

Technical Reports

ガチャ align

Sentence Alignment

Task:

- To align sentences to its corresponding translation

Assumption:

- Sentence alignment can be done monotonically

Challenges:

- non-1:1 alignments, insertions, deletions, incomplete translations

Assumption: Sentence alignment can be done monotonically

On the third day of this much-anticipated exhibition , the event will be open to the general public as a ticketed shopping event called the Blueprint Emporium .

Held in collaboration with Zouk , be prepared to be seen at this chic fashion party on 1 May at the F1 Pit Building , and take home limited edition and past season samples from Asian designers and labels that cannot be found anywhere else .

この大きな期待がされている展示会の3日目には、このイベントは一般市民に対してもチケット制のイベントとして開かれ、また、ブループリント百貨店と呼ばれる買い物イベントが、伝説的な地元のナイトクラブ・ズークとの共同で開かれます、最先端のファッションの才能が披露されます。

(*black text doesn't get translated)

Challenges: non-1:1 alignments, insertions, deletions, incomplete translations

Aiming to attract about 3000 visitors, the Blueprint Emporium allows exhibitors to test their brands on the Singapore market, as this event will showcase offerings from cutting-edge brands, many of which are not available commercially in Singapore.

一晩だけでも、予想観客動員数の300人は多目的の、50人のデザイナーとラウル（RAOUL）、オールドレスアップ（alldressedup）やウィキッド・ソング（Wykidd Song）などのデザイナーや一流地元ブランドを含む国内外の市場からの集団を収容することができるファッションクラブ空間で独特に圧倒的な魅力のある体験に迎えられます。

Sentence Alignment Approaches

Lexical methods

- corresponding sentences contain more corresponding words

Length-based methods

- sentences that correspond to each other are also similar in length (characters or words)

Combined methods

- use lexical cues in length-based settings

Gale-Church's Length Based model

- Define a distance based on the costs of aligning source to target sentences (for a fixed finite set of possible alignment types)
- Minimize this distance by finding the best alignment using dynamic programming → recursive definition of

$$D(i, j) = \min \left\{ \begin{array}{ll} D(i, j-1) & + \text{cost}(\text{align}_{0:1}, 0, t_j) \\ D(i-1, j) & + \text{cost}(\text{align}_{1:0}, s_i, 0) \\ D(i-1, j-1) & + \text{cost}(\text{align}_{1:1}, s_i, t_j) \\ D(i-1, j-2) & + \text{cost}(\text{align}_{1:2}, s_i, t_{j-1}..t_j) \\ D(i-2, j-1) & + \text{cost}(\text{align}_{2:1}, s_{j-1}..s_j, t_j) \\ D(i-2, j-2) & + \text{cost}(\text{align}_{2:2}, s_{j-1}..s_j, t_{j-1}..t_j) \end{array} \right.$$

Gale-Church Cost Function

- assume that each character in the source language generates c characters in the target language with variance, s^2 and distance function:
 - distance $\delta = (srclen - trglen * c) / \sqrt{srclen * s^2}$ is normally distributed and $P(\delta | aligntype)$ gives the probability of observing a specific length-pair
 - define prior probabilities of $P(aligntype)$
- finally the cost function:
- $$\log P(aligntype) P(\delta | aligntype)$$

Gale-Church Algorithm

- compute alignment costs for each sentence pair (i, j)
- start with 0^{th} source, 0^{th} target sentences and fill the entire table
- read the alignment path with minimal costs

Gale-Church Parameters

- empirically find parameters c , $s2$ and $P(\textit{aligntype})$ from example corpora
- Gale-Church used a German-English corpus and defined $c = 1$, $s2 = 6.8$

$$P(\textit{aligntype} = 1 : 1) = 0.89$$

$$P(\textit{aligntype} = 1 : 0) = 0.0099$$

$$P(\textit{aligntype} = 0 : 1) = 0.0099$$

$$P(\textit{aligntype} = 2 : 1) = 0.089$$

$$P(\textit{aligntype} = 1 : 2) = 0.089$$

$$P(\textit{aligntype} = 2 : 2) = 0.011$$

Gale-Church Tweak

- fixed parameters are based on German-English corpus but it works surprisingly well for most European language pairs.
- **What if we apply Gale-Church to non-European Languages?**
- **What if we tweak the fixed parameters?**

ガチャ align (GaCha align)

- An experiment to test how parameters, $\{c, s2$ and $P(\textit{aligntype})\}$ affects the accuracy of alignment for an English-Japanese corpus

Tasks:

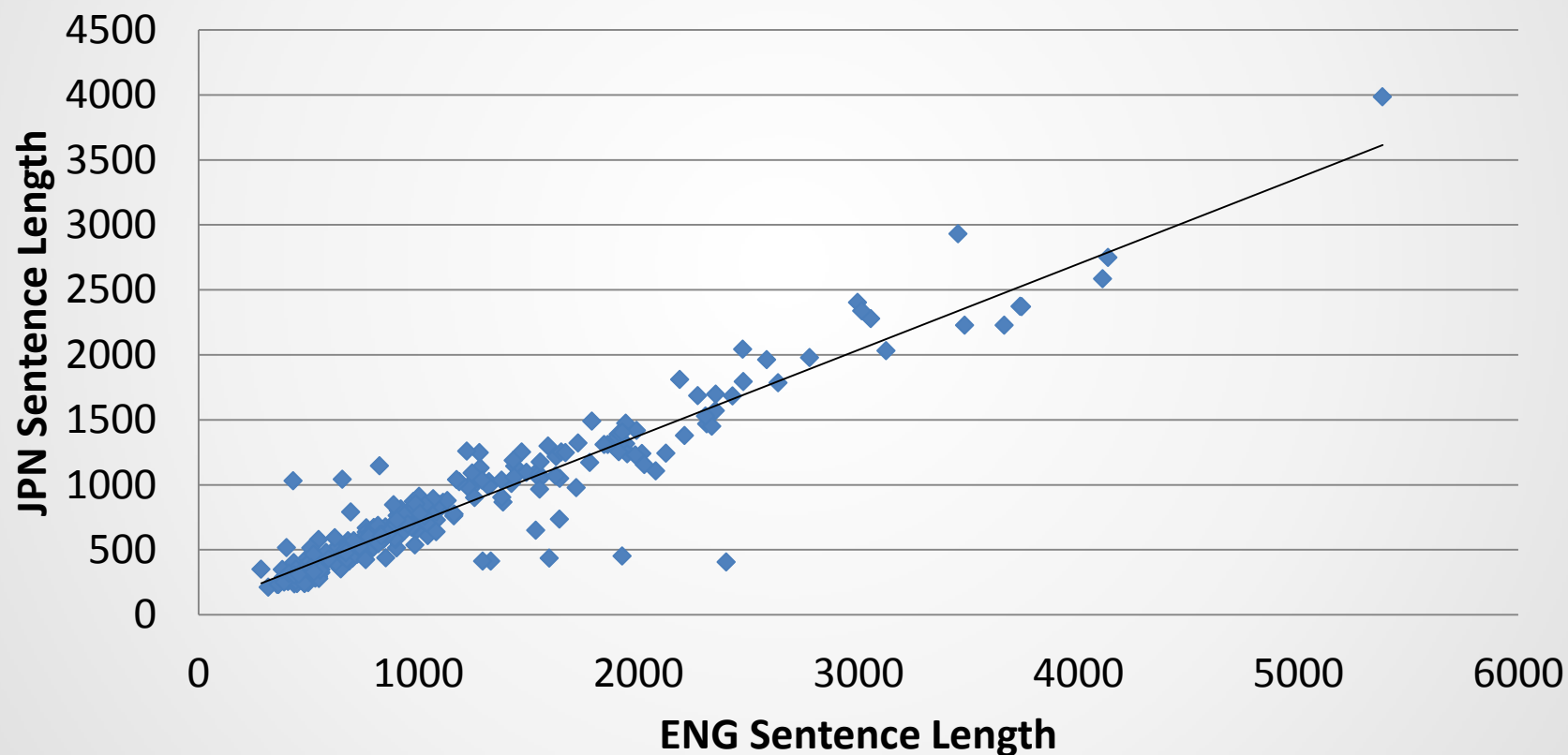
- Run Gale-Church alignments
 - with calculated parameters from the corpus
 - to determine what is the optimal value for the c and $s2$ for best accuracy

Data

- 1853 human-aligned ENG-JPN sentences from NTU-MC

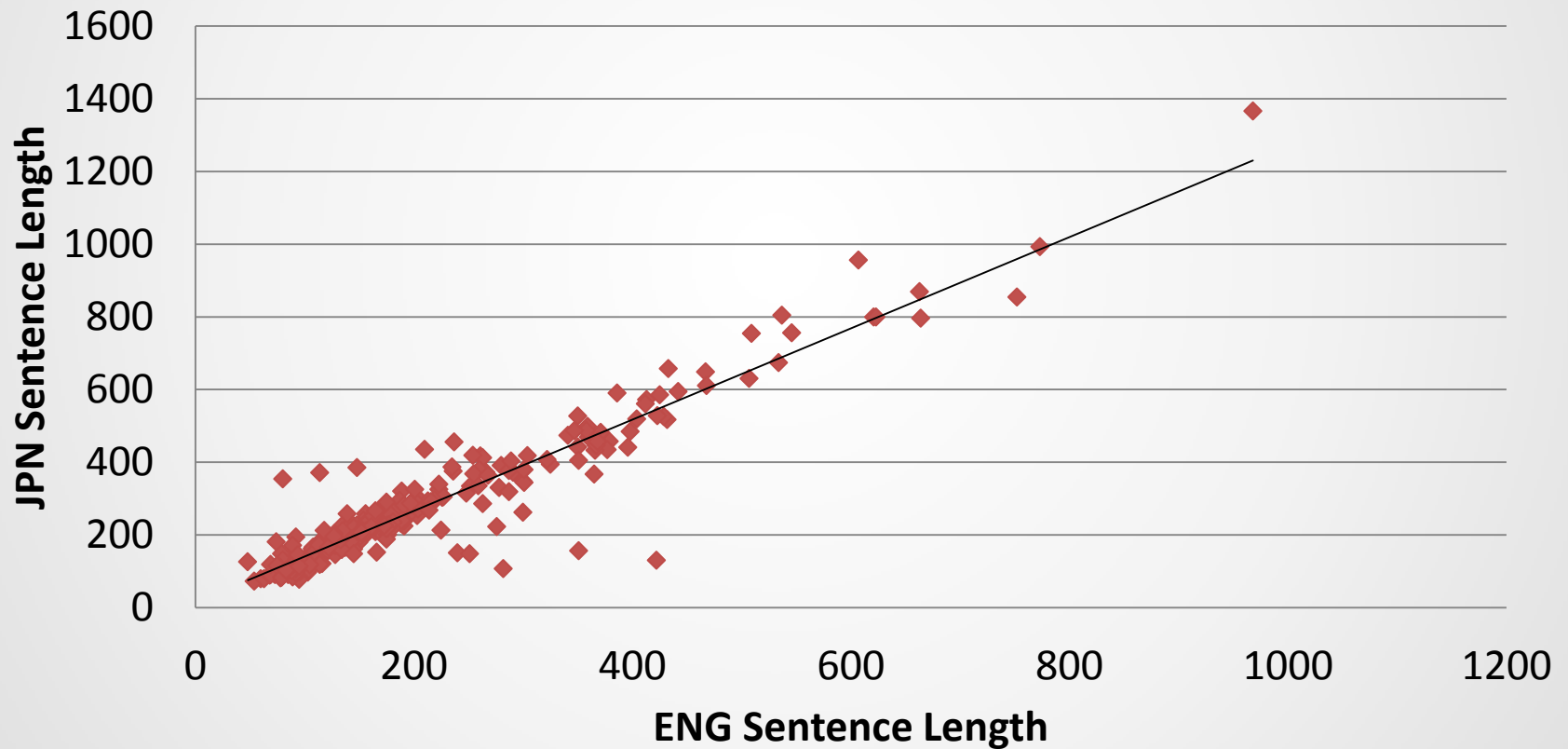
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Character Length $c = 0.711$, pearsonr = 0.9481



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Word Length $c = 1.33$, pearsonr = 0.9494



Results with calculated c and s^2

	Char-based	Word-based
c (mean)	0.711	1.332
s^2 (variance)	416.89	77.64

ガチャ align (c, s2 tweak)

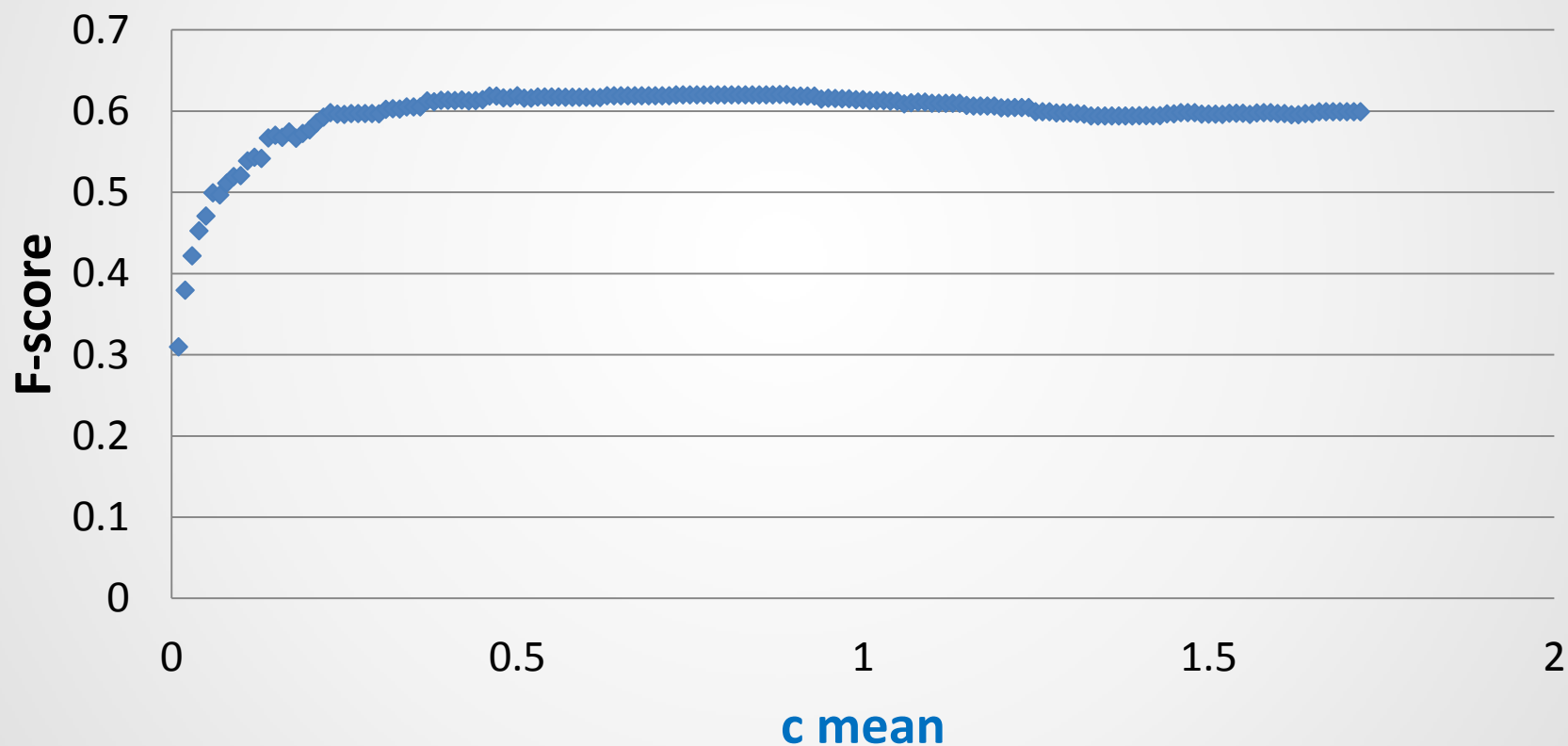
char-based	c	s2	Recall	Precision	F-score
Default	1	6.8	0.6657	0.5637	0.6104
c tweak	0.711	6.8	0.6724	0.5731	0.6188
s2 tweak	1	416.89	0.6224	0.5206	0.5670
c + s2 tweak	0.711	416.89	0.5758	0.4868	0.5276
word-based	c	s2	Recall	Precision	F-score
c tweak	1.33	6.8	0.6413	0.5531	0.5940
s2 tweak	1	77.64	0.6552	0.5483	0.5970
c + s2 tweak	1.33	77.64	0.6618	0.5602	0.6068

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	Original Gale-Church	Calculated from NTU-MC (eng-jpn)
$P(\text{aligntype} = 1 : 1)$	0.89	0.72153
$P(\text{aligntype} = 1 : 0)$	0.0099	0.05288
$P(\text{aligntype} = 0 : 1)$	0.0099	0.03022
$P(\text{aligntype} = 2 : 1)$	0.89	0.00702
$P(\text{aligntype} = 1 : 2)$	0.89	0.16352
$P(\text{aligntype} = 2 : 2)$	0.11	0.00216
$*P(\text{aligntype} = 1 : 3)$	-	0.01619
$*P(\text{aligntype} = 1 : 4)$	-	0.01619
$*P(\text{aligntype} = \text{others})$	-	0.00377
F-score with c =0.711	0.6188	0.6243 , *0.6091

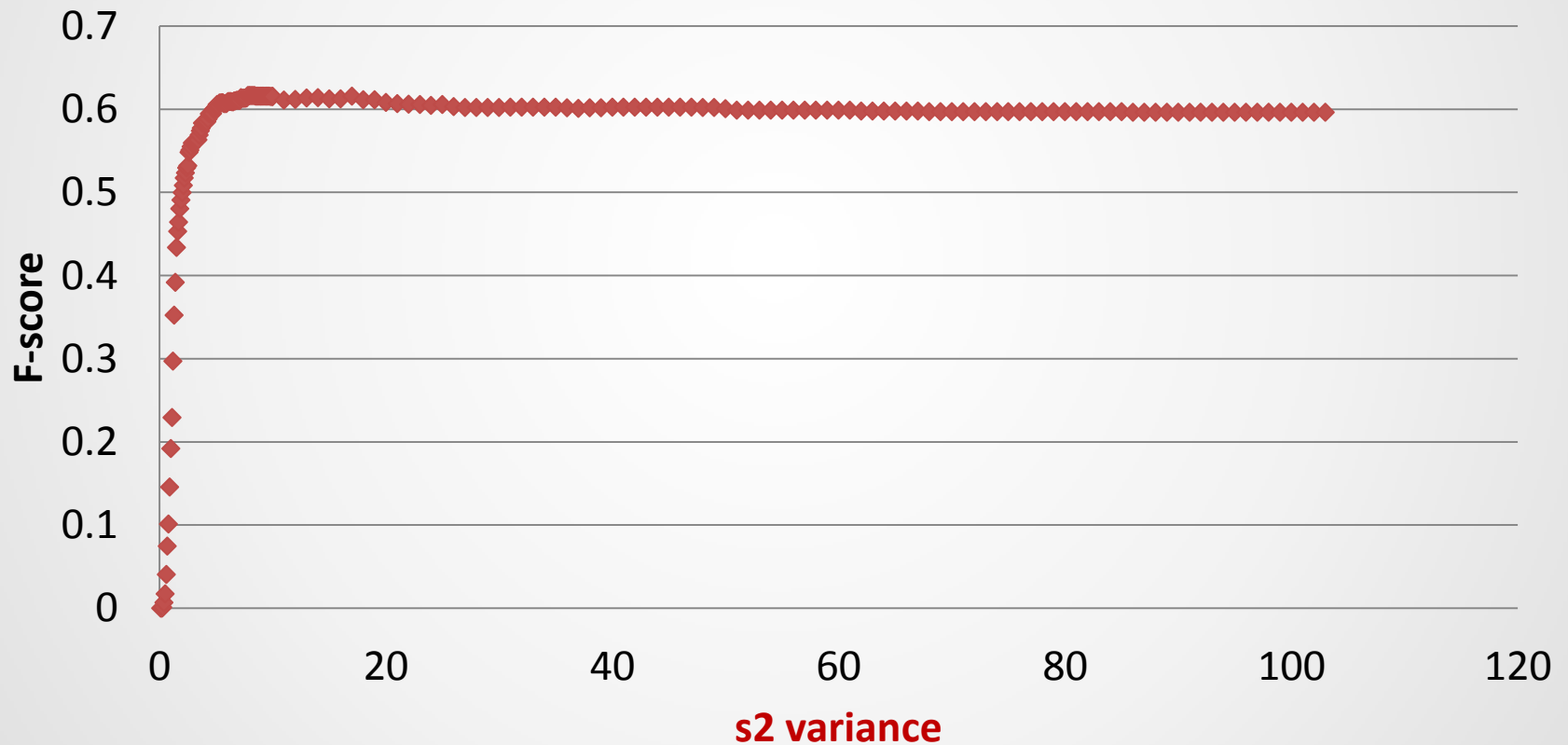
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Maximum F-score = 0.6201 @ mean = 0.88 to 0.89



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Maximum F-score = 0.6162 @ variance = 7.8 to 8.4



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Conclusion

- Gale-Church algorithm seems to be robust enough to be unaffected by language specific *length proportion* or *alignment types*
- Tweaking *character mean* between the src and trg text can be done simply and we have shown that it improves accuracy.
 - default: $c=1.00$, $s2=6.8$, $f\text{-score} = 0.6104$
 - calculated mean: $c=0.71$, $f\text{-score} = \underline{0.6188}$
 - optimal: $c=0.88$, $s2=7.8$, $f\text{-score} = 0.6199$
- Using text dependent *alignment types probabilities* don't affect performance too
 - default: $f\text{-score} = 0.6199$ tweaked: $f\text{-score} = \underline{0.6290}$

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Future Works

- Hybrid model by
 - using a dictionary and adding a weight to the length based:

$$\text{dic_weight} * \log P(\text{aligntype}) P(\delta | \text{aligntype})$$

English	Japanese
<p>Char kway teow , loosely translated as " stir-fried rice cake strips " , is made by stir-frying flat rice noodles (similar to the Italian tagliatelle) with light and dark soy sauce , a dash of belachan (shrimp paste) , tamarind juice , bean sprouts , Chinese chives , lap cheong (Chinese sausages) and cockles .</p>	<p>チャー・クウェイ・ティオは、“炒めた平たい米麺”を意味します。</p> <p>炒めた平たい米麺（イタリア料理のタリアテーレに似ている）を、薄味の醤油や少量のベラチャン（エビのすり身）、タマリンドジュース、もやし、ニラ、ラブチョン（腸詰）、ザルガイなどと伴に炒めて作ります。</p>
<p>In its original recipe , the rice noodles are also stir-fried in pork fat using crisp bits of pork lard , resulting in a distinctively rich taste .</p>	<p>オリジナルのレシピでは、少量のポークラードや豚脂を使って米麺も一緒に炒め、格別に濃厚な味にします。</p>
<p>In recent years , the dish as evolved into a healthier version with hawkers serving up more greens and adding less oil .</p>	<p>昨今の屋台では、緑黄色野菜をより多く用いて油分も控えめにした、より健康的なものを提供しています。</p>
<p>Char kway teow is easily available at most food centres in Singapore , such as at the Maxwell Road Hawker Centre , and it's also a signature dish at the Princess Terrace Café .</p>	<p>チャー・クウェイ・ティオは、「マックスウェル・ロード・ホーカーセンター」などのフードセンターで見つけることができます。</p> <p>また、「プリンセス・テラス・カフェ」の名物料理でもあります。</p>