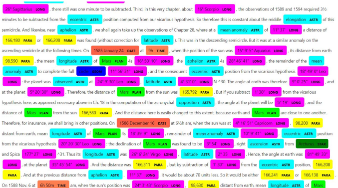
Refined Name Enitiy Recognition (NER) by A Customized SpaCy Model and Pattern Rules of RegEx

January 14, 2020



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- **Data Publishing.**

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- Proposing some ideas for the future work toward relation extraction and causal inference.

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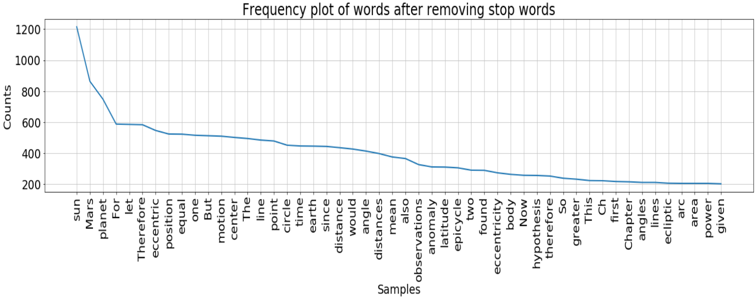

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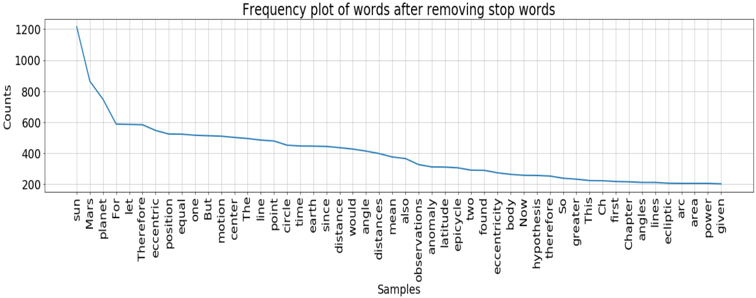
- Corpus includes 169231 tokens (roughly speaking; words) and 9513 unique tokens



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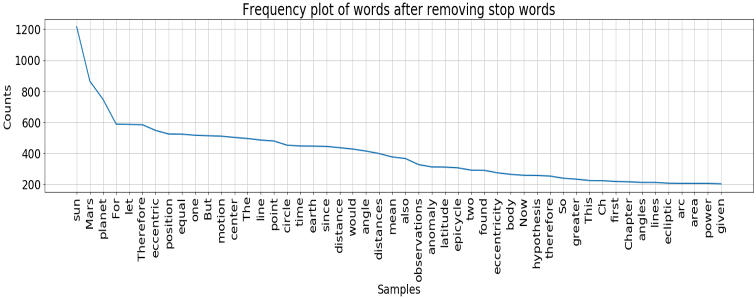
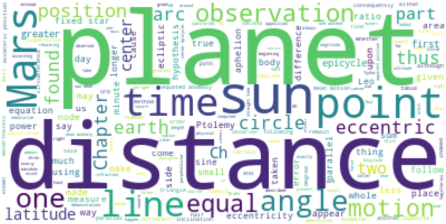


- You can find here the word cloud of the whole book and by chapters It can give us some simple initial and simple intution



	Sentence	SentIndex	Chapter	Paragraph	ParaIndex	ASO	Entities	CNER	La
158	On 1580 November 12 at 10h 50m,1 they set Mars down at 8° 36′ 50″ Gemini2 without mentioning the horizontal variations, by which term I wish the diurnal parallaxes and the refractions to be understood in what follows.	967	10	On 1580 November 12 at 10h 50m,1 they set Mars down at 8° 36′ 50″ Gemini2 without mentioning the horizontal variations, by which term I wish the diurnal parallaxes and the refractions to be understood in what follows. Now this observation is distant and isolated. It was reduced to the moment of opposition using the diurnal motion from the Prutenic Tables ^3.	218	{'act': 'set', 'subject': 'they', 'obj': 'Mars '}	[[1580 November 12, DATE], [10h 50m,1 they, TIME], [Mars, PLAN], [8° 36′ 50″, LONG], [diurnal, ASTR]]	[1, 1, 1]	
168	On 1582 December 28 at 11h 30m, they set Mars down at 16° 47′ Cancer by observation ^6.	977	10	On 1582 December 28 at 11h 30m, they set Mars down at 16° 47′ Cancer by observation ^6. The moment of opposition assigned by Tycho comes 46 minutes later, during which the planet retrogressed less than one minute. Tycho therefore puts it at 16° 46′ 16″ Cancer ^7. On an inserted sheet here, an attempt was made to correct for a refraction of two minutes. This was, I think, first trial of the theory of refraction then being developed. Nevertheless, he followed the observed value unchanged, thus declining to consider the planet as something which could alter its position. Nor was there any need for correction, since it was in Cancer, beyond the reach of refraction, and was in mid-sky where, in Cancer, there is no longitudinal parallax.	221	{'act': 'set', 'subject': 'they', 'obj': 'Mars '}	[[1582 December 28, DATE], [11h 30m, TIME], [Mars, PLAN], [16° 47′ Cancer, LONG], [observation, ASTR]]	[1, 1, 1]	
175	On 1585 January 31 at 12h 0m, Mars was placed at 21° 18′ 11″ Leo ^8.	984	10	On 1585 January 31 at 12h 0m, Mars was placed at 21° 18′ 11″ Leo ^8. The diurnal motion, by comparison of observations, was 24′ 15″. The moment of opposition followed at 19h 35m, 7 hours and 35 minutes later. To this period belongs 7′ 41″ of diurnal motion westward. Therefore, at the designated moment, it would have been at 21° 10′ 30″ Leo, which is what was accepted. There is no mention of parallax. Nothing had to be done about refraction, because Mars was high and at mid-sky. I therefore find the bit of advice in the table about refraction (properly) ignored.	222	{'act': 'placed'}	[[1585 January 31, DATE], [12h 0m, TIME], [Mars, PLAN], [21° 18′ 11″ Leo, LONG]]	[1, 1, 1]	

Here you can see first 10 sentences with different attribute that has been add by our analysis and can be used later for classification,relation extraction and...

- Data is annotated entity by entity using regex pattern.

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- Annotated data is merged and now the training data is ready!

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- A model word representation Bert with100 iteration and batch size 16 has been used for NER classifcation

Metrics	ents_p'	ents_r	ents_f	
GEOM	100	99.85	99.92	
LONG	99.76	99.88	99.82	
PARA	98.51	99.76	99.13	
TIME	97.97	97.00	97.48	
STAR	84.61	74.15	79.04	

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- The evaluation result per entity and overall is calculated by comparing with gold stadnard format as follows:

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- you can see here an example of explicit observation extraction.

183	On 1587 March 7 at 19h 10m they deduced the position of Mars from the observations, which was 25° 10′ 20″ Virgo.	On 1587 March 7 at 19h 10m they deduced the position of Mars from the observations, which was 25° 10′ 20″ Virgo. This they kept in the table, but changed the time to 17h 22m. The difference of 1h 48m multiplied by a diurnal motion of 24′ gives the same number of minutes and seconds (that is, 1′ 48″), no more. It therefore should have been 25° 8′ 32″ Virgo, which also approaches nearer the point opposite the sun. The difference is of practically no importance ^9.

- you can see here an example of explicit observation extraction.
- moreover, you can find how the text has been structured.

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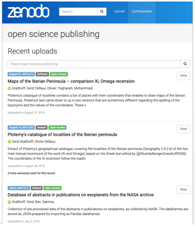

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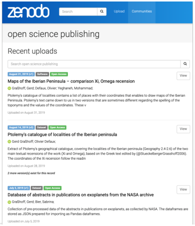

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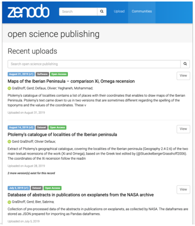

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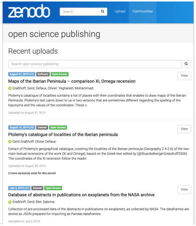

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- find the best way to presnt the result in way that non-technical readers could be able to go thorough the text and computational analysis (voila)

Ma, Y.; Zhou, G.; Wang, S.; Zhao, H.; Jung, W. SignFi: Sign Language Recognition Using WiFi. Proc. ACM Interact. Mob. Wearable Ubiquitous Technol. 2018, 2, 23.

Muller, Machine Learning and AI for the sciences – Towards Understanding