

IMPROVED METHOD FOR CONVERSION OF PHONETIC CHINESE TO CHARACTER CHINESE

Field of the Invention

This invention relates to automated methods for converting phonetic

5 Chinese to character Chinese.

Background of the Invention

Because the Chinese language uses thousands of characters--in contrast to the English language's use of 26 characters--the development of modern Chinese word processing equipment is a substantial problem. Obviously, a typewriter

10 keyboard consisting of thousands of keys is impractical.

Phonetic (sometimes called phonemic) input schemes, based on the use of normal keyboards, are often used to input Chinese character text into a computer or word processor. These schemes known as the Mandarin Phonetic system in Taiwan and the Hanyu Pinyin system in the People's Republic of China involve the

15 transliteration, for example, of the five following Chinese characters:

台 灣 有 颱 風

respectively into the five following single syllables reflecting the pronunciation of the Chinese characters:

tai2 wan1 you3 tai2 feng1

20 Because this phonemic input does not require special keyboards or the mastery of special coding schemes, its use is advantageous. However, since the number of pronounced Chinese syllables is significantly fewer than the number of Chinese characters, it suffers from the problem of ambiguity. A well-educated Chinese may be expected to know about 6000 characters, but the number of syllables

25 is about 1200. Thus, one syllable may represent many different characters. For example, in the Hanyu Pinyin transliteration system, all the following characters are pronounced shi4:

是	事	市	式	世	示	士	視	識	試	適	室
勢	釋	氏	飾	侍	逝	誓	仕	嗜	恃	拭	噬
弑	弑	筴	柿	爽	舐	荏	伫	促	俾		

Not surprisingly, translation of the 5-syllable phrase

tai2 wan1 you3 tai2 feng1

into Chinese characters presents over 21,000 (i.e., $9 \times 5 \times 6 \times 9 \times 9$) different combinations of characters since "tai2", "wan1", "you3", "tai2", and "feng1"

5 respectively represent at least 9, 5, 6, 9, and 9 different Chinese characters.

As described in their paper ("Removing the Ambiguity of Phonetic Chinese by the Relaxation Technique," Computer Processing of Chinese & Oriental Languages Vol. 3, No. 1, May 1987) Lin and Tsai, attempting to overcome the above-described ambiguity problem, propose a method of converting phonetic

10 Chinese syllables to character Chinese using the relaxation process widely used in image analysis problems, such as edge detection, curve detection and shape recognition. More specifically, they employ the relaxation process to obtain the optimal path through the lattice of possible characters, making use of the lexical probabilities of the characters given the syllables and the transition probabilities of
15 adjacent characters and adjacent syllables.

Summary of the Invention

I have discovered a simpler, yet equally effective, method for converting phonetic Chinese into Chinese characters. In accordance with the principles of my invention, this conversion is effected by obtaining the optimal path through the
20 lattice of possible Chinese characters by calculating only the probability of adjacent Chinese characters appearing in text.

More specifically, an automated method is disclosed for converting n phonetic Chinese syllables S_1 through S_n in a text into n Chinese characters C_1 through C_n . In accordance with my method, for each Chinese syllable S_i , I generate
25 a group of Chinese characters C_{i1} through C_{iz} , collectively referenced as C_{i1-z} , possibly corresponding thereto. Then I compute the optimal path through the lattice of possible Chinese characters C_{i1-z} through C_{n1-z} to derive the mostly likely n Chinese characters C_1 through C_n corresponding to the n phonetic Chinese syllables S_1 through S_n . This optimal path is beneficially computed, in accordance with the
30 principles of my invention, by deriving the probability of the use of adjacent Chinese characters C_i and C_{i-1} in said text based upon the frequency of the ordered appearance of said adjacent Chinese characters in a large corpus of Chinese text, multiplying together the derived probabilities, and selecting the path with the highest probability as the optimal path.

Brief Description of the Drawing

Further features and advantages of my invention will become apparent from the following detailed description, taken together with the drawing, in which:

FIG. 1 is a flow diagram of the Chinese syllable to Chinese character conversion method according to my invention;

FIG. 2 is a Chinese character matrix indicating the possible Chinese characters ($C_1 - C_9$) corresponding to each of the particular Chinese syllables ($S_1 - S_5$);

FIG. 3 discloses the same Chinese character matrix shown in FIG. 2, but, where for ease of explanation, the Chinese characters have been replaced by English letters "a" through "h";

FIG. 4 is a matrix reflecting the number of times representative Chinese characters ("a" through "h") appeared adjacent one other in a large corpus of Chinese character text; and

FIG. 5 shows how the frequency numbers of FIG. 4 are used to compute the optimum path through the matrix of FIG. 3 to thereby select the Chinese characters most likely corresponding to the particular Chinese syllables.

Detailed Description

I will now describe my invention in terms of selecting the optimal path through the lattice of Chinese characters generated from a phonetic input. So consider the example below in which the phonetic Chinese syllables ("tai2", "wan1", "you3", "tai2", "feng1") each refer respectively to one of the Chinese characters found beneath it. Thus, "wan1" refers to one of the five Chinese characters beneath it, and "you3" refers to one of the six Chinese characters beneath it.

tai2	wan1	you3	tai2	feng1
台	灣	有	台	風
颱	彎	友	颱	豐
抬	琬	莠	抬	峰
檯	婉	酉	檯	封
臺	剋	勣	臺	鋒
檯		牖	檯	蜂
苔			苔	瘋
抬			抬	楓
郃			郃	丰

The circled Chinese characters represent the most likely path through the lattice, that is, the sentence in Chinese: "Taiwan has typhoons."

I will now describe the methodology used to select the optimum path in the above-described example in accordance with the principles of my invention.

- 5 Step 1 of FIG. 1 identifies the starting point of my method as the Chinese syllables S_1 through S_n or, in the illustrative example, S_1 through S_5 corresponding to the Chinese syllables "tai2", "wan1", "you3", "tai2" and "feng1", respectively. In accordance with step 2 of FIG. 2, Chinese characters C_1 through C_z , representing all possible Chinese characters corresponding to a particular Chinese syllable, are
- 10 generated for each Chinese syllable S_i , where the numbers of such characters "z" will vary for each Chinese syllable. Thus, with reference to FIG. 2 which shows the matrix formed by the generation of such Chinese characters, "z" is 9 for "tai2" since "tai2" represents one of nine possible characters, "z" is 5 for "wan1" since "wan1" represents one of five possible characters, "z" is 6 for "you3" as shown, and "z" is 9
- 15 for "feng1" since, as shown, "feng1" represents one of nine possible characters.

- In accordance with step 3 of FIG. 1, the generated Chinese characters are formed into a matrix where the characters generated for the first syllable form the left hand column C_1 and the characters generated for the last syllable (number n) form the right hand column C_n . Thus, with reference to the matrix in FIG. 2,
- 20 Chinese character C_1 for "tai2" (which is really character C_{11} in the matrix following the numbering methodology of step 3 of FIG. 1) is the first Chinese symbol under "tai2", Chinese character C_2 for "tai2" (which is character C_{12} in the matrix) is the second Chinese symbol under "tai2", Chinese character C_3 for "tai2" (which is character C_{13} in the matrix) is the third Chinese symbol under "tai2", and
- 25 so on until Chinese character C_9 for "tai2" (which is character C_{19} in the matrix) is the ninth Chinese symbol under "tai2".

- Chinese character C_1 for "wan1" is the first Chinese symbol under "wan1" and would be designated C_{21} in accordance with the numbering scheme of FIG. 1 since it is the first symbol for the second character. Similarly, Chinese
- 30 character C_2 for "wan1" is the second Chinese symbol under "wan1" and would be designated C_{22} in accordance with the numbering scheme of FIG. 1 since it is the second symbol for the second character. Finally, the last Chinese character C_5 for "wan1" is the fifth Chinese symbol under "wan1" and would be designated C_{25} in accordance with the numbering scheme of FIG. 1 since it is the fifth symbol for the
- 35 second character.

Chinese character C_1 for the last symbol "feng1" (i.e., S_5) is the first Chinese symbol under "feng1" and would be designated C_{51} in accordance with the numbering scheme of FIG. 1 since it is the first symbol for the fifth character. The last symbol in the matrix found at its bottom right corner is C_9 for "feng1" since it is the ninth symbol under "feng1" and is designated C_{59} since it is the ninth symbol for the fifth character.

FIG. 3 depicts the same Chinese character matrix as that shown in FIG. 2, but, where for ease of explanation later of the selection of the optimum path through this matrix, some of the Chinese symbols have been replaced by the English designations "a" through "h". Thus, "a" in FIG. 3 represents the first symbol under "tai2" in FIG. 2, "b" in FIG. 3 represents the second symbol under "tai2", "c" in FIG. 3 represents the first symbol under "wan1" in FIG. 2, and so on. These designations are arbitrary (except identical Chinese characters obviously are identified by the same English designation) and serve only to facilitate further explanation of the principles of my invention.

FIG. 4 shows a frequency matrix reflecting the use of adjacent Chinese characters (C_{i-1} , C_i) in a large corpus of Chinese text. This matrix was derived by analysis of the corpus and noting the number of times ordered pairs of Chinese characters appeared in the text. Again, for ease of explanation, the same English letters "a" through "h", used in FIG. 3, are again used respectively to identify the adjacent Chinese characters shown in FIG. 4. The letter "i" refers to the Chinese sentence delimiter (Chinese 'period'), and as hereinafter explained, is used to delineate the beginning and the end of a phrase of Chinese text. Since the number of Chinese characters is quite large, FIG. 4 shows only a representative portion of the actual frequency matrix which is approximately a 6000 by 6000 matrix and represents use of 6000 Chinese characters.

FIG. 4 shows the number of times ordered pairs of Chinese characters (C_{i-1} , C_i) appeared in the corpus. For example, the pair "aa" appeared 5 times in the Chinese corpus since the number 5 appears for $C_{i-1} = a$, and $C_i = a$. Similarly, the pair "ab" appeared 0 times since the number 0 appears for $C_{i-1} = a$, and $C_i = b$. On the other hand, the pair "ac" appeared 1513 times (see, $C_{i-1} = a$, and $C_i = c$). One notices immediately, that while the pair "ac" is found quite frequently in Chinese text, its inverse -- namely "ca" -- is found only infrequently (e.g., "1" in FIG. 4). One notices readily that many pairs were not found at all in the corpus -- see, for example, "ab", "af", "ah", "ba", "bb", "bc" and so on.

The letter "i" in FIG. 4 refers to the Chinese sentence delimiter and is used to pad both ends of an input sequence. Thus, where "a" is the first character, the ordered pair (C_{i-1}, C_i) really constitutes the pair "ia". Similarly, where "a" is the last character, the ordered pair (C_{i-1}, C_i) really constitutes the pair "ai". FIG. 4, therefore, indicates that "a" was the first character 322 times (see $C_{i-1} = i, C_i = a$) and that "a" was the last character only 22 times (see $C_{i-1} = a, C_i = i$).

Returning now to FIG. 3, our goal is to find the optimum path through the 9 by 5 matrix (comprising C_{11} through C_{59}) using the frequency information of FIG. 4 to thereby identify the five Chinese characters most likely corresponding to the five Chinese syllables "tai2 wan1 you3 tai2 feng1". More specifically, there are 21,870 possible paths through matrix (i.e., $9 \times 5 \times 6 \times 9 \times 9$). One path would be that shown in the first row in FIG. 3 -- "aceag". Another path would be that shown in the second row -- "bdfbh". Of course, by combining the first and second rows, other possible paths can be derived easily -- e.g., "adfbh", "bdeag", etc. In fact, FIG. 5 shows the 32 combinations possible by combining just the symbols in the first and second rows in FIG. 3. The path "aceag" is listed as the first path in FIG. 5, "aceah" is listed as the second path, "acebg" the third ... and "bdfbh" the last path.

FIG. 5, also, shows to the right of each path the frequency calculation used to derive the optimum path. For example, the frequency calculation (or probability calculation) for the first path "aceag" is derived by looking up in FIG. 4 the individual ordered occurrence frequencies for the following pairs "ia", "ac", "ce", "ea", "ag", "gi" -- namely, 322, 1513, 26, 25, 2, 41. Then these six numbers are multiplied together to derive the score for the use of the Chinese characters "aceag" resulting in the large number 25,967,013,800 shown in line 1 of FIG. 1. The only other probable path is shown on the third line -- namely, "acebg" -- and, in fact, represents the optimum path because the calculated probability number 30,121,736,008 is larger than any other calculated number. Thus, the Chinese characters "acebg" are selected as representing those characters defined by the Chinese symbols "tai2 wan1 you3 tai2 feng1" and, in fact, recite the sentence in Chinese: "Taiwan has typhoons."

In more mathematical terms, I have discovered that the optimal path through the matrix represented by the Chinese characters can be derived by reference to the following formula where "c" refers to a Chinese character and $p(c_i | c_{i-1})$ refers to the probability of Chinese character i, given the previous adjacent occurrence of Chinese character i-1:

$$\operatorname{argmax} \prod_{i=1}^{n-1} p(c_i | c_{i-1})$$

Conceptually one enumerates all of the possible paths through the lattice and then picks the best path according to the above formula. Of course, truly enumerating all possible paths is computationally expensive. However, as indicated in the above formula and previous discussion, the score for the entire path is computed by multiplying out the probabilities (frequencies) for adjacent pairs of characters, and one can therefore significantly reduce the paths that one has to consider by using dynamic programming (= Viterbi algorithm) techniques. In formal terms, note that for any syllable syl_i , and for each $c_{i,j}$ (a Chinese character corresponding to syl_i), one is considering the scores of all paths which end in $c_{i,j}$. Note however, that we only need to keep the best path ending in $c_{i,j}$. This is because when we move on to the next syllable syl_{i+1} and consider all characters $c_{i+1,k}$, we want to compute the scores for paths ending in the pair of characters $c_{i,j}c_{i+1,k}$ (as well as the scores for paths ending in other characters corresponding to the pair $\text{syl}_i \text{syl}_{i+1}$); we do this computation for each path ending in $c_{i,j}$, by multiplying the score for that path by the frequency of occurrence of the pair of characters $c_{i,j}c_{i+1,k}$, $\text{freq}(c_{i,j}c_{i+1,k})$. However, it is clear even before we perform these calculations that we really only need to consider the best scoring path ending in $c_{i,j}$: since the same multiplier $\text{freq}(c_{i,j}c_{i+1,k})$ is used to compute the frequency of each path ending in the pair $c_{i,j}c_{i+1,k}$, the best path ending in the pair $c_{i,j}c_{i+1,k}$ will simply be the best path ending in $c_{i,j}$ concatenated with the character $c_{i+1,k}$. We therefore discard all but the best scoring path leading up to $c_{i,j}$. Thus, rather than keeping around $cc_1 \cdot cc_2 \dots cc_{i-1}$ paths (where cc_m is the number of characters possibly corresponding to syl_m) ending in $c_{i,j}$, we only keep one. To illustrate, let us return to FIG. 5, which, as we have noted, represents a subset of the possible paths through the lattice given in FIG. 3. Suppose that we are considering syllable "wan1", in the third column of FIG. 5. Note that there are four distinct possible paths illustrated in FIG. 5 which end in possible transliterations of "wan1", namely:

- "iac", the initial subpath of the final paths numbered 1-8;
- "iad", the initial subpath of the final paths numbered 9-16;
- "ibc", the initial subpath of the final paths numbered 17-24;
- "ibd", the initial subpath of the final paths numbered 25-32.

Following the above formal description, we can eliminate "ibc" because its score $0 \times 0 = 0$ obviously does not compete with the score of the other path ending in "c", namely "iac", whose score at this point is $322 \times 1513 = 487,186$. (Note that in practice the value 0 is not actually used, but rather an arbitrarily chosen very small valued constant; this is for technical reasons which do not affect the discussion here.) Naturally, there is no way that longer paths "ibce", "ibcf" can compete with "iace" or "iacf", respectively, either; this is because at the point we compute the scores for those paths, we are multiplying the scores for the pairs "ce" and "cf" with the scores for the paths "ibc" and "iac" and since we already know that "ibc" is an inferior candidate to "iac", it follows that "ibce" and "ibcf" are inferior candidates, respectively, to "iace" and "iacf". The subpath "ibc" can therefore be eliminated from further consideration. Both of the paths ending in "d", "iad" and "ibd" have 0 scores, and in principle we could eliminate both of these, but in practice one of them -- this case "iad" -- is kept around; it will be eliminated in later steps. At this point, then, we have eliminated two paths -- "ibc" and "ibd" -- and retained two -- "iac" and "iad". Moving on to the syllable "wan1" in the fourth column of FIG. 5, we now have to consider the following possible paths:

- "iace", the initial subpath of the final paths numbered 1-4;
- "iacf", the initial subpath of the final paths numbered 5-8;
- "iade", the initial subpath of the final paths numbered 9-12;
- "iadf", the initial subpath of the final paths numbered 13-16.

Of those paths ending in "e", namely "iace" and "iade", the former has a value of $322 \times 1513 \times 26 = 12,666,836$ and the latter a value of $322 \times 0 \times 0 = 0$; "iade" is therefore eliminated from further consideration. Of those paths ending in "f", namely "iacf" and "iadf", both have a value of 0 ($322 \times 1513 \times 0$ and $322 \times 0 \times 0$ respectively); as above, one of these -- in this case "iacf" -- is kept around in practice. At this point we have kept only two paths "iace" and "iacf", and have eliminated six other possibilities; "iade" and "iadf" were eliminated on this step; and "ibce", "ibcf", "ibde", "ibdf" were eliminated on the previous step because of the elimination of "ibc" and "ibd" on that step.

The frequency statistics of FIG. 4 used in the illustrative embodiment were derived from a corpus of 2.6 million characters of Chinese newspaper text. Estimates of the probabilities of the appearance of 2 adjacent Chinese characters can be derived by dividing the frequency of occurrence of a sequence by the size of the

corpus. However, since all probability estimates thus derived represent the frequency divided by the corpus size, in practice, the corpus size is omitted from the calculation since it does not affect the maximization (thus the use of frequency rather than estimated probability in the above illustrative description of my method).

- 5 To evaluate the effectiveness of my method, seven short samples of text of differing length were chosen, representative of various writing styles ranging from very classical to colloquial:

i) Ad [classical]

ii) Report [classical]

- 10 iii) Newspaper social commentary taken from the training corpus [semi-classical]

iv) Essay [more colloquial]

v) Narrative [more colloquial]

vi) Short story [colloquial]

vii) Exposition [colloquial].

- 15 The performance of my method is given in the table in terms of percentage correct by character (*hit rate*) for each of the styles; the hit rate from my method is specified in the third column and should be compared with the hit rate achieved by merely picking the most common character given the pronunciation, which is given in the second column:

	Style	Lexical Probabilities Only	My Method
	i [class.]	76%	93%
	ii [class.]	73%	90%
	iii [semi-class.]	76%	98%
5	iv [more coll.]	69%	73%
	v [more coll.]	72%	86%
	vi [coll.]	71%	89%
	vii [coll.]	71%	92%
	Total	73%	90%

- 10 A trend evident in these data is that there is some dependence upon style: my current training corpus is heavily classical in style since it is mostly derived from newspapers. As a consequence, texts (i-iii) which are classical in style are better rendered than the more colloquial texts, with the exception of (vii). I expect that this style dependence would become less marked if the training corpus were expanded to
- 15 include other styles.



Claims:

1. An automated method ^{of} ~~for~~ converting signals from an English language keyboard, the signals representing n phonetic Chinese syllables S_1 to S_n in a text, into signals for input to a computer process, the signals for input representing
5 n Chinese characters C_1 to C_n , the method comprising the steps of:
generating for each signal representing a Chinese syllable S_i a group of signals representing Chinese characters C_{i1} to C_{iz} , collectively referenced as C_{i1-z} , corresponding thereto, to form a lattice of signals representing Chinese characters C_{11-z} to C_{n1-z} ,
10 computing a score for each of one or more paths through the lattice of signals representing Chinese characters C_{11-z} to C_{n1-z} each path reflecting a sequence of n Chinese characters corresponding to the n phonetic Chinese syllables S_1 to S_n ; and
generating signals representing n Chinese characters C_1 to C_n based on
15 the path through the lattice having a score reflecting a most likely sequence of character signals in the lattice;
CHARACTERIZED IN THAT:
said computing step comprises deriving a probability of the use of adjacent Chinese characters C_i and C_{i-1} in said text based upon a frequency of the
20 ordered appearance of said adjacent Chinese characters in a large corpus of Chinese text, and determining a score for each of the one or more paths through the lattice by multiplying together the derived probabilities.
2. The method of claim 1 further characterized in that:
said computing step is effected using dynamic programming techniques.
- 25 3. The method of claim 1 wherein the corpus of Chinese text comprises one or more corpora of Chinese texts representative of one or more Chinese language styles.
4. An automated method ^{of} ~~for~~ converting signals from an English language keyboard, the signals representing one or more phonetic syllables of
30 Chinese, into signals for input to a computer process, the signals representing one or more Chinese characters, the method comprising the steps of:



generating for each signal representing one or more phonetic syllables a group of signals representing one or more Chinese characters corresponding thereto to form a lattice of signals representing characters; and

- 5 computing a measure of likelihood for each of one or more paths through the lattice of signals representing characters, each path reflecting a sequence of characters corresponding to the one or more phonetic syllables, the measure of likelihood of the path based upon one or more measures of likelihood of occurrence of successive characters of the path appearing in a corpus of Chinese text; and
- 10 generating signals representing a plurality of Chinese characters based on the path through the lattice having a measure of likelihood reflecting a most likely sequence of character signals in the lattice

CHARACTERIZED IN THAT:

- the computing step comprises: deriving for a path a measure of likelihood of the path's sequence of Chinese characters by combining one or more
- 15 measures of likelihood of occurrence of successive pairs of characters of the path appearing in a corpus of Chinese text.

5. The method of claim 4 wherein measures of likelihood for successive pairs of character signals are combined by multiplication.

6. The method of claim 4 wherein the corpus of language text comprises

20 one or more corpora of language texts representative of one or more language styles.

7. The method of claim 4 wherein the measure of likelihood is a probability.

8. The method of claim 4 wherein the measure of likelihood is a frequency.