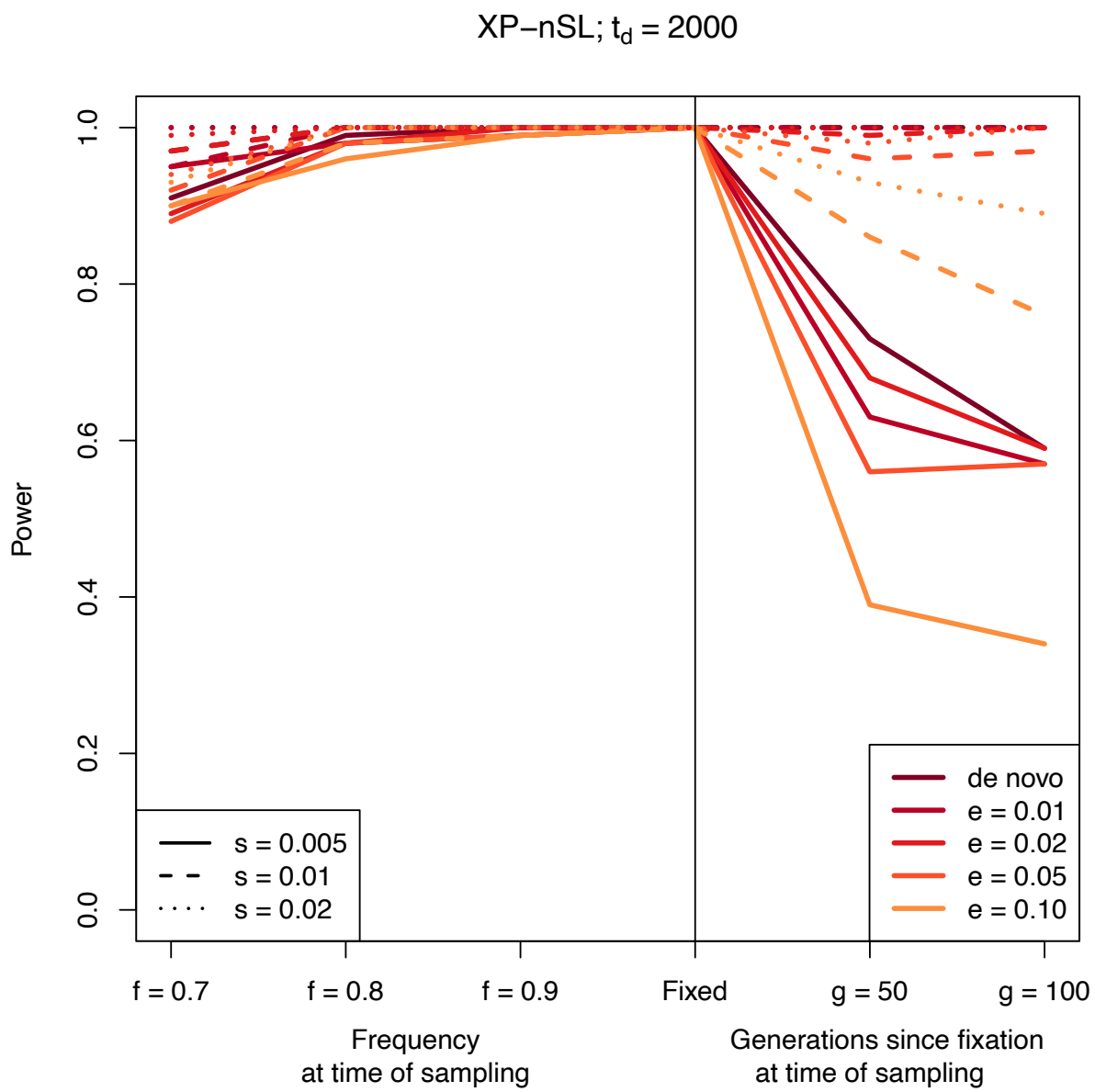

Figure S7. Demo 1 XP-nSL $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

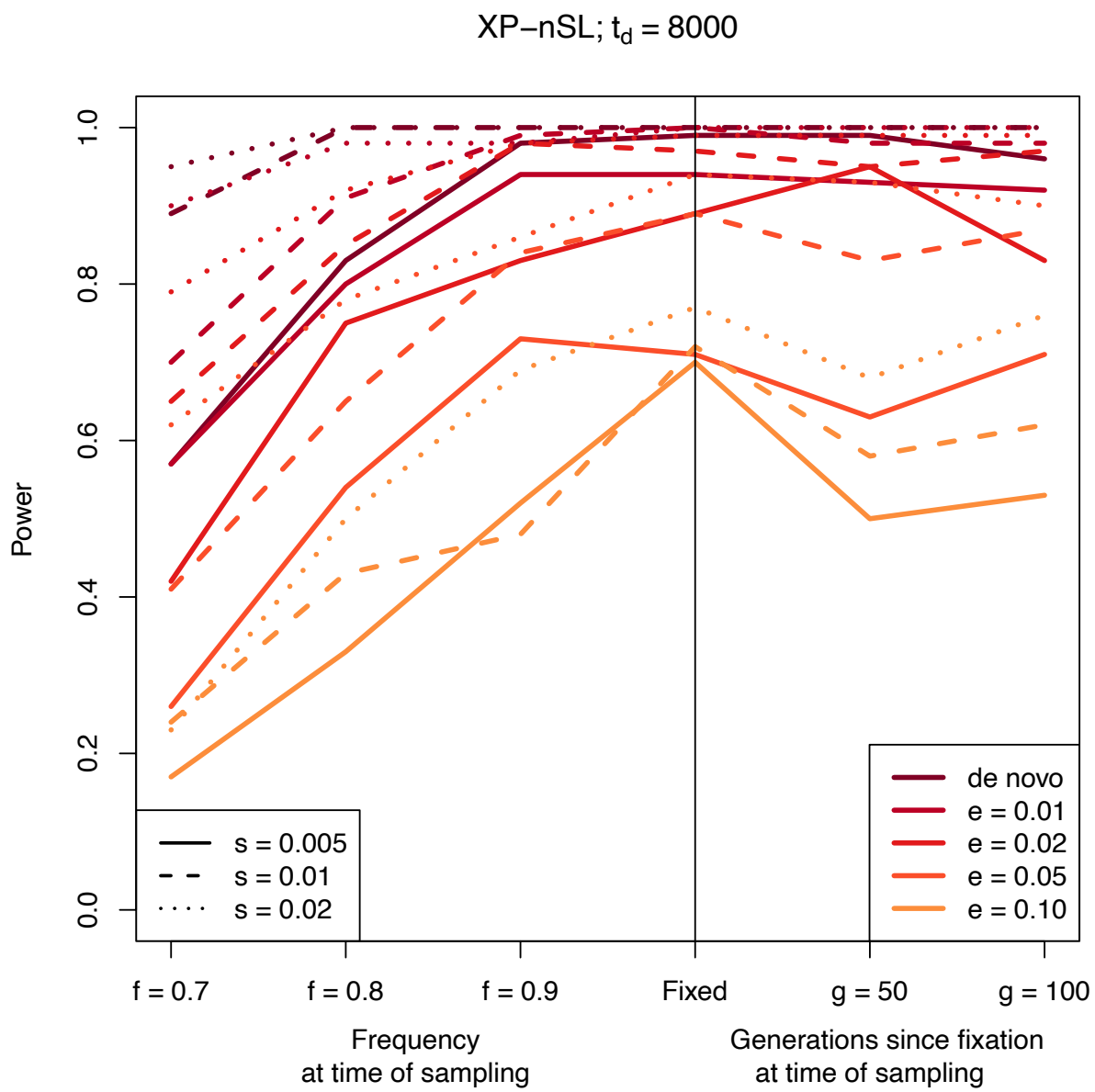

Figure S8. Demo 1 XP-nSL $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

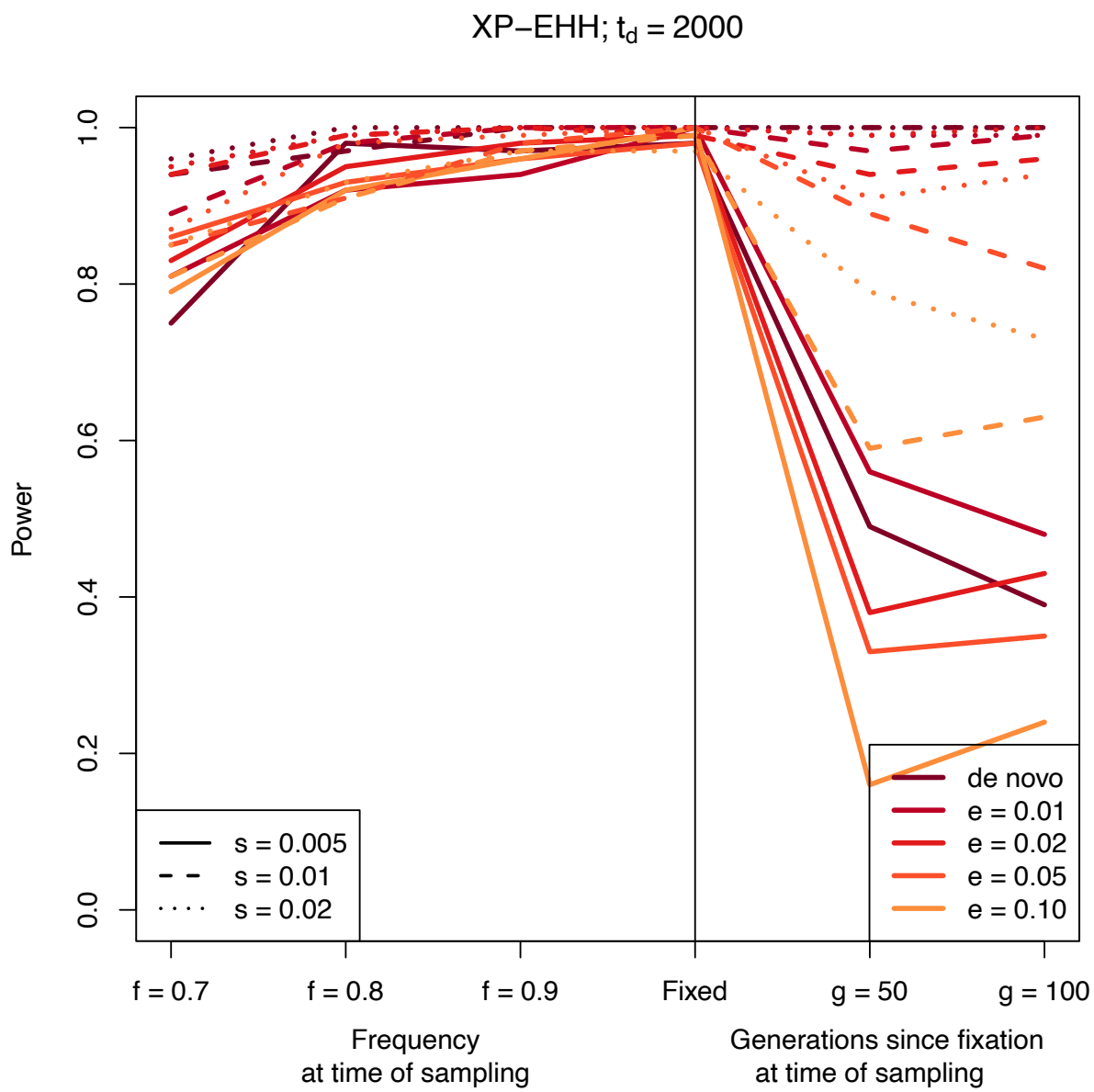

Figure S9. Demo 2 XP-EHH $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

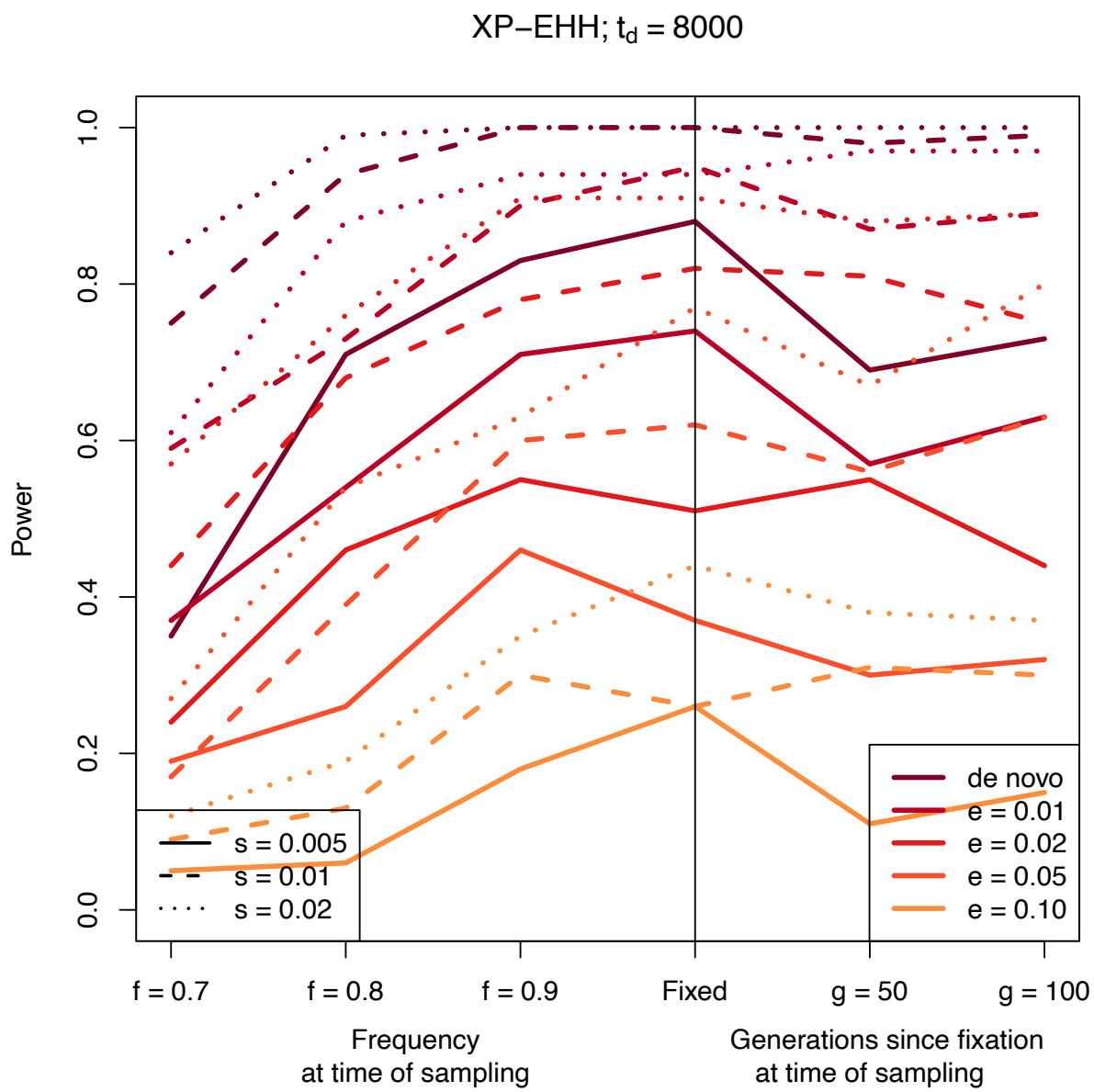

Figure S10. Demo 2 XP-EHH $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

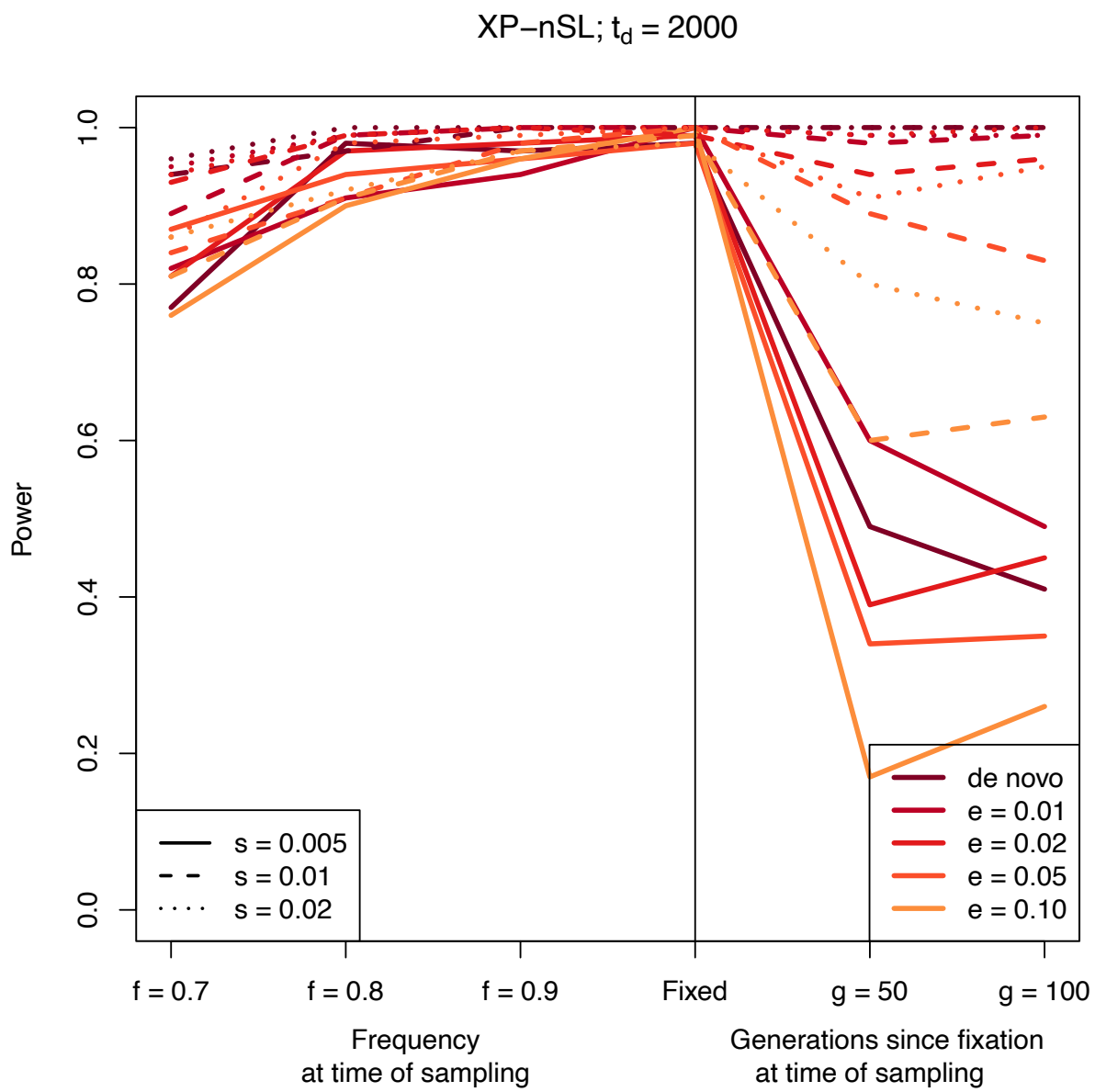

Figure S11. Demo 2 XP-nSL $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

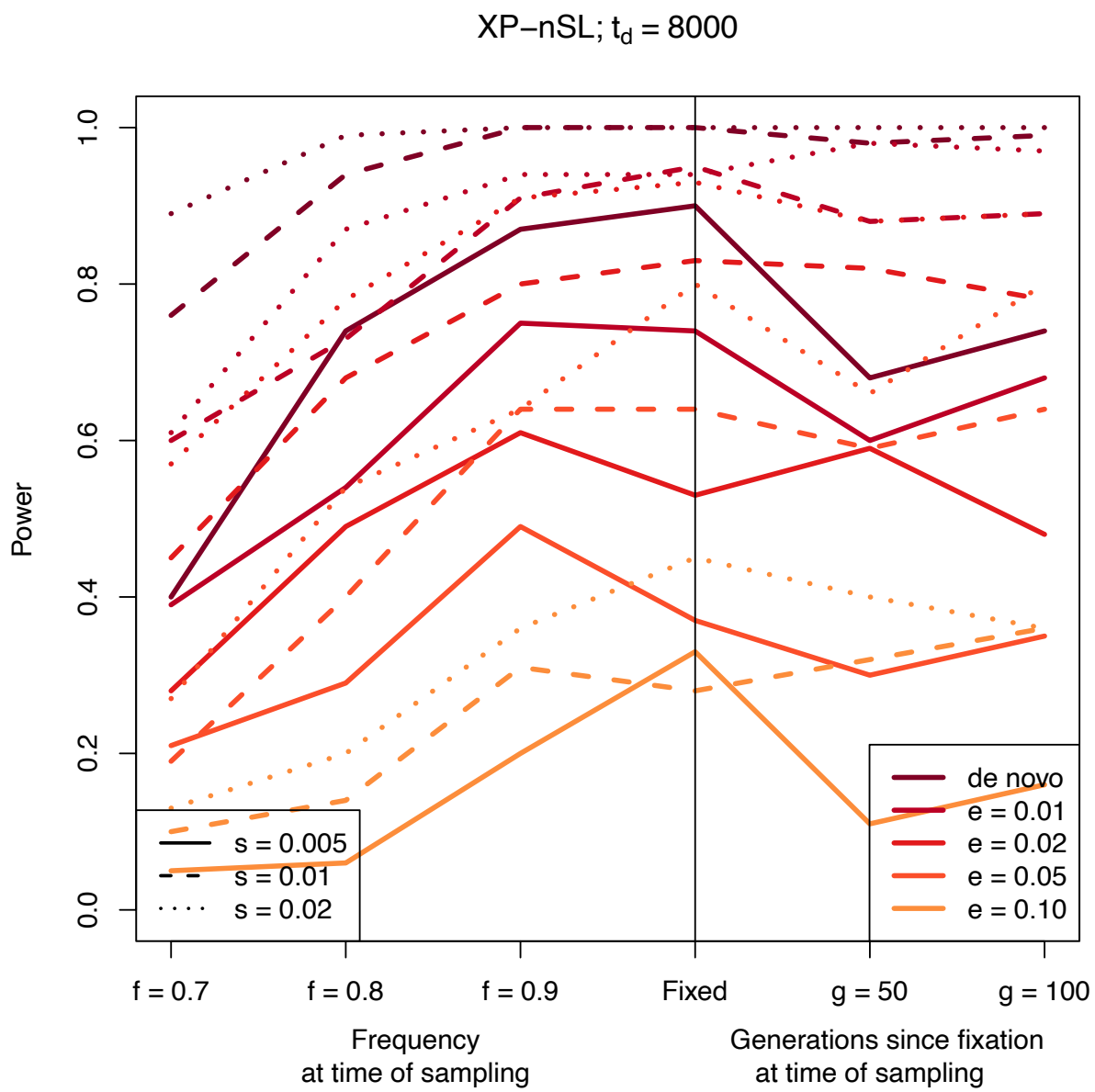

Figure S12. Demo 2 XP-nSL $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

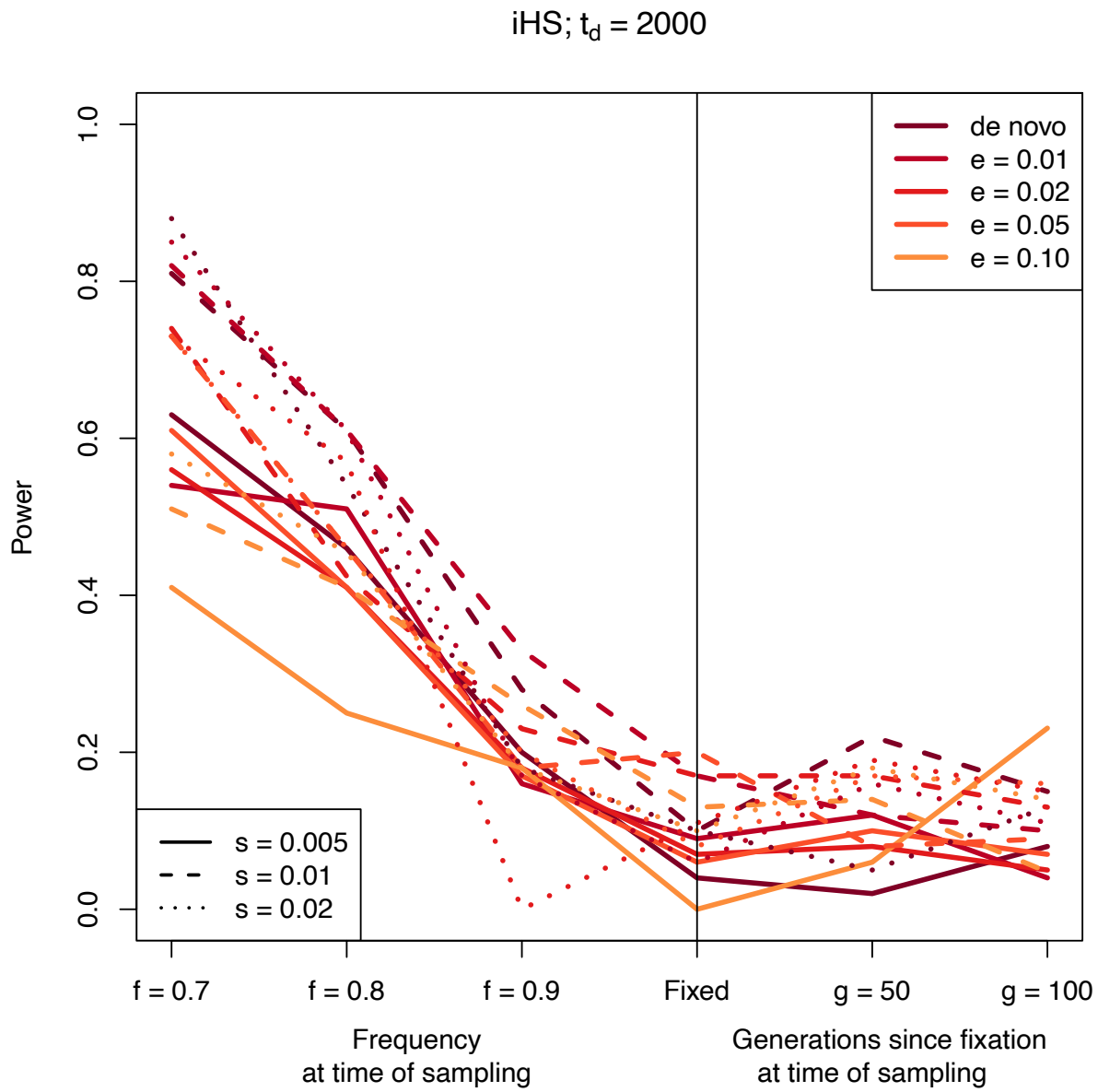

Figure S13. Demo 3 iHS $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

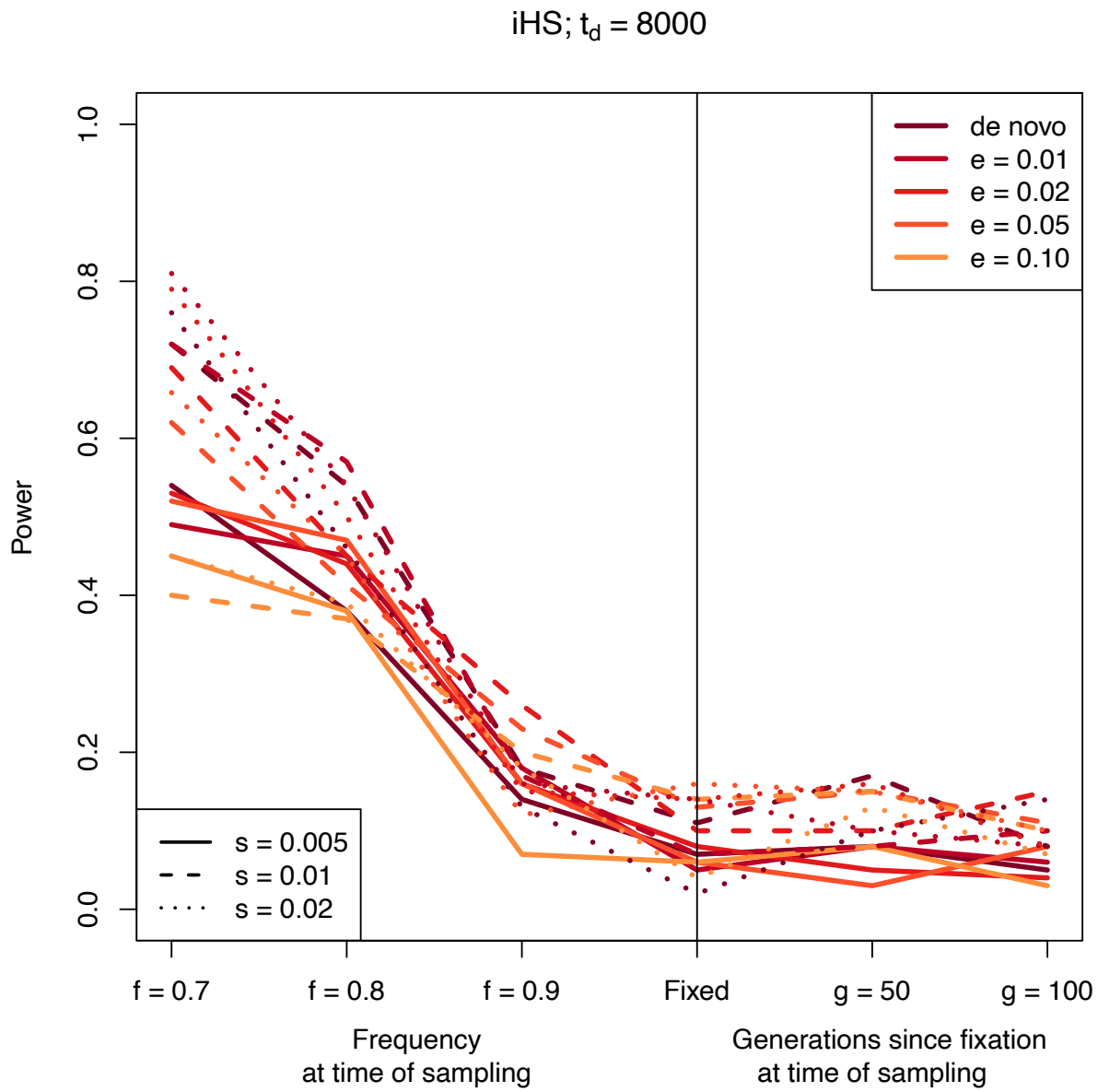

Figure S14. Demo 3 iHS $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

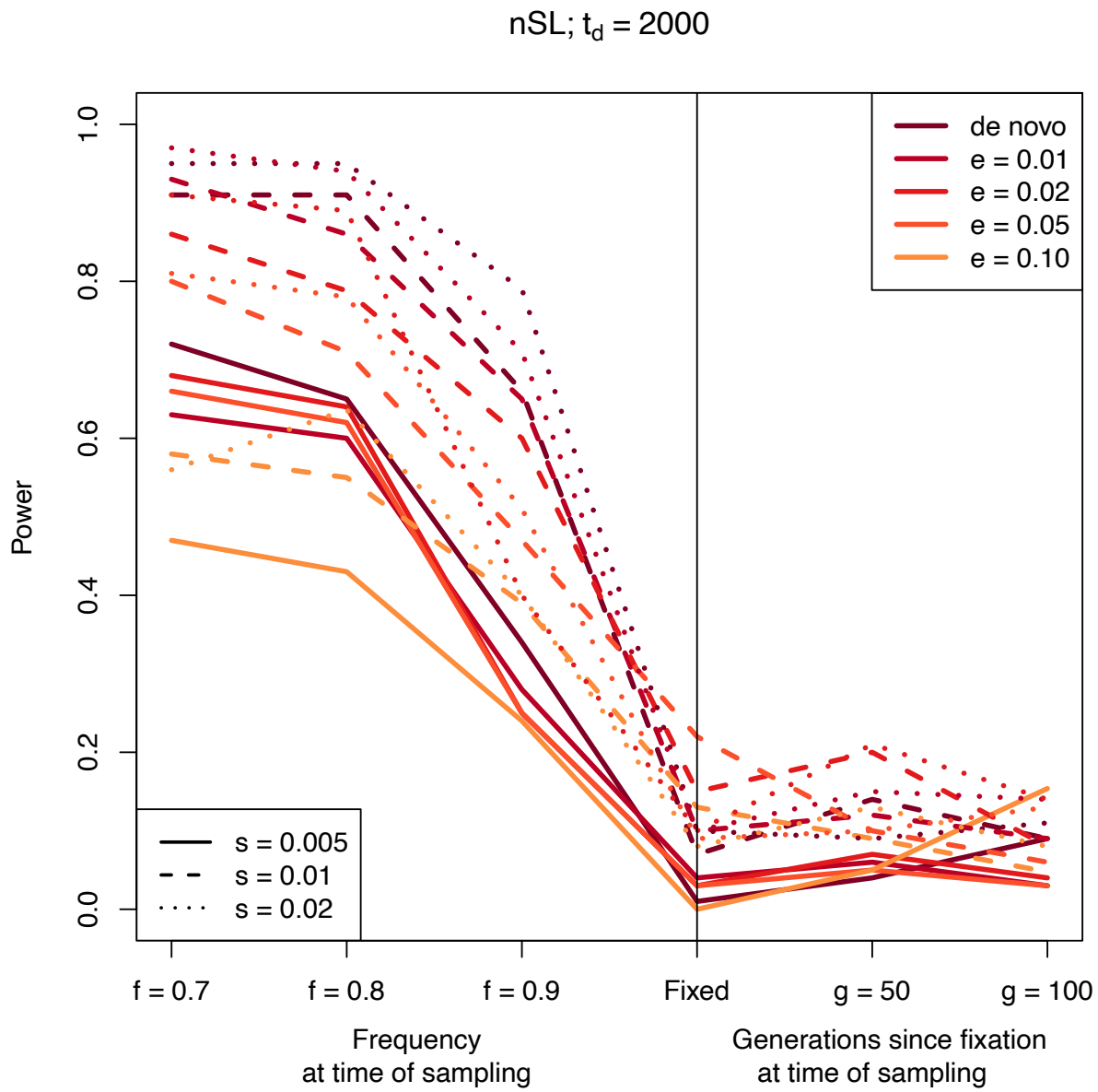

Figure S15. Demo 3 nSL $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

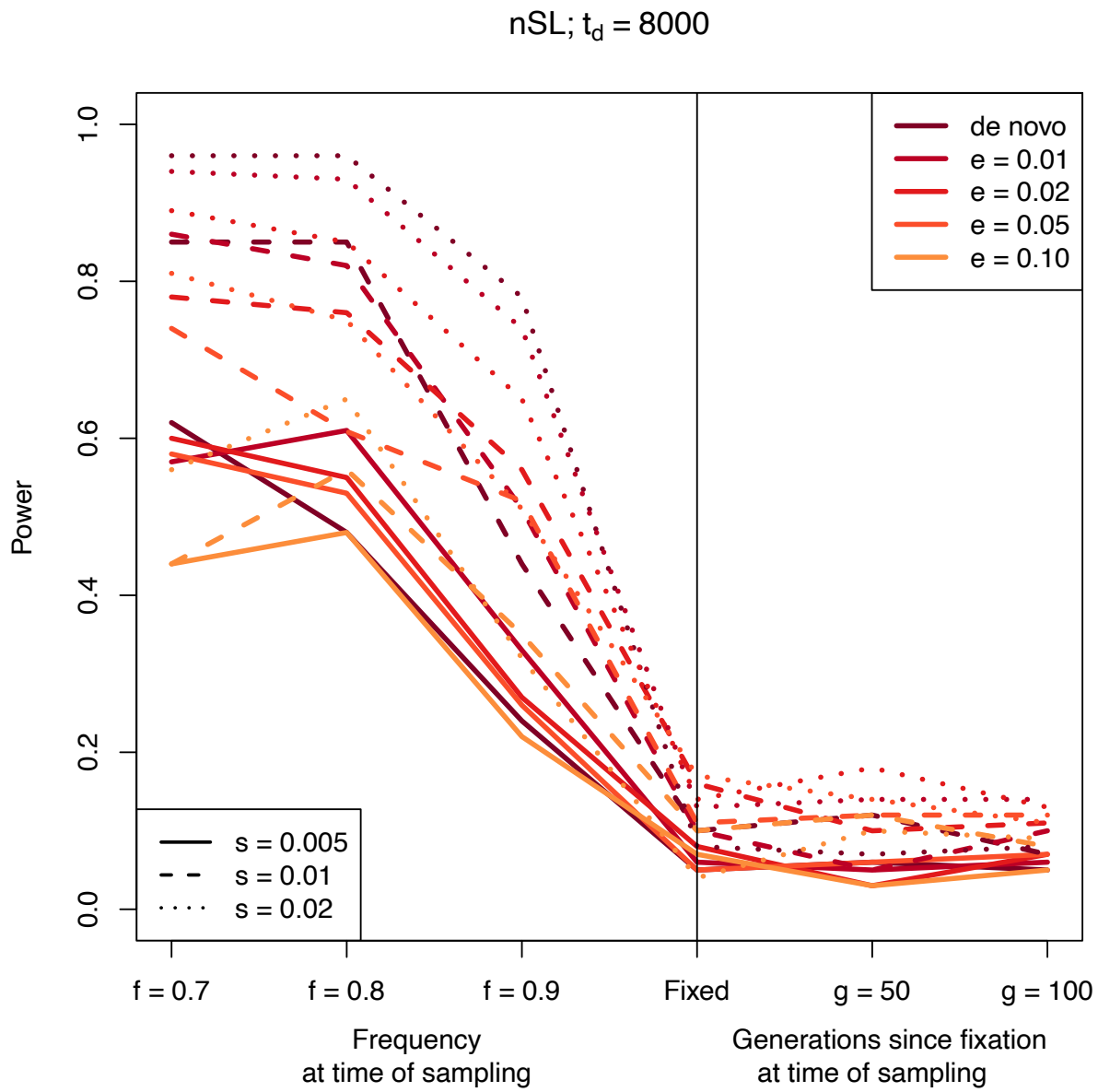

Figure S16. Demo 3 nSL $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

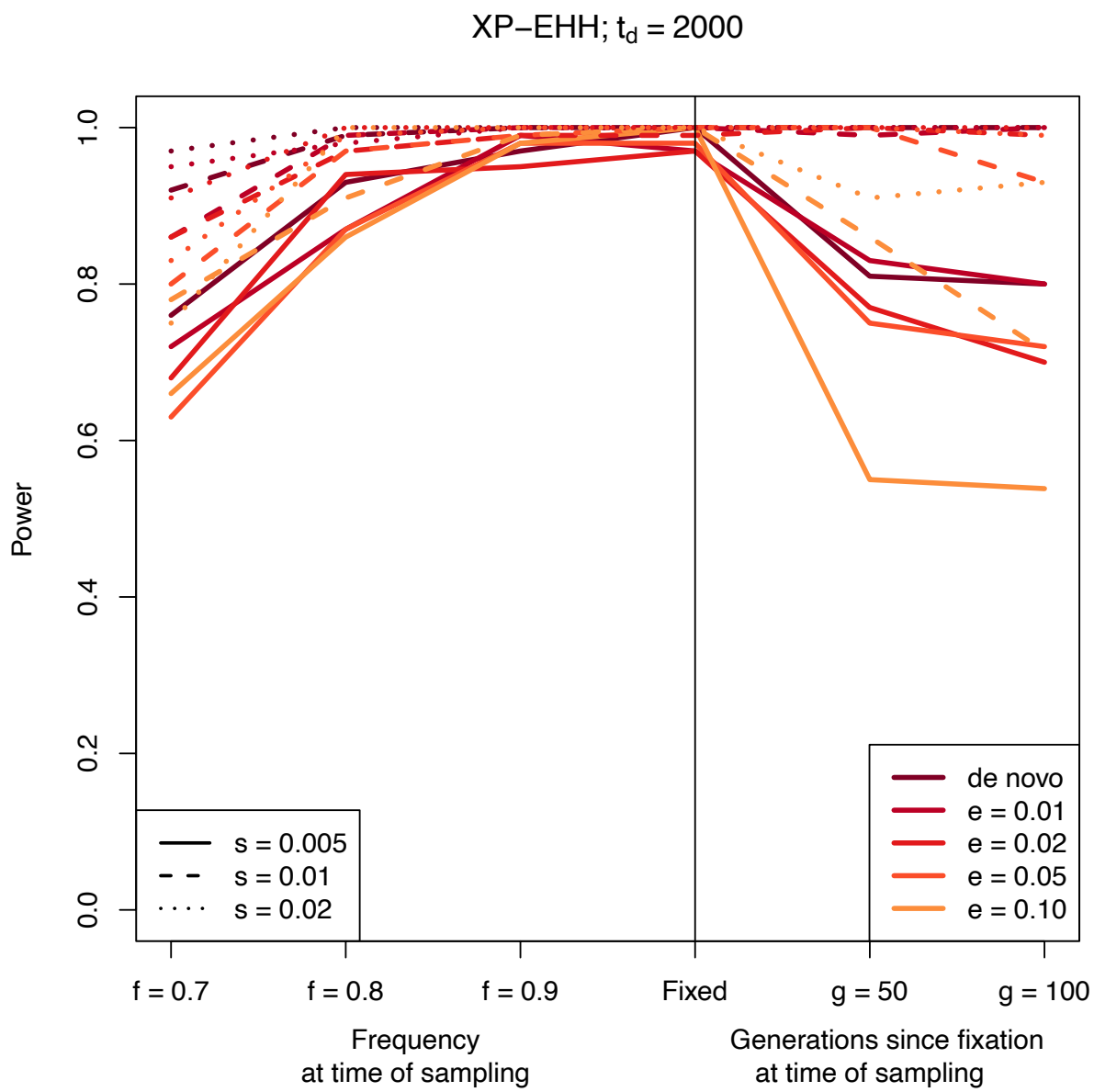

Figure S17. Demo 3 XP-EHH $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

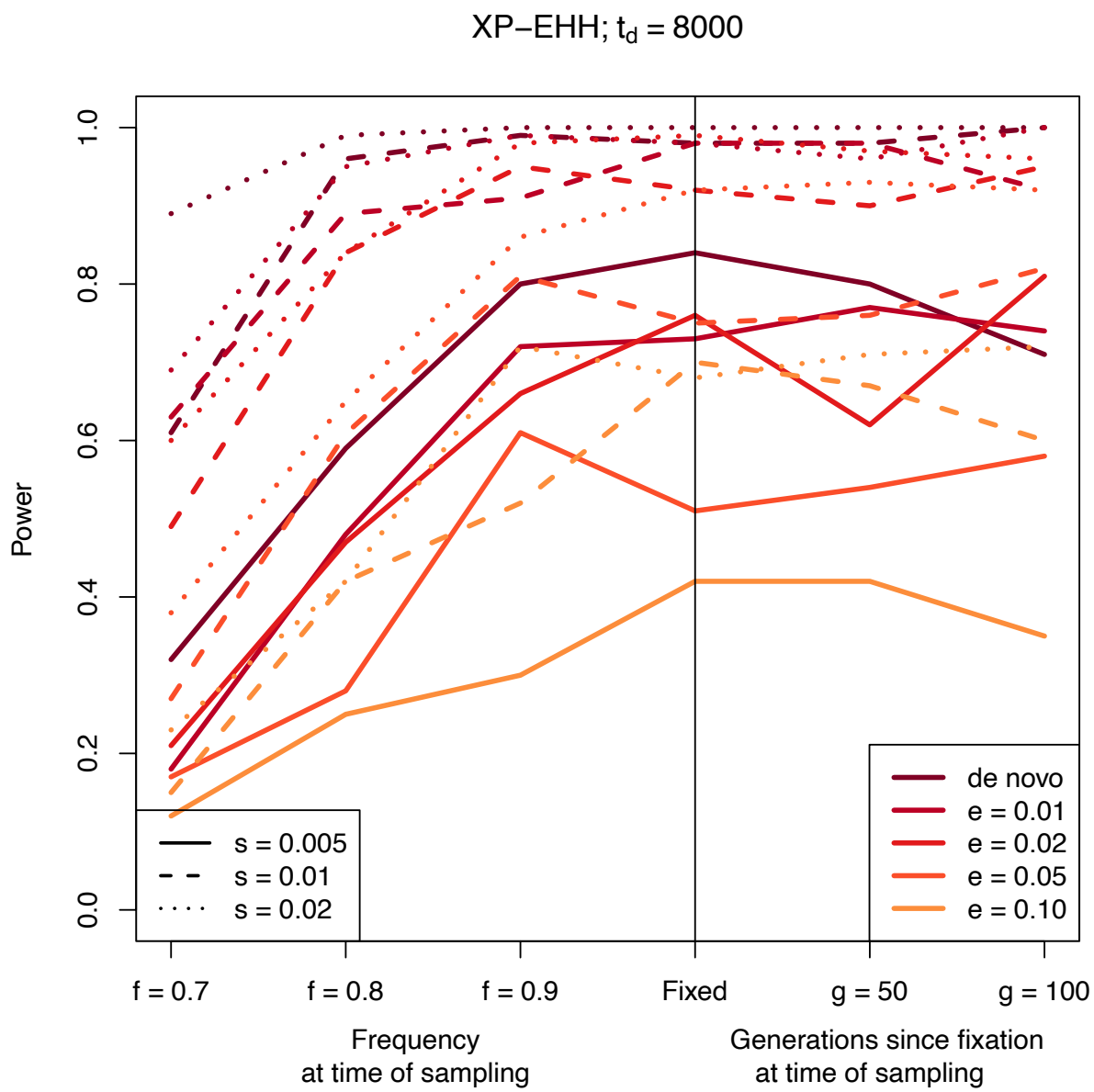

Figure S18. Demo 3 XP-EHH $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

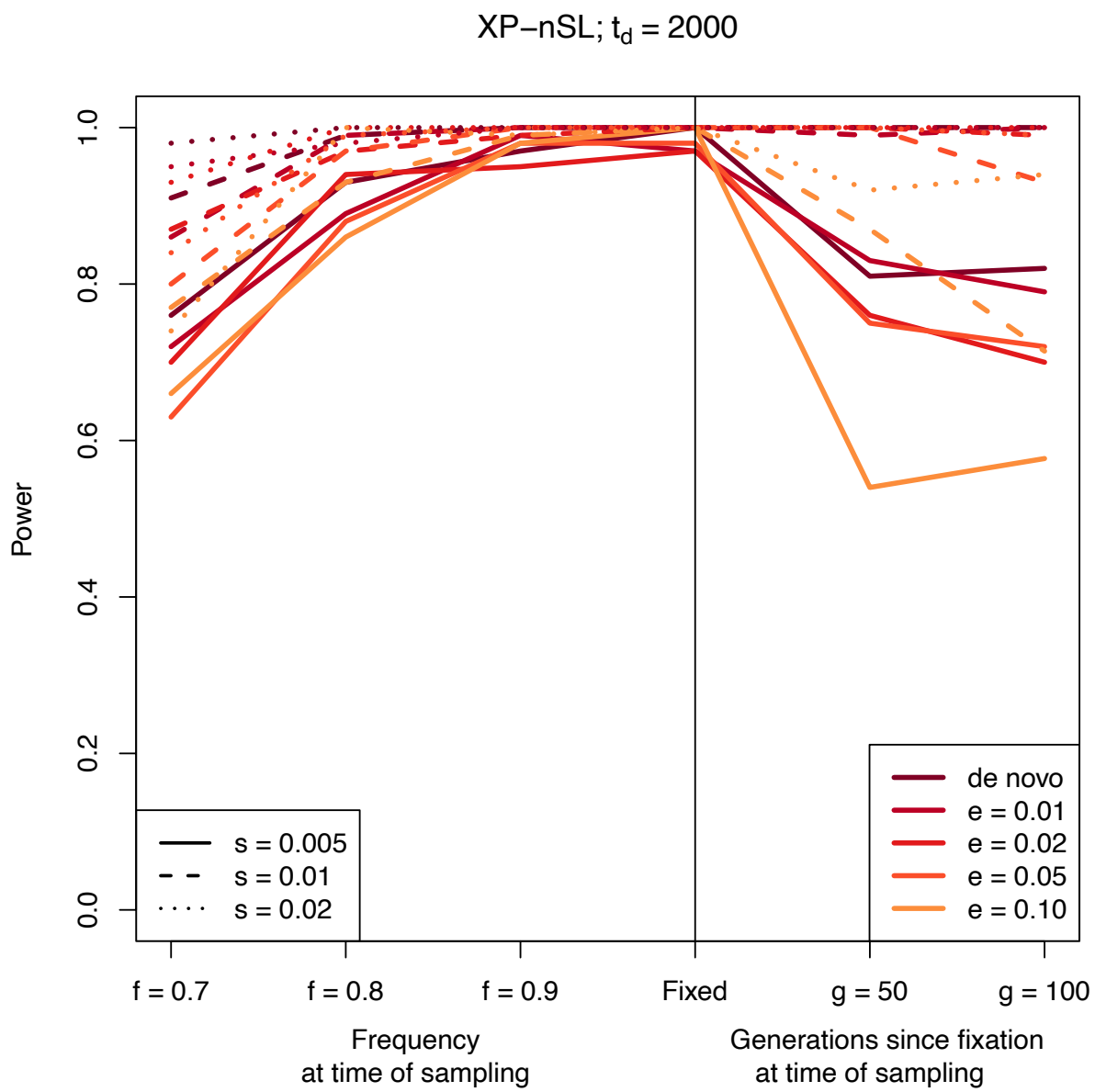

Figure S19. Demo 3 XP-nSL $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

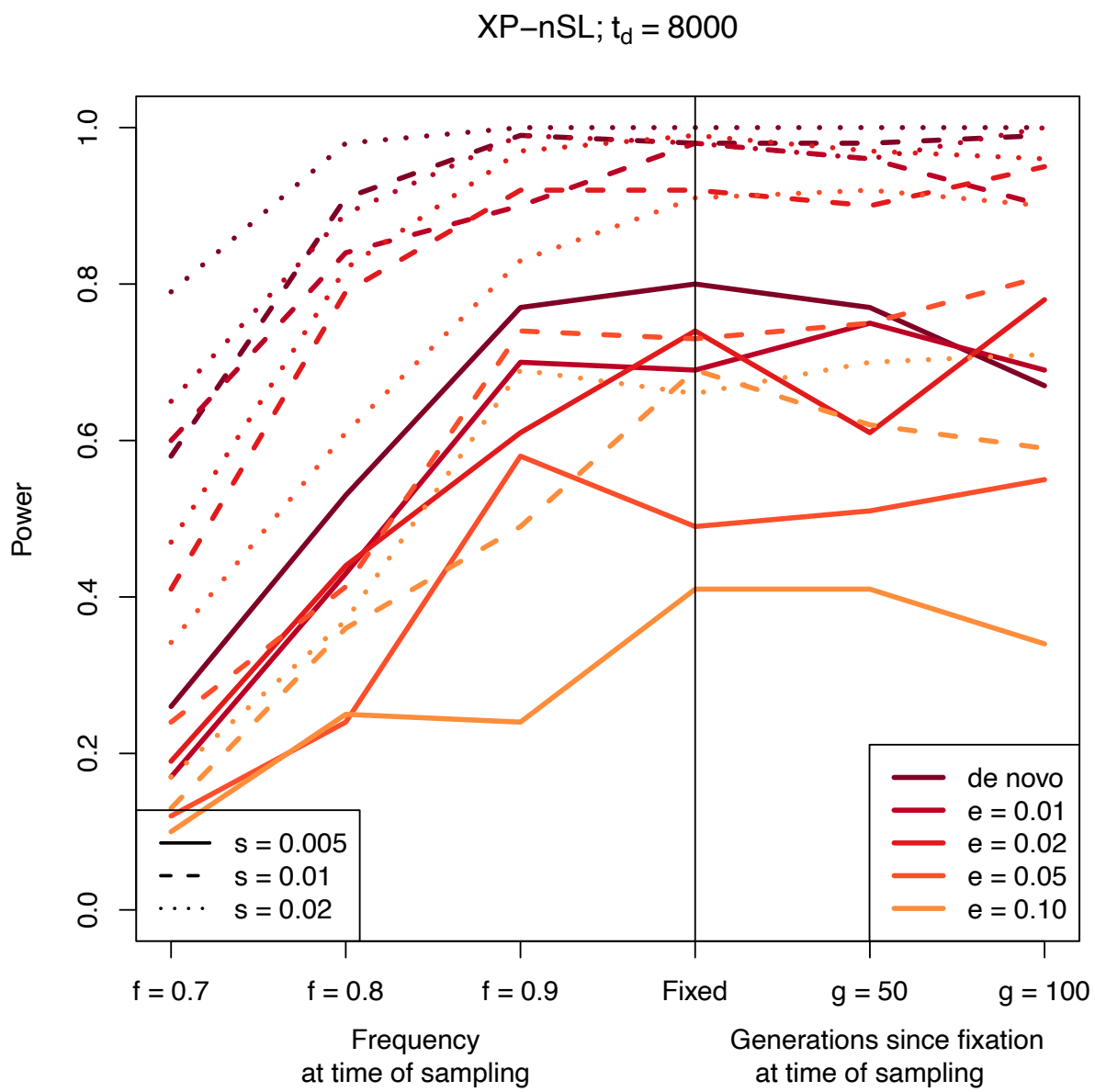

Figure S20. Demo 3 XP-nSL $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

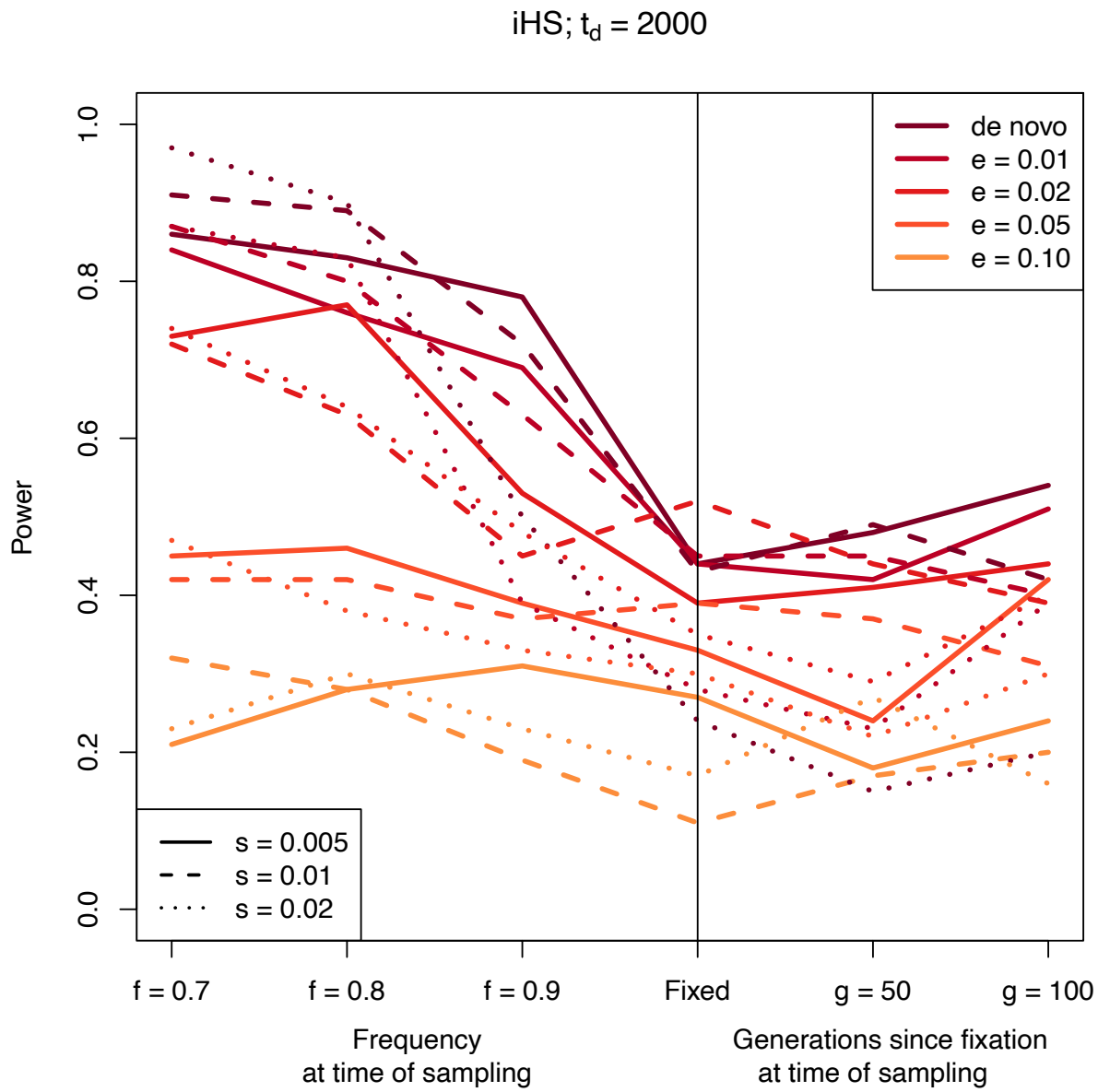

Figure S21. Demo 4 iHS $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

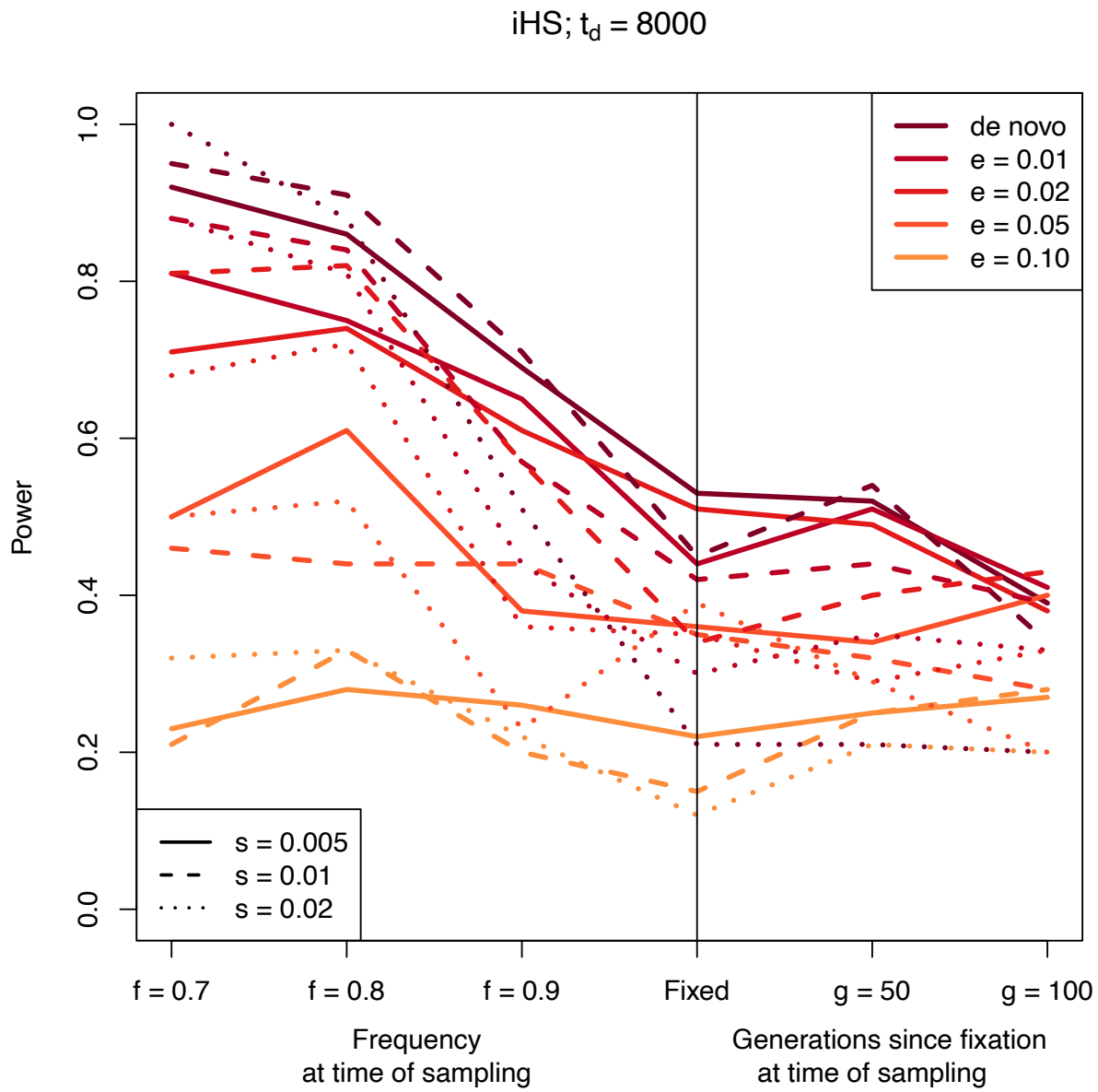

Figure S22. Demo 4 iHS $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

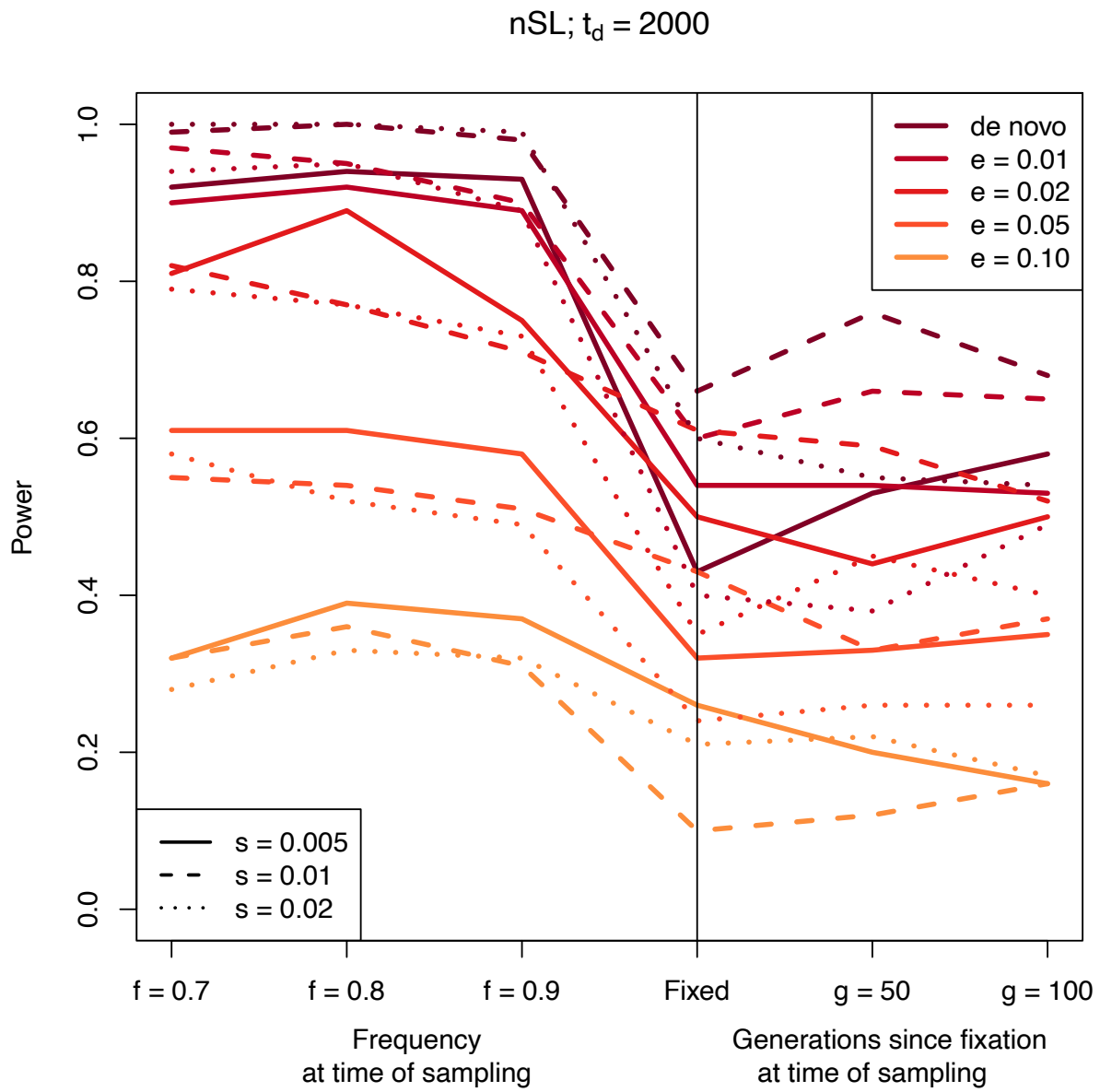

Figure S23. Demo 4 nSL $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

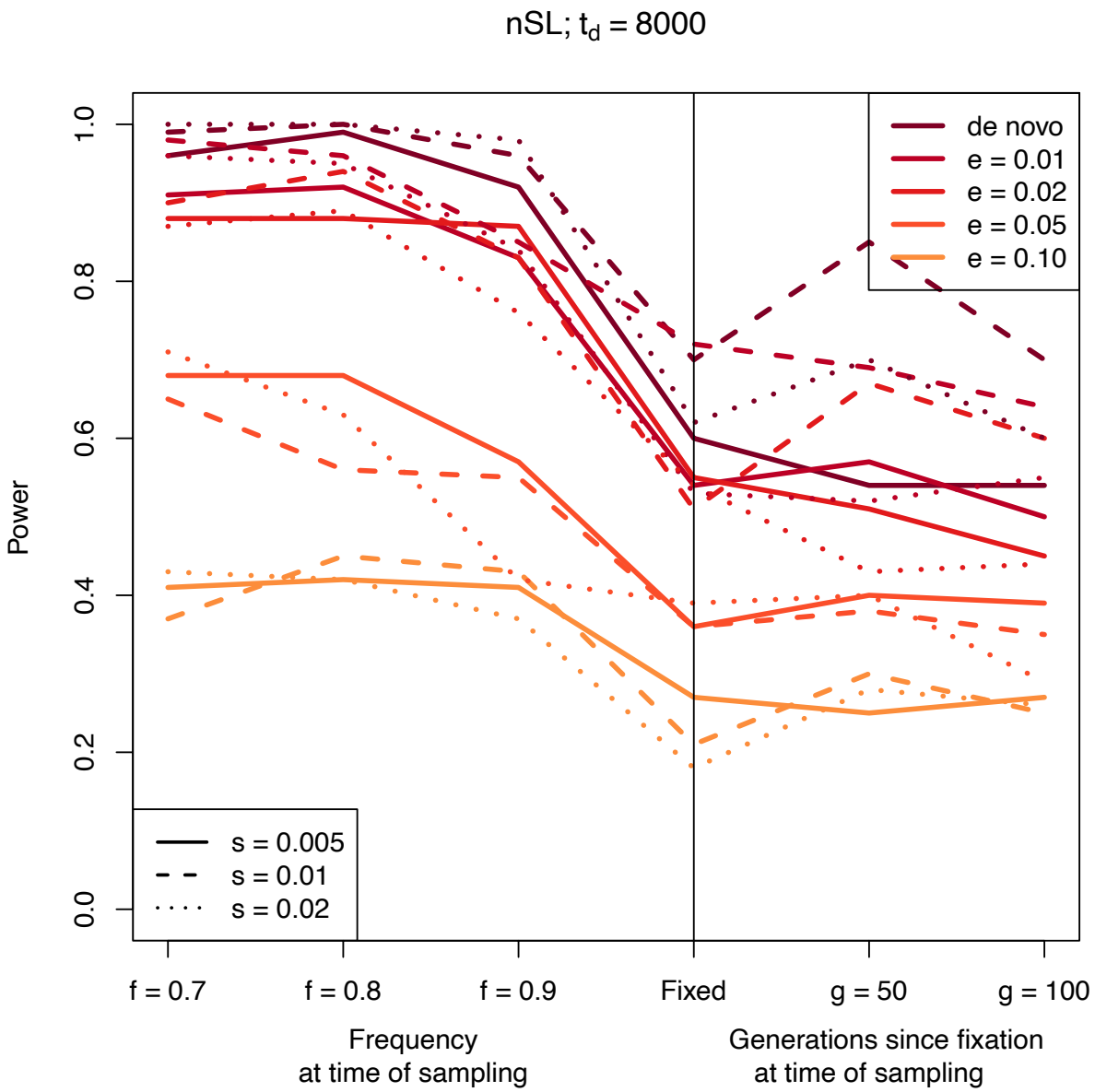

Figure S24. Demo 4 nSL $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

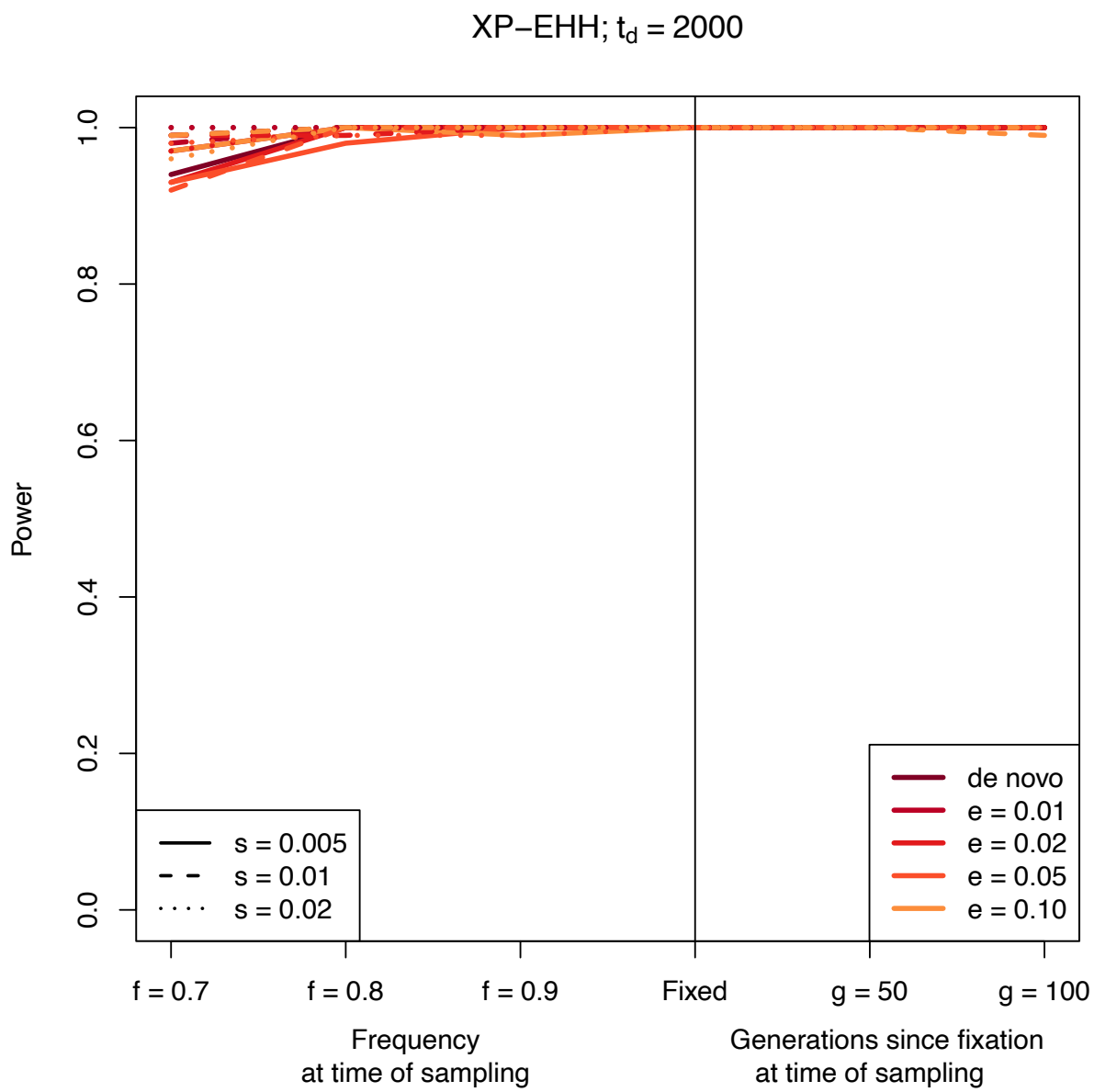

Figure S25. Demo 4 XP-EHH $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

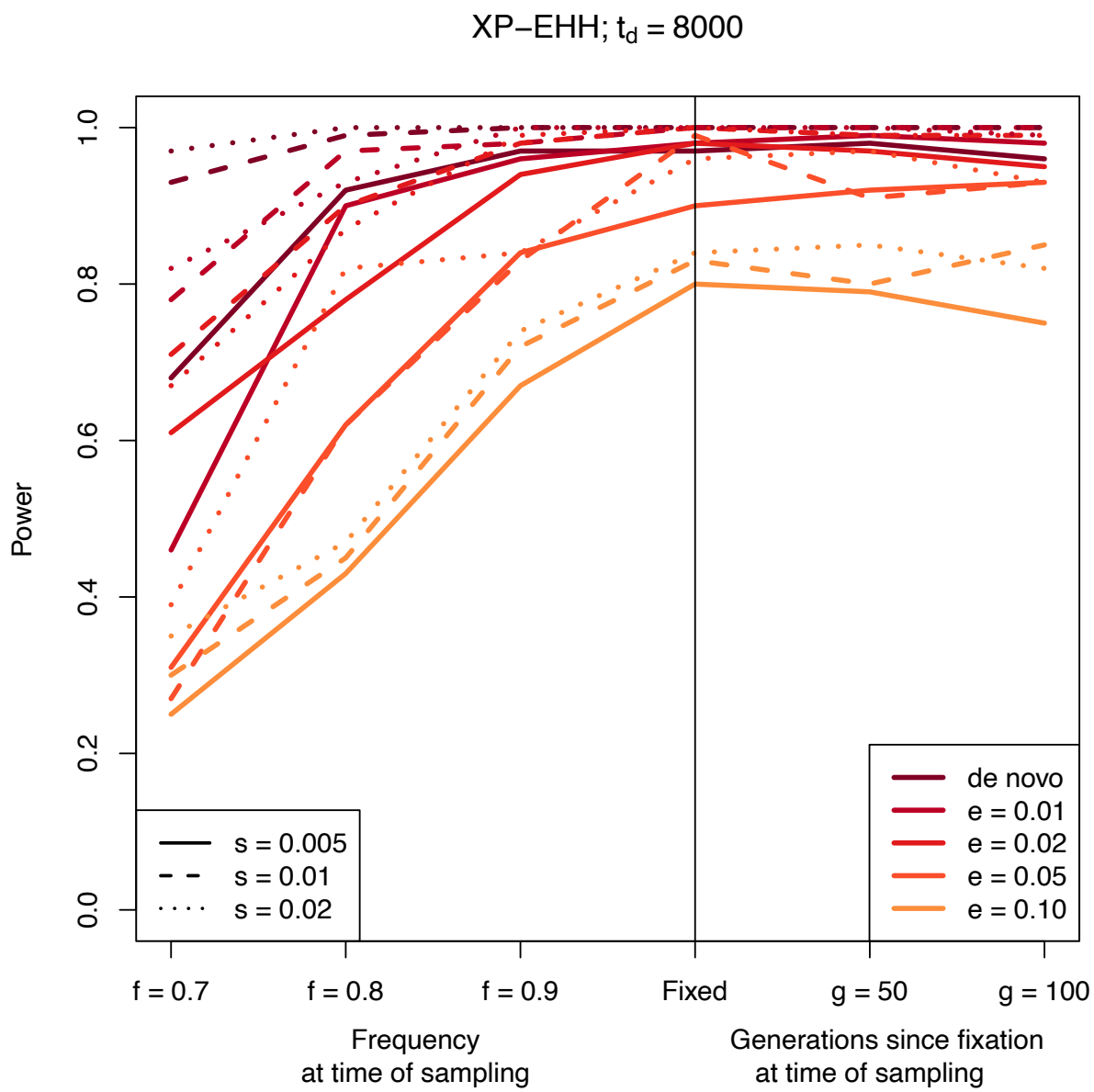

Figure S26. Demo 4 XP-EHH $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

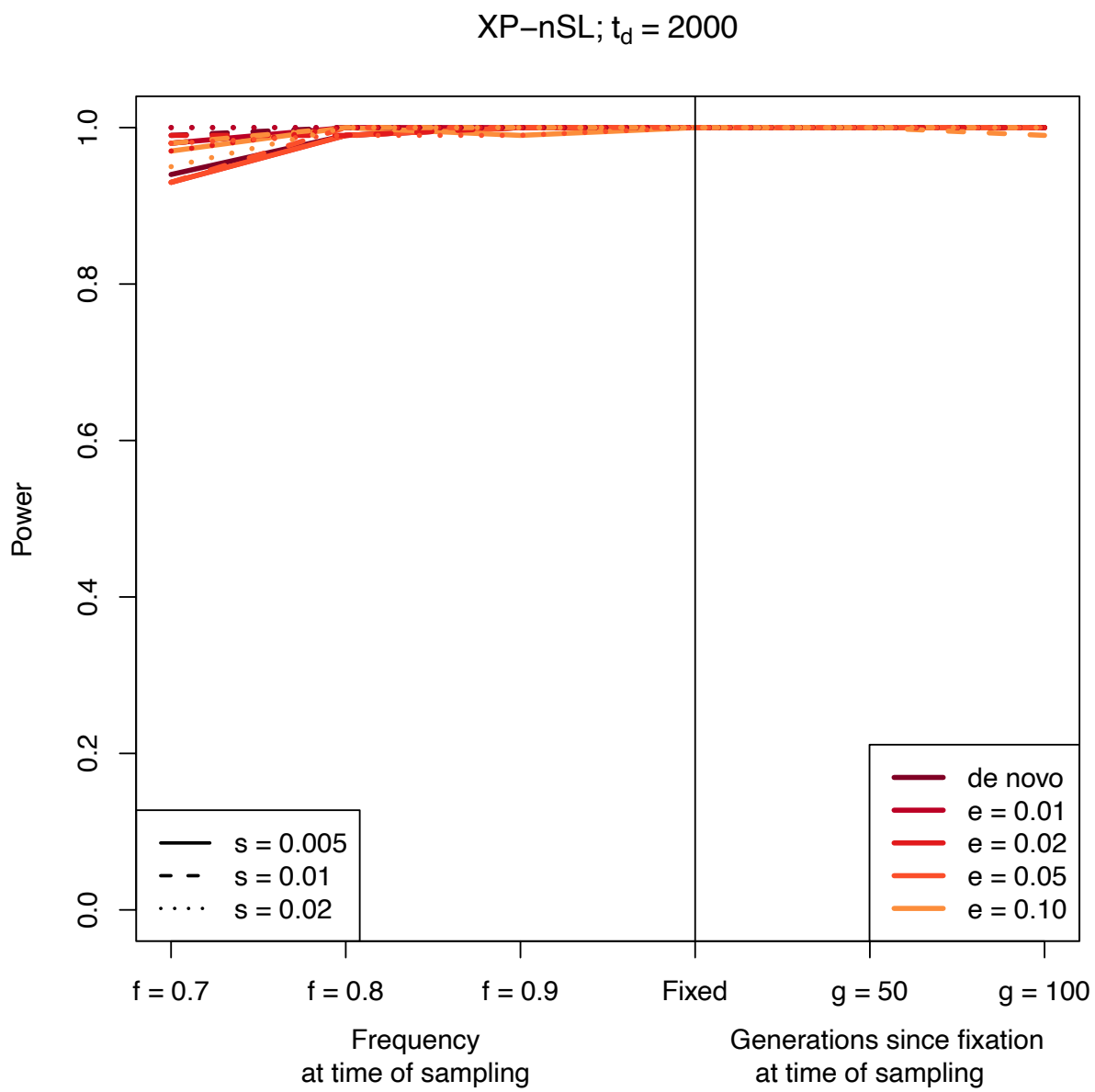

Figure S27. Demo 4 XP-nSL $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

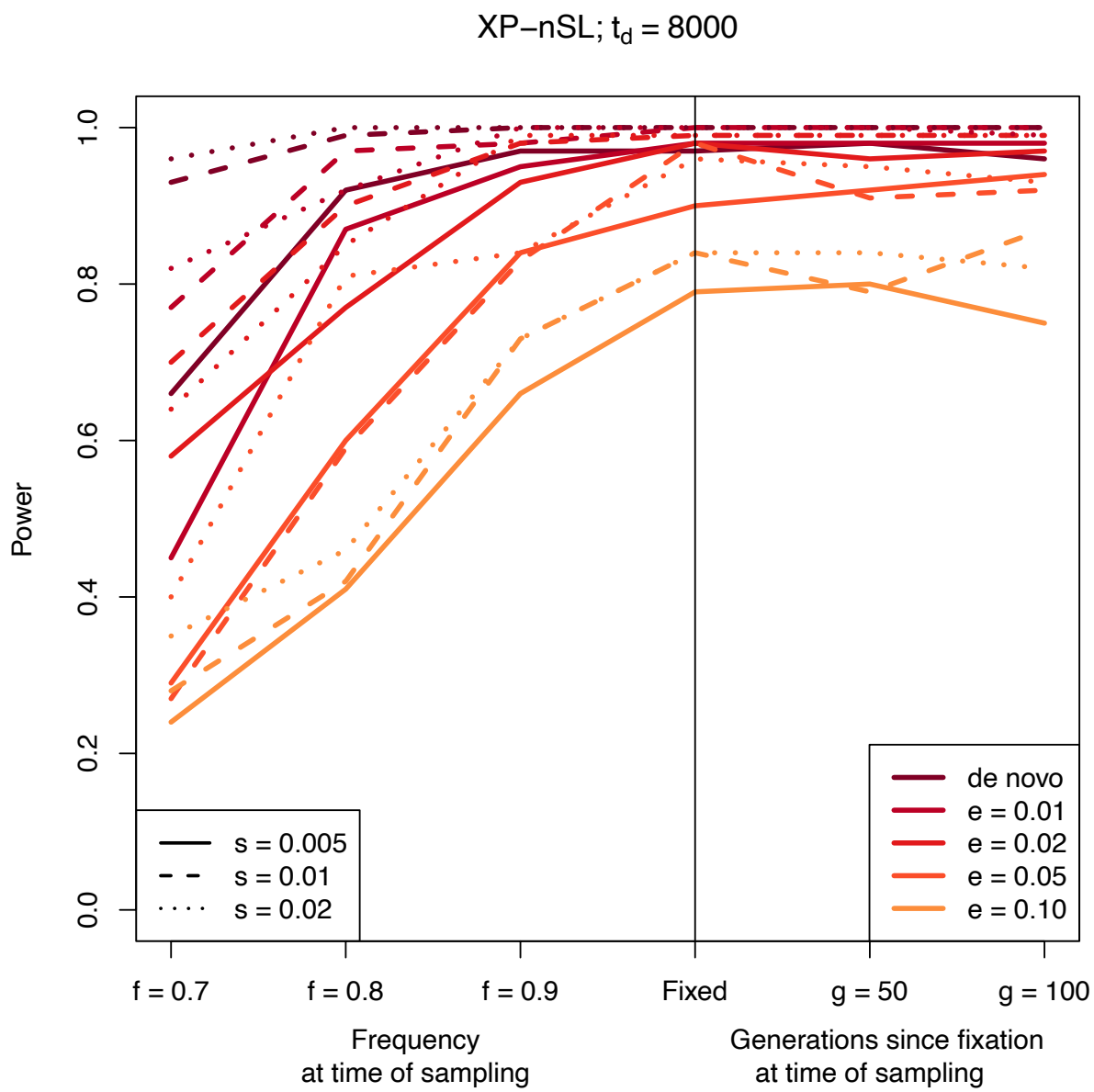

Figure S28. Demo 4 XP-nSL $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

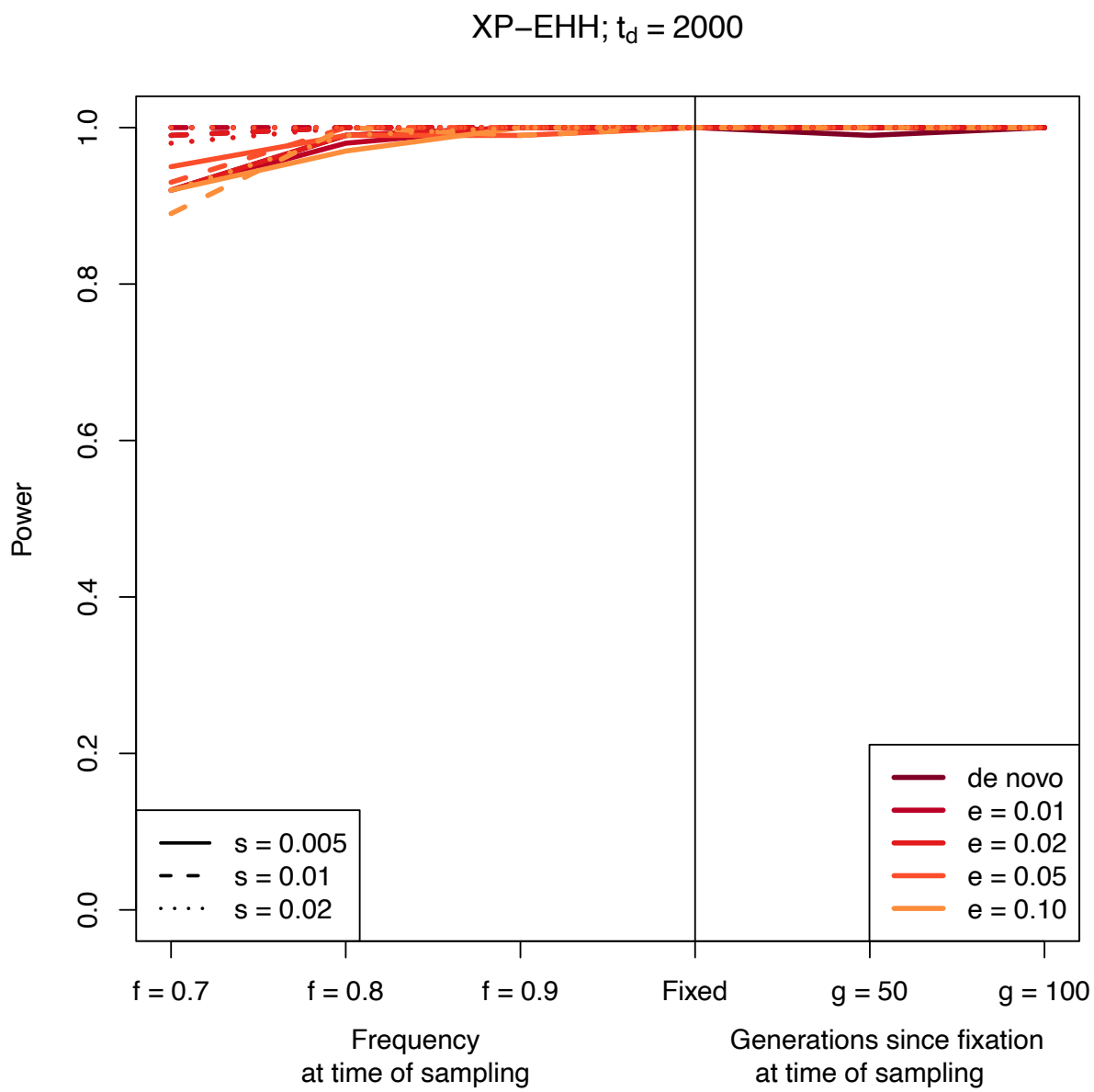

Figure S29. Demo 5 XP-EHH $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

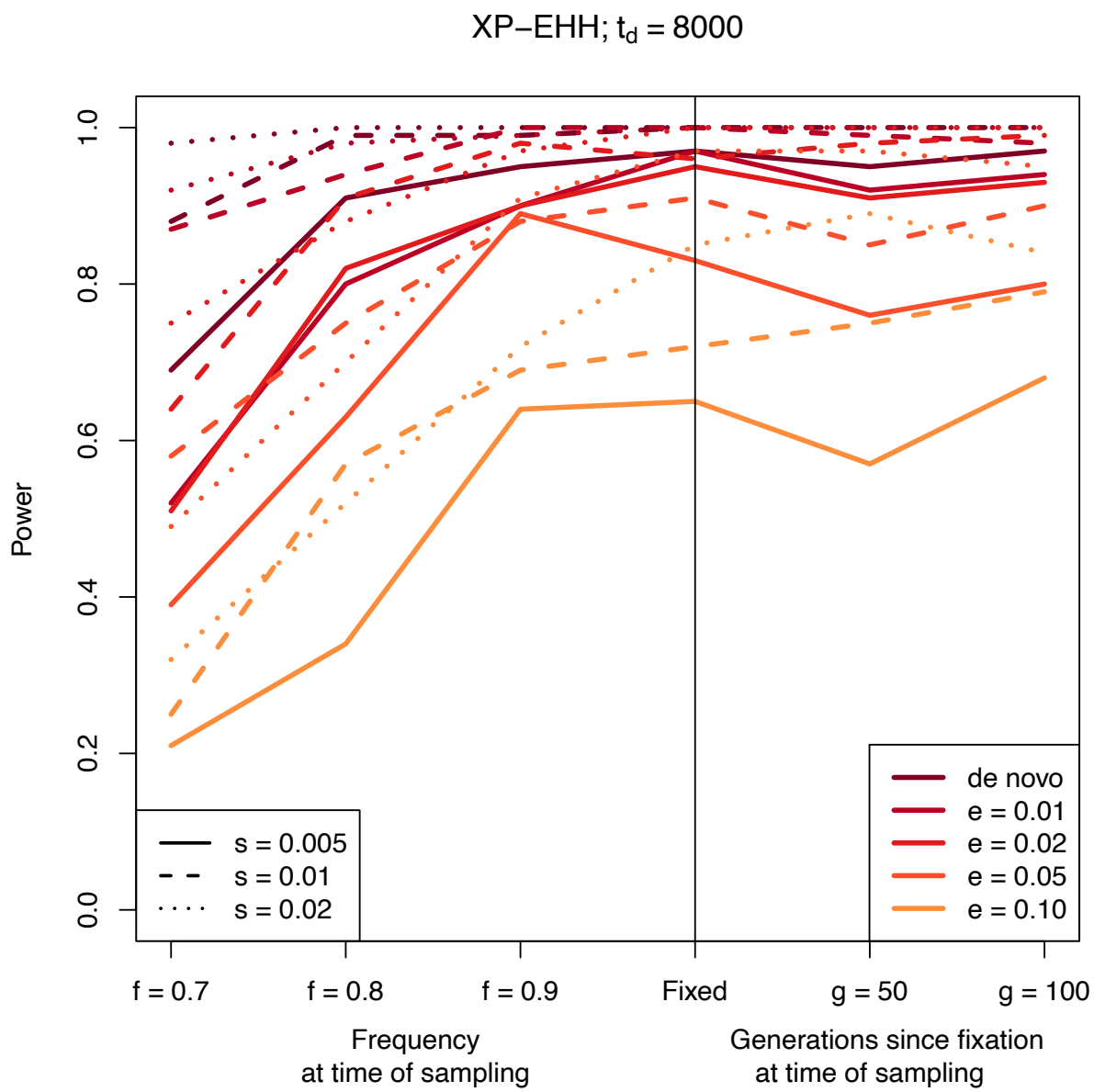

Figure S30. Demo 5 XP-EHH $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

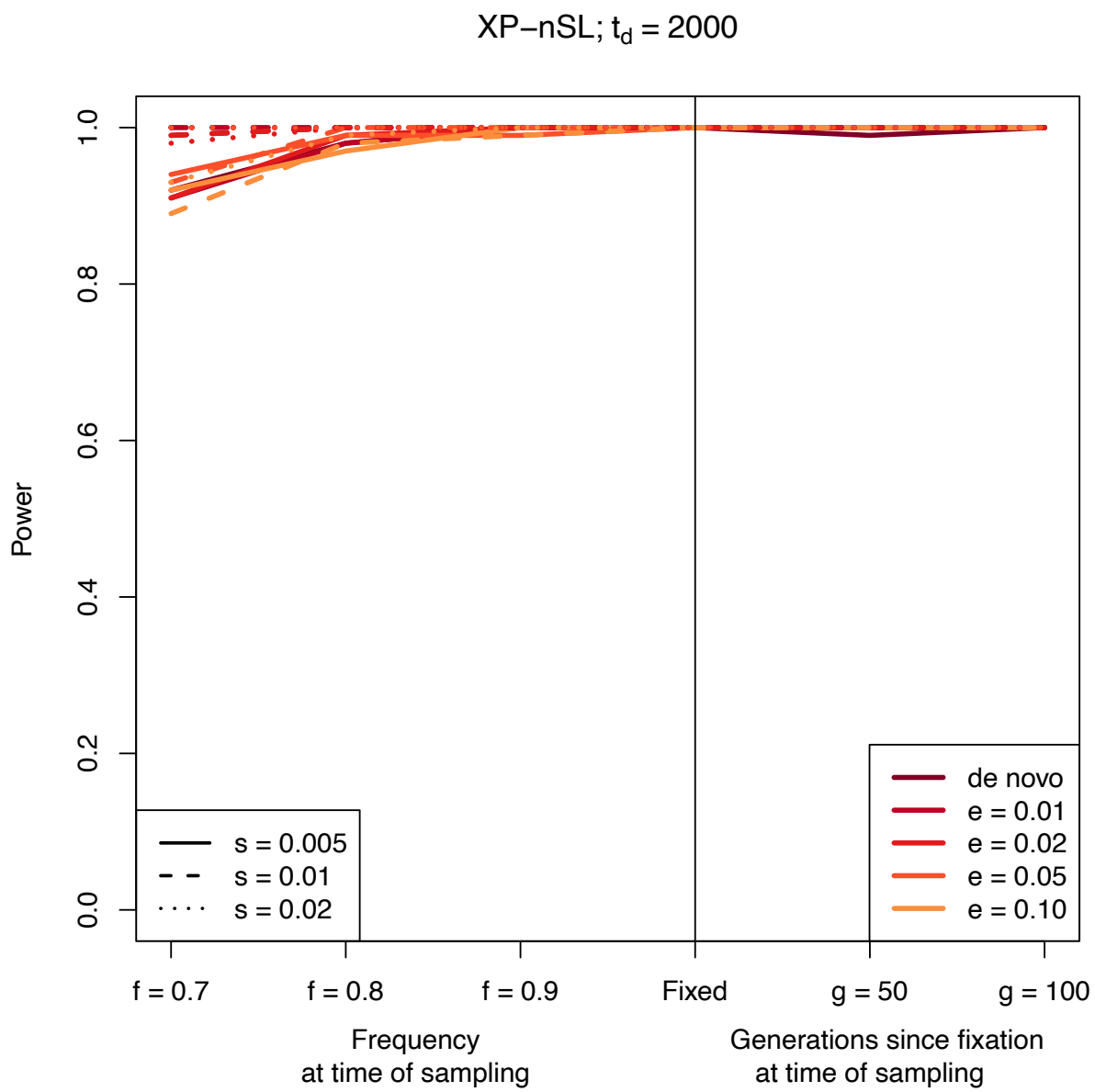

Figure S31. Demo 5 XP-nSL $t_d = 2000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

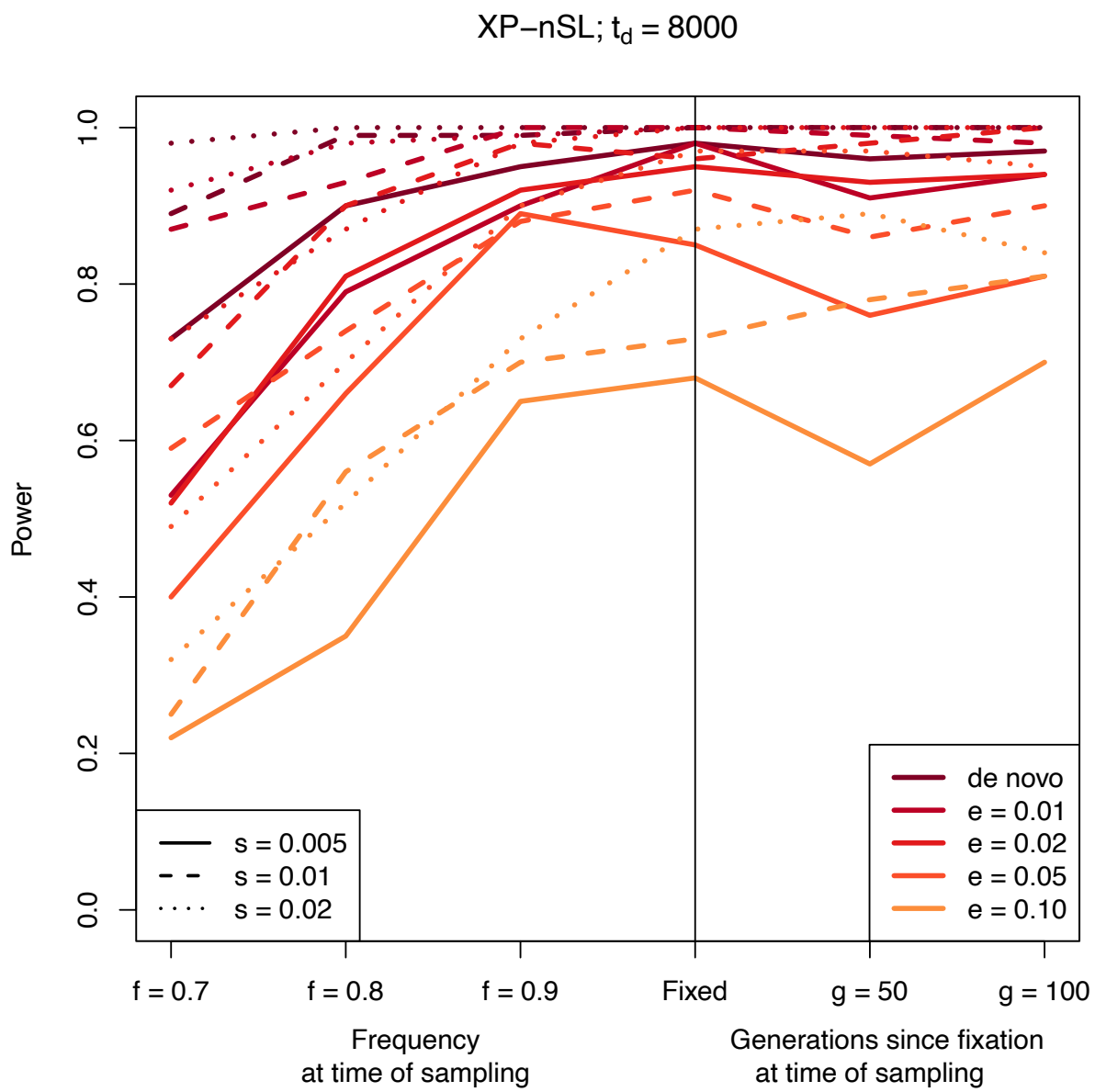

Figure S32. Demo 5 XP-nSL $t_d = 8000$ power curves. s is the selection coefficient, f is the frequency of the adaptive allele at time of sampling, g is the number of generations at time of sampling since fixation, e is the frequency at which selection began, and t_d is the time in generations since the two populations diverged.

Table 1. Demographic history parameters for simulations.

	N_A	N_0 at split	N_0 at present	N_1 at split	N_1 at present	t_d
Demo 1	10,000	10,000	10,000	10,000	10,000	2,000/4,000/8,000
Demo 2	10,000	10,000	10,000	5,000	5,000	2,000/4,000/8,000
Demo 3	10,000	5,000	5,000	10,000	10,000	2,000/4,000/8,000
Demo 4	10,000	10,000	50,000†	10,000	10,000	2,000/4,000/8,000
Demo 5	10,000	10,000	10,000	10,000	50,000†	2,000/4,000/8,000

†The reached via exponential growth starting 2,000 generations ago.

Table S1. False positive rate computed from neutral simulations for varying t_d and demographic history.

		$t_d = 2000$	$t_d = 4000$	$t_d = 8000$
iHS	Demo 1	0.013	0.1	0.009
	Demo 3	0.007	0.013	0.007
	Demo 4	0.015	0.018	0.008
nSL	Demo 1	0.01	0.015	0.008
	Demo 3	0.008	0.011	0.007
	Demo 4	0.014	0.021	0.014
XP-EHH	Demo 1	0.013	0.013	0.016
	Demo 2	0.017	0.009	0.015
	Demo 3	0.01	0.011	0.012
	Demo 4	0.012	0.014	0.014
	Demo 5	0.011	0.012	0.013
XP-nSL	Demo 1	0.014	0.011	0.013
	Demo 2	0.019	0.011	0.012
	Demo 3	0.011	0.011	0.012
	Demo 4	0.012	0.012	0.014
	Demo 5	0.011	0.012	0.014

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