Genomes from different strains of the same bacterial species often differ substantially (up to 30%) in gene content [1-6]. There are two general ways to account for such gene content variability ('patchy distribution') among closely related genomes: strain-specific loss of genes after divergence from a common species ancestor that contained the genes, and strain-specific gain of genes after divergence from an ancestor that lacked them. Gain might be effected through lateral gene transfer (LGT), duplication (paralog creation) or, much less likely, de novo creation.

Several recent publications have attempted to assess rates of within-species gain and loss using parsimony-based approaches applied to gene presence/absence data, in the context of a reference strain phylogeny [7-11]. Similar parsimony-based approaches have also been taken for inferences of gene gain/loss at larger phylogenetic distances [12-14].

In such analyses, a pattern like that shown in Figure 1a would be interpreted to indicate a single event of gain of a gene X not present in the species ancestor, after the separation of taxa 4 and 5. Explaining this distribution as the result of loss of a gene X initially present in the ancestor would, in contrast, require a minimum of four separate events, a seemingly less parsimonious scenario. However, reasoning by parsimony in such a situation requires difficult-to-test assumptions about the relative frequency of gain and loss events (that, for instance, losses are not four times more frequent than gains). Moreover, such reasoning is simply beside the point if we have some other sort of knowledge about the relevant processes that suggests we might be misled by appearances. Here we do know that gain (at least when it occurs by gene duplication or LGT) could be effectively instantaneous, but that loss will more commonly proceed gradually, through intermediates we might call pseudogenes and gene remnants (regions recognizable as gene-derived only by synteny and statistically significant sequence similarity to the parent gene). There is thus an inherent asymmetry between gain and loss both in terms of defining and of detecting them, and failure to recognize gene remnants will inevitably lead to mistaking a situation like that shown in Figure 1b (in which a gene present in the species’ ancestor has deteriorated in all lineages but one) for the situation in Figure 1a (in which a gene absent from the ancestor has been gained in a single lineage). Our goal in the present analysis was to assess how often such mistakes might be made.

Although prokaryotic genomes have traditionally been viewed as efficiently packed with functioning genes, and mutationally biased towards rapid deletion of dysfunctional regions [15], there are new indications that significant numbers of pseudogenes persist in some genomes [16-18]. In addition, detailed analyses show that in reduced genomes such as those of Rickettsia, intergenic regions often represent decaying remnants of genes [19]. Some categorization more nuanced than ‘presence’ versus ‘absence’ might thus better capture genome history. But for gain-and-loss surveys of the sort cited there may seem to be no alternative to the binary approach. A gene is considered ‘present’ if represented by an open reading frame (ORF) showing significant similarity in sequence (with arbitrarily chosen significance cutoff) and having similar length.
to a query gene; otherwise it is scored as ‘absent’. We systematically screened groups of closely related genomes (see Additional data files 1-3) for gene-family presence/absence patterns using several common criteria. When potential gene remnants detectable by less stringent methods are included, the number of gene families for which events of gain or loss within a species might be inferred (because they are scored as present only in some strains) can drop by as much as 90% (or as little as 7%) - on average about 60%. The extent to which recognition of such remnants will decrease estimates of the rates of gain of genes by LGT and increase estimates of the gene content of species’ ancestors will depend on how recognition affects inferred patterns of presence and absence as displayed on a phylogeny of the species’ strains. Each gene family must be individually examined, and where there is frequent between-strain recombination, not only is strain phylogeny a problematic concept [20], but it will sometimes be the case that gene remnants are themselves acquired by LGT.

We have assessed the impact of more complete recognition of gene remnants in the simplest cases, those species for which only three genomes are available. We calculated the number of presence/absence patterns that change under different match-length requirements for the eight such groups in our dataset (Figure 2). Any change in any of the

Figure 1
Illustration of parsimony inference from a gene/presence pattern and a reference tree topology. (a,b) Results of parsimonious inferences for the same gene family, with different criteria used to define presence/absence patterns. In (a) genes are divided into only two categories, present and absent, while in (b) the absent genes are further classified into gene remnants and genuinely absent.
The analysis of patchily distributed gene families that change their state (present or absent) in different genomes under two different selection criteria for gene families. Eight groups of three genomes each were analyzed. In one selection scheme, a match-length requirement of 85% in BLASTN was imposed (stringent selection), while in the other there was no match-length requirement in BLASTN (relaxed selection). Corresponding gene families constructed under the two criteria were compared and classified into all possible types of gene families (total $3^3 = 27$). Of these, only those types of gene families (12) where at least one gene is present under both criteria, and where at least one gene changes its state under the two criteria, are shown. They are coded as filled circles (present under both criteria), empty circles (absent under both criteria) and half-filled circles (absent under the stringent criterion and present under the relaxed criterion). Numbers in the figure indicate the number of patchily distributed gene families that change their state when under two different selection criteria. The last row is the total number of gene families for which differences in history might be incorrectly inferred, expressed as a percentage of total gene families detected as present in one or two, but not three, genomes in a genome group. The total number of gene families used in the calculation is listed in the second table in Additional data file 2. Branches on the three-taxon tree are denoted as a, b, c and d. G, gain; L, loss; A, ambiguous (both gain and loss are equally parsimonious); C, core (that is, present in all three genomes). The subscript refers to the branch on which the event is inferred. For the list of genomes in each group see Additional data file 3.
practical concern and of theoretical interest that we really do not have a definition for gene loss. It is not clear where - along the line from the appearance of the first subtly deleterious regulatory or missense mutation to the deletion of the last nucleotide - we would agree to declare a gene to be lost. Parsimony-based inferences depend on how we make that declaration, but most quantitative treatments of gene loss in evolution avoid this question altogether. Moreover, in recombinogenic species, the possibility of exchange of remnants of inactivated genes between lineages means that there will be additional difficulties in reconstructing the decay process for individual genes. Indeed, in highly recombinogenic lineages, it is often impossible to sequence remnants of inactivated genes between strains of the same species. Ideally, all gene families would be examined for even highly decayed remnants.

**Additional data files**

The following additional data are available online with this paper. Additional data file 1 contains Materials and methods for the analyses performed. Additional data file 2 describes in detail the comparison of different BLAST-based criteria for presence/absence detection. Additional data file 3 is a table listing the composition of the analyzed genome groups.

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**References**

1. Welch RA, Burland V, Plunkett G 3rd, Redford P, Roesch P, Rasko D, Buckles EL, Liu SR, Boutin A, Hackett J, et al.: Extensive mosaic structure revealed by the complete genome sequence of uropathogenic Escherichia coli. Proc Natl Acad Sci USA 2002, 99:17020-17024.
2. Rasko DA, Ravel J, Okstad OA, Helgason E, McLysaght A, Baldi PF, Gaut BS: Pseudo-genes in bacterial genomes. Nucleic Acids Res 2005, 33:3125-3132.
3. Paulsen IT, Press CM, Ravel J, Kobayashi DY, Myers GSA, Mavrodi DV, DeBoy RT, Kyrpides NC, Olsen GJ, et al.: The genome sequence of Bacillus cereus ATCC 10987 reveals metabolic adaptations and a large plasmid related to Bacillus anthracis pXO1. Nucleic Acids Res 2004, 32:977-988.
4. Altschul SF, Madden TL, Schaffer AA, Zhang J, Zhang Z, Miller W, Lipman DJ: Gapped BLAST and PSI-BLAST: a new generation of protein database search programs. Nat Biotechnol 1997, 15:215-217.
5. Perna NT, Magrini V, Cieslewicz MJ, Donati C, Medini D, Ward NL, Angiuoli SV, Crabtree J, Jones AL, Durkin AS, Crabtree J, Jones AL, Durkin AS, et al.: Genome analysis of multiple pathogenic isolates of Streptococcus agalactiae: implications for the microbial “pan-genome”. Proc Natl Acad Sci USA 2005, 102:13950-13955.
6. Hsu J, Hong X, Liu X, Bierne N, Wright F, et al.: An integrated genomic approach to identify genes associated with population structure in Drosophila melanogaster. PLoS Genet 2006, 2:e15.
7. Richter JD, Ross AG, Dehal PS: The small genome of the myxobacterium Sorangium cellulosum. Proc Natl Acad Sci USA 2007, 104:8827-8832.
8. Casjens SR, Krause K, Banfield JF: Memory in the myxobacterial genome. Mol Biol Evol 2007, 24:1702-1712.
9. Brown JR, Liu X, Bierne N, Wright F, et al.: An integrated genomic approach to identify genes associated with population structure in Drosophila melanogaster. PLoS Genet 2006, 2:e15.
10. Hsu J, Hong X, Liu X, Bierne N, Wright F, et al.: An integrated genomic approach to identify genes associated with population structure in Drosophila melanogaster. PLoS Genet 2006, 2:e15.
11. Marri PR, Bannantine JP, Paustian ML, Golding GB: Lateral gene transfer in Mycobacterium avium subspecies paratuberculosis. Can J Microbiol 2006, 52:560-569.
12. Kunin V, Ozouzovs CA: The balance of driving forces during genome evolution in prokaryotes. Genome Res 2003, 13:1589-1594.
13. Ortutay C, Gaspari Z, Toth G, Jager E, Vida M, Gaspari Z, Toth G, Jager E, Vida M: Algorithms for computing parsimonious evolutionary scenarios for genome evolution, the last universal common ancestor and dominance of horizontal gene transfer in the evolution of prokaryotes. BMC Evol Biol 2003, 3:
14. Liu Y, Harrison PM, Kunin V, Gerstein M: Comprehensive analysis of pseudogenes in prokaryotes: widespread gene decay and failure of putative horizontally transferred genes. Genome Biol 2004, 5:R64.
15. Paabo S, Andersson SG: Pseudo-genes, junk DNA, and the dynamics of Rickettsia genomes. Mol Biol Evol 2001, 18:829-839.
16. Li R, Chen W, Hsieh C, Wu D: Small change: keeping pace with microevolution. Nat Rev Microbiol 2005, 2:483-495.
17. Altschul SF, Madden TL, Schaffer AA, Zhang J, Zhang Z, Miller W, Lipman DJ: Gapped BLAST and PSI-BLAST: a new generation of protein database search programs. Nucleic Acids Res 1997, 25:3389-3402.
18. Hanage W, Fraser C, Spratt B: Fuzzy species among recombinogenic bacteria. BMC Biol 2005, 3:6.
19. Thompson JD, Higgins DG, Gibson TJ: CLUSTAL W: improving the sensitivity of progressive multiple sequence alignment through sequence weighting, position-specific gap penalties and weight matrix choice. Nucleic Acids Res 1994, 22:4673-4680.
20. Swofford D: PAUP*: 4.0 Beta Version, Phylogenetic Analysis Using Parsimony (and Other Methods) Sunderland, MA; Sinauer Associates: 1998.
21. van Dongen S A cluster algorithm for graphs. Technical Report INS-R0010.
Amsterdam: National Research Institute for Mathematics and Computer Science in the Netherlands; 2000.

26. Konstantinidis KT, Tiedje JM: Genomic insights that advance the species definition for prokaryotes. Proc Natl Acad Sci USA 2005, 102:2567-2572.

27. Pearson WR. Effective protein sequence comparison. Methods Enzymol 1996, 266: 227-258.

28. Fraser-Liggett CM: Insights on biology and evolution from microbial genome sequencing. Genome Res 2005, 15:1603-1610.

29. Nierman WC, DeShazer D, Kim HS, Tettelin H, Nelson KE, Feldblyum T, Ulrich RL, Ronning CM, Brinkac LM, Daughtery SC, et al: Structural flexibility in the Burkholderia mallei genome. Proc Natl Acad Sci USA 2004, 101:14246-14251.