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## The modal distribution of protein isoelectric points reflects amino acid properties rather than sequence evolution

Georg F. Weiller<sup>1, 2</sup>, Gilles Caraux<sup>3</sup> and Nicole Sylvester<sup>2</sup>

Two-dimensional gel electrophoresis, a routine application in proteomics, separates proteins according to their molecular mass ( $M_r$ ) and isoelectric point (p/). As the genomic sequences for more and more organisms are determined, the  $M_r$  and p/ of all their proteins can be estimated computationally. The examination of several of these theoretical proteome plots has revealed a multimodal p/ distribution, however, no conclusive explanation for this unusual distribution has so far been presented. We examined the p/ distribution of 115 fully sequenced genomes and observed that the modal distribution does not reflect phylogeny or sequence evolution, but rather the chemical properties of amino acids. We provide a statistical explanation of why the observed distributions of p/ values are multimodal.

Keywords: Extremophiles / Genome / Isoelectric point / Multimodal distribution

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#### 1 Introduction

Proteins exhibit very sharp pls and can therefore be very well separated by IEF [1]. Images of 2-DE gels depicting the proteomes of various organisms have been widely published, and interactive websites have been established where the individual protein spots of these images provide a link to further information on the respective protein. For organisms for which the genomic sequence is determined it is also possible to create an image of their theoretical proteome by plotting the theoretical pl of all predicted proteins against their theoretical  $M_r$ . Bjellqvist et al. [2, 3] have shown that the pl of a denatured linear protein, as observed in 2-DE, can be calculated with high accuracy. While most 2-DE gels utilise narrow pl gradients only revealing expressed proteins within a limited pl range (e.g., pH 3-7, or pH 7-9), theoretical proteomes typically display the distribution of  $M_r$ s and p/s of the entire proteome over the entire pl range (pH 0-14).

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Our bioinformatics laboratory forms part of an ongoing Sinorhizobium proteome project for which we have produced several theoretical proteome plots. Curiously, a roughly bimodal distribution of the pls of the proteins was observed in all plots, with the scarcity of proteins within the pH 7.4-7.6 range being the most significant feature. Other investigators have made similar observations. Already in the early 1980s Gianazza and Righetti [4] showed that protein  $M_r$ s and p/s are not normally distributed. Urguart et al. [5] observed a bimodal pl distribution in Mycobacterium bovis. VanBoegelen et al. [6] analysed several bacteria and also observed a bimodal distribution of pls with peaks centred around pH 5.5 and pH 9. The authors assumed that this distribution reflected the intracellular pH of the cell. As proteins are generally poorly soluble near their pl [7] it would appear that most proteins have evolved to avoid a pl close to the pH of the cytoplasm, which is assumed to be near neutrality. Schwartz et al. [8] extended this analysis to several eukaryotes for which they reported a trimodal distribution. They confirmed the previously observed bimodal distribution for bacteria and archea and presented an additional peak near pH 7 as being characteristic for eukaryotes representing nuclear proteins. The authors also speculate that the pl of eukaryotic proteins reflects the subcellular localization, and suggest that theoretical pl calculations may

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provide a way of tentatively assigning subcellular localizations to proteins that are identified in sequenced genomes but not further characterised. Independently of the studies above, we have analysed the pl distributions of 115 completely sequenced genomes or chromosomes spanning all three kingdoms of life as well as mitochondria, chloroplasts and viruses. Our results differ in several aspects from previous reports and we arrive at a different conclusion. Here, we briefly describe the breadth of observed variations and offer an explanation for the distribution of protein pl values found in different organisms.

### 2 Methods

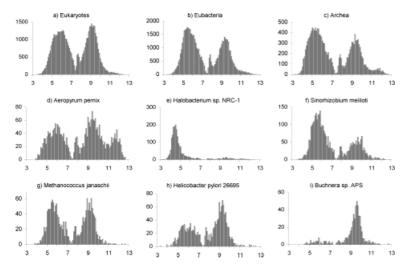
The predicted protein sequences of 115 complete genomes (available in January 2001) were obtained from the National Centre for Biotechnology Information [9] and the Protein Information Resource [10]. The data span 115 different genomes and included 8 Archea, 32 Eubacteria, 5 Eucaryotes (Arabidopsis thaliana, Caenorhabditis elegans, Drosophila melanogaster, Saccharomyces cerivisiae, Plasmodium falciparum), 48 mitochondria, 9 chloroplast, 9 viruses of prokaryotes and 4 viruses of eukaryotes. Programs for predicting the  $M_r$  and pl from predicted proteins and for binning the data given in the histograms were written in C. The pl values were calculated using a stepwise approach with the pH value increasing by 0.1 until the positive and negative charges in the protein were equal. Note that the calculated pl depends considerably on the set pK values assumed for the ionisable groups. When several different sets of published pK values were used the predicted pl of some proteins differed by up to 1 pH unit. We have observed that the shapes of the pl distributions vary considerably if other pK values are used, but the multimodal shape persists. All the data presented here use the pK values of amino acids described by Bjellqvist *et al.* [2], which were defined by examining polypeptide migration between pH 4.5 to 7.3 in an IPG gel environment with 9.2 M and 9.8 M urea at 15°C or 25°C, and are therefore close to the values obtained in 2-DE.

The programs for generating random sequences according to a specified multinomial model were written in C using *rand1* and *gasdev* as random number generators [11]. The first one was used to generate random sequences according to a specified multinomial distribution model. The second one is a normally distributed random deviation generator. We used it to approximate multinomial distribution by a normal distribution. The *pl* distribution histograms were plotted with MS Excel (Microsoft Corporation) and Sigmaplot (Statistical solutions, Saugus, MA, USA).

#### 3 Results and discussion

## 3.1 Distribution of estimated protein p/ in genomic sequences

Figure 1 shows the theoretical p/ distributions found in 135 621 proteins derived from 115 fully sequenced genomes or chromosomes. The first three histograms summarise the proteins of eukaryota, eubacteria and archea, respectively. Most proteins form part of one of two major clusters with peaks at pH 5.5 and 9.5, respectively, representing the previously observed bimodal distribution of p/ values. Very few proteins have a p/ between 7 and 8 except around pH 7.8 where a small and narrow third peak is found. On the acidic side of the spectrum the distribution raises sharply. No protein in the analysed dataset has a p/ below 3.0 and only 394 (less than 0.3%) pro-



**Figure 1.** Histograms of computationally estimated p/ distributions in steps of 0.1 p/. The p/ values ranging from 3–13 are given on the abscissa and the number of proteins on the ordinata.

teins have a pl below 4. In contrast, the alkaline side has a long tail. Here, 9706 proteins (more than 7%) have a pl of 10 or above. This is due to a small group of proteins which have a alkaline pl that forms a fourth peak around pH 12. This fourth peak is more visible in archea and is particularly pronounced in the archaeon Aeropyrum pernix (Fig 1d).

Besides this, the distribution in the different kingdoms is remarkably similar. The only apparent differences being that in eukaryota the main alkaline peak is slightly higher that the acidic peak, while the reverse is true for both prokaryotic kingdoms. These differences are however, not specific for the kingdoms. For instance, the distributions of *S. cerivisiae* and *D. melanogaster* are very similar to the eubacterial histogram (panel b) with the main acidic peak being the most prominent (data not shown). The last 6 panels (d–i) of Fig. 1 show some individual prokaryotic species. Note that the pH of the four peaks is the same for all organisms, however, the relative abundance of proteins belonging to the four peaks differs largely in individual species. These differences do not reflect phylogeny.

In most organisms neither of the two major peaks contain more than about 75% of the proteins. These distributions are exemplified in panels f-h. In a small number of organisms the discrepancy is however, more pronounced, with Halobacterium (panel e) and Buchnera (panel i) forming the extremes in the data that we examined. At least for these species, and possibly also for Aeropyrum (panel d), the relative amount of proteins in the individual peaks appears to reflect the environment for which the species are adapted as their extreme pl distribution correspond to extreme environments and all of these organisms can be classified as extremophiles. The adaptation of the extreme halophilic archaeon Halobacterium sp. NRC-1 to its environment has been analysed by Kennedy et al. [12]. On the other extreme of the distribution spectrum is the eubacterium Buchnera sp. Although a close relative to Escherichia coli [13], which has a pl distribution similar to the one given in Fig. 1b, Buchnera is an obligate endocellular symbiont that can only survive in the bacteriocytes of aphids. The extremely high average pI (pH = 9.6) of this species has previously been observed by Shigenobu et al. [14] and has been attributed to the high lysine usage of the species. Aeropyrum is an extreme thermophile archaeon growing at temperatures of up to 100°C [15]. It is however, not clear whether the fourth and alkaline peak (pH 12) observed in this species is associated with heat resistance, as such a peak is absent from other thermophiles like the archaeon Methanococcus janaschii (Fig. 1g).

While the relative abundance of proteins forming the individual peaks differs in different organisms, the pH of the peaks and especially of the troughs remains constant,

e.g., there is always a trough at pH 7.4 and at pH 8.0. We were interested in whether this characteristic shape of pl distribution reflects genetic evolution. Have all proteins evolved to be charged at physiological pH as has been suggested? There are two impediments to answering this question; first we do not know the pl of native proteins, as theoretical calculations can only estimate the pl of denatured polypeptides. Further, the actual physiological pH cannot be measured in vivo, and exact estimates therefore, do not exist. However, it may be reasonable to assume that the physiological pH is often close to neutrality and that the distribution of native protein pls follows the distribution of denatured polypeptides. Have sequences evolved to avoid pl values coinciding with the observed troughs?

To answer these questions we have analysed the pl distribution of randomly generated protein sequences and found that the distributions closely follow the distributions of actual sequences. Clearly, nucleotide composition as well as amino acid composition varies widely between organisms, and this will result in differences in the average pl of the proteins. In simulations that model the spread and variation of sequence length as well as amino acid compositions of real organisms, random sequences have produced the same pl distribution as actual sequences. We conclude therefore that sequences have not evolved to conform to a specific pl, but rather the observed pl distribution is a statistical consequence that follows from the chemical properties of the amino acids that make up the proteins. Below we give a statistical explanation as to why the modal distribution of pl values is to be expected without assuming any selective pressures.

## 3.2 Distribution of estimated protein p/ in the absence of evolutionary constraints

By definition, the pI of a protein is the pH at which the positive charge is equal to the negative charge. At this pH, the sum of charges, over all amino acids, is null. In proteins there are four ionisable groups that can assume positive charges. These are the three amino acids lysine (K), argine (R) and histidine (H) as well the N-terminus (N-term). Negative charges can be assumed by the four amino acids tyrosine (Y), cysteine (C), aspartate (D) and glutamate (E) and the C-terimus (N-term) of the protein. Let N-term of all positive charges and N-term be the sum of all negative charges of a given protein. Both values depend on the pH and the pK values of the ionisable groups in the protein. We define N-term as the set of amino acids charged positively and N-term as the set of amino acids charged negatively.

Hence, the positive charge  $f^+$  of a given protein and its negative charge  $f^-$  can be defined by:

$$f^+ = \sum_{i \in I^+} n_i f_i^+$$
 and  $f^- = \sum_{i \in I^-} n_i f_i^-$  Eq. 1

where  $f_i^+$  is the elementary charge assumed by an amino acid of type i, and  $n_i$  is the number of amino acids of type i in the current protein. We have  $n_{N-\text{term}} = n_{C-\text{term}} = 1$ . By definition, the pl value of a protein is the solution of the Eq.

$$f^{-}(pH) + f^{-}(pH) = 0$$
 Eq. 2

For a given amino acid,  $f_i^+$  or  $f_i^-$  is estimated from Henderson-Hasselbach's Eq. by:

$$f_i^+(pH) = \frac{1}{1 + 10^{pH - pK_i}}$$
  
 $f_i^-(pH) = f_i^+(pH) - 1$ 

$$= \frac{10^{pH-pK_i}}{1+10^{pH-pH_i}}$$

where  $0 \le f_i^+$  (pH)  $\le 1$  and  $0 \le f_i^-$  (pH)  $\le 1$ .

All  $f_i^+$  functions have the same shape and can be deduced from another one by translation. The same is true for the  $f_i^-$  functions. They are a sigmoid function of pH with an inflection point at pH = pK<sub>i</sub>. This point is a symmetric point where

$$f_i^+$$
 (pk<sub>i</sub>) = 1/2 or  $f_i^-$  (pK<sub>i</sub>) =1/2.

In each function  $f_i^+$  decreases with increasing pH, ranging from 1 to 0. Reciprocally,  $f_i^-$  is an increasing function of pH, ranging from 0 to 1. The  $f_i^+$  curve has a relatively steep zone extending about 1.0 pH unit on either side of the pK<sub>i</sub>. This zone is the well-known buffering region as large changes in the charge result in only slight variations of pH. The entire graph resembles a step centred at the pK<sub>i</sub>.

From these properties, and from Eq.(1), the positive charge  $f^+$  is a descending stair function (Fig. 2) varying from

$$n_{\mathsf{l}^+} = \sum_{\mathsf{i} \in \mathsf{l}^+} n_\mathsf{i}$$

to 0

Each step is associated with an amino acid and is centred on its pK value. The height of each step is  $n_i$  and depends only on the frequency of the associated amino acid in the protein. Note that the position (pH) of each step is independent of the protein!. Only its height (charge) changes depending on the amino acid composition.

For amino acids with close pK values the steps are smoothed forming a single step as the buffering region of both amino acids intersects (e.g., K and R in Fig. 2).

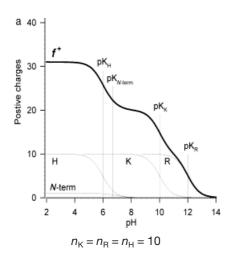
Symmetrically, and for the same reasons,  $f^-$  is a rising stair function varying from 0 to

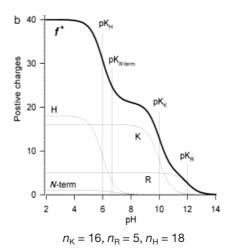
$$n_{\mathsf{l}^-} = \sum_{\mathsf{i} \in \mathsf{l}^-} n_{\mathsf{i}}.$$

The pl value is the abscissa of the intersection point between the ascending and descending stair graphs (Fig. 3).

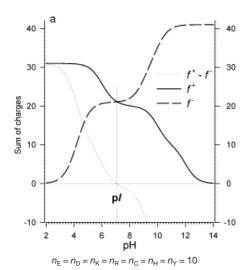
## 3.3 The influence of the shape of $f^+$ and $f^-$ on p/position

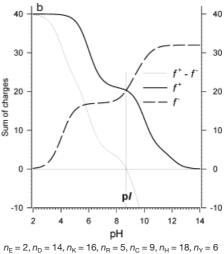
Equation 2 does not have a simple solution and it would be very difficult to study the relationship between pl and the amino acid composition analytically. Consequently, we followed the influence of each amino acid graphically and used simulation to exhibit the distribution of pl around a set of random sequences. First, it is easy to establish from Fig. 3 that each amino acid charge does not have the same effect on pl values. If the number of histidines  $(n_H)$  increases in a sequence, the height of the associated





**Figure 2.** Positive charge curve  $(f^{+})$  as a function of pH. The additive contribution of each amino acid is given in thin lines and their numbers  $(n_{\rm K}, n_{\rm R}, n_{\rm H})$  are indicated for part a) and b) of the Figure.





**Figure 3.** Positive  $(f^+)$  and negative  $(f^-)$  charge curves. The pI value is the abscissa of the intersection point between  $f^+$  and  $f^-$ , e.g. where the sum of charges $(f^+-f^-)$  is 0. The numbers of each amino acid  $(n_{\rm E}, n_{\rm D}, n_{\rm K}, n_{\rm R}, n_{\rm C}, n_{\rm H}, n_{\rm N}, n_{\rm H})$  are indicated for part a) and b) of the Figure.

(left most) step in  $f^+$  increases accordingly. This does not significantly modify the curve to the right hand side of this step. In Fig. 3 for instance, the number of histidines  $(n_{\rm H})$  changes from 10 (a) to 18 (b). Note that this modification does not alter the pl value by much. In general, an amino acid with a positive charge and a low (acidic) pK does not normally have a large effect on the pl. For the same reasons, an amino acid with a negative charge and a high (alkaline) pK does not have a large effect on the pl.

In contrast, a change in the occurrence of an amino acid with positive charge and high pK, or an amino acid with negative charge and a low pK, have a strong effect on the pl of a protein. In Fig. 3, the value of  $n_K$  increases from 10 to 16, but at the same time,  $n_R$  decreases from 10 to 5. Both amino acids are associated with the same step of the  $f^+$  curve. The total height of the step associated with these amino acids grows slightly and the pl moves slightly to the right. The stair shapes of the  $f^+$  and f functions explain the uneven effect that a change in the number of charged ionic groups has on the resulting pl of the protein. If the p*I* appears in a zone where the  $f^+$  curve is relatively flat, a small change in  $f^-$  has a larger effect on the pl than if  $f^+$  is in a buffer zone (near a pK value). The reverse is true for  $f^-$ . For example, in Fig. 3a the p*l* is 7.1 and the  $f^-$  curve is relatively flat in this zone:

$$\frac{\partial f^{-}(\mathsf{pI})}{\partial \mathsf{pH}} = 0.92.$$

If  $f^+$  increased by one charge unit, the p/ would increase to 7.8 ( $\Delta_{\rm pl}=+0.7$ ).

By contrast, in Fig. 3b the p*l* is 8.7 and  $f^-$  curve is growing in this zone:

$$\frac{\partial f^{-}(\mathsf{pI})}{\partial \mathsf{pH}} = 5.82.$$

If  $f^+$  would increase by one charge unit, the p*I* would only raise to 8.8 ( $\Delta_{pl} = +0.1$ ).

Thus, the density distribution of pI values must be low on the areas where  $f^+$  or  $f^-$  are flat. For example, pI distribution should be particularly small between 7 and 8, where  $f^+$  and  $f^-$  are both relatively flat.

### 3.4 Simulated p/ distributions

To test this prediction let us consider that the amino acid composition of a sequence is the result of a random process. Each  $n_i$  is now a random variable with a specific distribution, and the p/ therefore represents a random value. We are interested by its distribution but as we do not know the exact distribution of  $n_i$  for a set of similar proteins we cannot know the distribution of the p/ without further assumptions. For a fixed sequence length

$$I = n_{l^+} + n_{l^-} = \sum_{i \in l^+ \cup l^-} n_i,$$

we suppose that the  $n_i$  are coming from a multinomial distribution with l trials and  $p_i$  probabilities. Thus, for a fixed amino acid, the mean and SD of the number of its occurrences are:

$$v_i = E(n_i) = I p_i$$
 and  $\sigma_i = \sqrt{V(n_i)} = \sqrt{I p_i (1 - p_i)}$ .

We have to choose values for parameters I and  $p_i$ . To get realistic values for I we modelled a set of sequence lengths on an existing proteome (we chose Sinorhizobium meliloti), and randomly chose I from this set in our simulations. For  $p_i$  we explored two cases. For the first one, we studied the theoretical case where  $v_i = Cst$  (Fig. 3a), *i.e.* 

where the probability of occurrence is the same for each ionisable amino acid. For the second case, we chose an heterogenous set of values for  $p_i$  given in Fig. 3b.

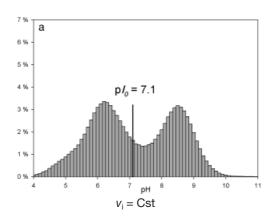
Furthermore, from Eq. (1) we can see that  $f^+$  and  $f^-$  are linear expressions of the  $f_i^+$  and  $f_i^-$  functions. In these expressions each  $f_i^+$  and  $f_i^-$  are weighted by  $n_i$ . As these weights are integers the pl distribution reflects a discontinuous phenomenon when amino acid distributions change. This discontinuity complicates our study by introducing local modifications in pl distribution. Then, in a first step, we chose to eliminate discontinuity effects by considering that  $n_i$  has a continuous value. We simulate  $n_i$  by the normal approximation of a multinomial distribution. We suppose that

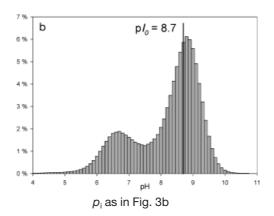
$$n_i \equiv N(v_i = I p_i, \sigma = \sqrt{I p_i (1 - p_i)}).$$

We obtain an excellent approximation of multinomial distribution if we round  $n_i$  to the nearest integer. Without rounding  $n_i$ , we find a smooth distribution for pl with two peaks

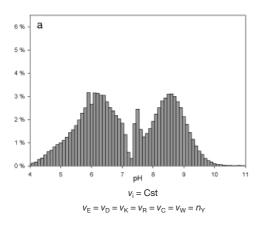
(Fig. 4) for each case studied. We also confirm our initial prediction by observing that the p/ distribution is small for a pH between 7 and 8 (relatively flat zone for  $f^+$  and  $f^-$ ).

If we round  $n_i$ , obtained with normal approximation, or if we use a multinomial generator, we reintroduce  $n_i$  discontinuity. Then  $f^+$  and  $f^-$  fluctuate across a finite set and the pl cannot take all pH values. Some of them are impossible, some are more frequent values. A region where we can find the effects of these discontinuity points is for a pH between 7 and 8. In this zone the pl is more sensitive. As we discuss below, any modification of  $f^+$  or  $f^-$  has an amplified effect on the pl. This is the reason why, in Fig. 5, a multimodal distribution arises. Between pH 7 and 8 we observe a zone with a low pl value distribution as well as a zone with a local peak. This peak corresponds to the third peak found in the pl distribution of most organisms and is most apparent in large protein sets. This peak does not have any biological significance but arises from the discrete definition of  $f^+$  and  $f^-$  across  $n_i$ .





**Figure 4.** Simulated pI distribution using a continuous normal distribution perturbation. A line indicates  $pI_0$ , the pI of a sequence with  $n_i = v_i$ , see section 3.4.



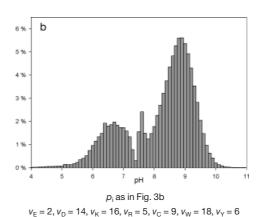


Figure 5. Distribution of pl with a multinomial distribution, see section 3.4.

## 4 Concluding remarks

Positive and negative charges of a protein are respectively, descending and ascending functions of the pH. The underlying sigmoid shape of the pK curves of the individual chargeable groups causes both functions to assume an uneven stair-like shape. The intersection of the curves represents the protein's pl and is more likely to occur at a pH where the curves are steep. Consequently, some pl values are attained by a large number of different compositions of charged amino acids while combinations of amino acids resulting in other pl values can hardly be obtained. An uneven distribution of pl values must therefore be expected when a sufficiently large number of proteins are examined. Our simulations show that the distribution of randomly generated pl distributions concur with the multimodal distributions of pl values observed in all organisms. Most proteins form part of one of the two major clusters with peaks at pH 5.5 and 9.5 respectively, although minor peaks can be detected at pH 7.8 and pH 12. Note that the exact positions of these peaks depend on the pK values assumed for the pl calculations. This characteristic location of the distribution peaks is observed in the organisms of all kingdoms and is also found in chloroplasts, mitochondria and viruses. However, depending on the amino acid usage, the number of proteins that form part of the different clusters differ in different organisms. In extreme cases the acidic or alkaline cluster may be missing almost entirely. These unusual distributions have only been found in extremophiles, and almost certainly reflect the extreme environment in which these organisms live.

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