## MT Summer Term 2021 Ex2: SMT Intuition; Probability and Noisy Channel.

- 1. What is great about RBMT?
- 2. What is not so good about RBMT?
- 3. Why do we try to use machine learning or statistical estimation from data for MT?



4. In your own words, explain  $\hat{e} = \underset{e}{\operatorname{argmax}} P(e|f)$ .



- 5. What kinds of data do we need for SMT?
- 6. Given this bitext, which symbol is the likely Chinese symbol for "chicken"? Which symbol is the likely symbol for "soup"?

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					CLASSIC SOUPS Sm.	Lg.
六	炒	雞	2	57.	House Chicken Soup (Chicken, Celery,	
					Potato, Onion, Carrot)	2.75
雞	1	飯		58.	Chicken Rice Soup1.85	3.25
雞	3	麵		59.	Chicken Noodle Soup1.85	3.25
鹰	東	李	杏	60.	Cantonese Wonton Soup1.50	2.75
*	茄	季	-	61.	Tomato Clear Egg Drop Soup1.65	2.95
3	2	5	备	62.	Regular Wonton Soup1.10	2.10
酸	身	*	*	63. ₹	Hot & Sour Soup	2.10
季	7	主		64.	Egg Drop Soup1.10	2.10
李	7	F	*	65.	Egg Drop Wonton Mix1.10	2.10
豆	窟	莱	*	66.	Tofu Vegetable SoupNA	3.50
雞	Æ.	米	3	67.	Chicken Corn Cream SoupNA	3.50
23	图 3	E 米	:	68.	Crab Meat Corn Cream SoupNA	3.50
海	1	¥	*	69.	Seafood SoupNA	3.50

7. Given the following bitext and word alignment (indicated in terms of colour codes), (i) estimate a word based probabilistic translation dictionary (a translation model), (ii) find the best

I love the boy.
J'aime le garçon.
I love the dog.
J'aime le chien.
They love the dog.
Ils aiment le chien.
They talk to the girl.
Ils parlent à la fille.
I talk to the mother.
Je parle à la mère.

"translations" of



- They love the girl
- I talk to the dog

into French under the model and (iii) compute the probabilities for the best translations under the model based on the word-based translation probabilities, assuming that the probabilities are independent of each other:  $P(f_1 f_2 \dots f_n | e_1 e_2 \dots e_n) = \prod_{i=1}^n P(f_i | e_i)$ .

- 8. Given a sequence of n numbers/measurements/numerical observations  $x_1x_2 \cdots x_n$ , please define
  - Population mean, sample mean



- Population variance, sample variance
- Population standard deviation, sample standard deviation
- 9. Explain the notions of sample space, outcome and event in set-based formalisations of probability.



- 10. What is a Laplace experiment and a Laplace probability?
- 11. Probabilities can be estimated from counts (relative frequencies). Give the following observations, estimate the probabilities:

	Car	Lorry	M-cycle	Cycle	Pedes.	n
#(x)	124	49	7	64	271	515
$\frac{\#(x)}{n}$						
P(x)						

12. Please complete the following table describing the Boolean Algebra of events:

$$A \cap B = B \cap A$$

$$A \cap (B \cap C) =$$

$$A \cap (B \cup C) =$$

$$A \cap A = A$$

$$A \cap (A \cup B) = A$$

$$\overline{A \cap B} =$$

$$A \cap \emptyset = \emptyset$$

$$A \cap \bar{A} =$$

$$\bar{\bar{A}} = A$$

$$A \cup B =$$

$$A \cup (B \cup C) = (A \cup B) \cup C$$

$$A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$$

$$A \cup A =$$

$$A \cup (A \cap B) =$$

$$\overline{A \cup B} = \overline{A} \cap \overline{B}$$

$$A \cup \Omega = \Omega$$

$$A \cup \bar{A} =$$

13. Define conditional probability:

$$P(A|B) =_{def} \dots$$



14. Given a fair 6-sided dice, what are

$$P(odd) = \cdots$$

$$P(odd|prime) = \cdots$$

$$P(prime|odd) = \cdots$$

$$P(even|prime) = \cdots$$

$$P({5}|odd) = \cdots$$

$$P(\{2,5\}|odd) = \cdots$$

$$P(\emptyset|odd) = \cdots$$

- 15. When are two events mutually exclusive, when are two events independent?
- 16. Give the specific and the general version of the addition rule of probabilities:

$$P(A \cup B) = \cdots$$

$$P(A \cup B) = \cdots$$

17. Give the specific and the general version of the multiplication rule of probabilities:

$$P(A \cap B) = \cdots$$

$$P(A \cap B) = \cdots$$

- 18. Give the complement rule of probability:  $P(\bar{A}) = \cdots$
- 19. Expand the following using the chain rule of probability:  $P(w_1 \ w_2 \ ... \ w_n) = \cdots$
- 20. What is the prior, the likelihood and the posterior in Bayes rule:

$$P(A|B) = \frac{P(B|A) \times P(A)}{P(B)}$$

- 21. Prove Bayes Rule.
- 22. Why is Bayes rule useful?
- 23. In your own words, relate the fundamental rule of statistical machine translation (SMT) to the noisy channel model (NC):

$$\hat{e} = \underset{e}{\operatorname{argmax}} P(e|f) = \underset{e}{\operatorname{argmax}} \frac{P(f|e) \times P(e)}{P(f)} = \underset{e}{\operatorname{argmax}} P(f|e) \times P(e)$$

What is the translation model, what is the language model, what is the source model, what is the channel model, what is the prior, the likelihood and the posterior? In what sense is this a MAP (maximum a posteriori) decision rule?

- 24. Rule of total probability: can you show in terms of a drawing why  $P(A) = P(A \cap B) \cup P(A \cap \overline{B})$
- 25. Total probability: why is  $(A \cap B) \cup P(A \cap \overline{B}) = P(A|B) \times P(B) + P(A|\overline{B}) \times P(\overline{B})$