# 運用機器語言模型預測與分析中文媒體之報導框架

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# Introduction - 動機

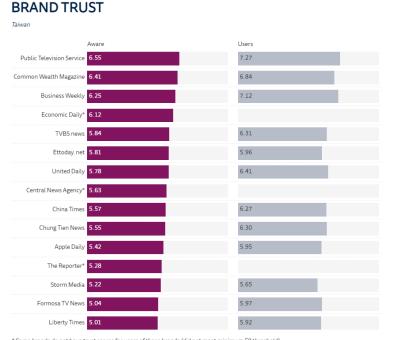
• 台灣媒體的現況:

現代獲取報導更為快速、便利,民眾經常閱讀單一媒體的習慣

媒體開源節流,聳動下標、立場報導,以吸引更多受眾;

賣新聞⇔立場報導; Internet penetration (Taiwan):88%

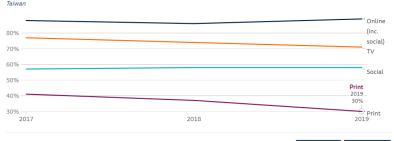
# NEWS SOURCES LAST WEEK: TAIWAN Online Used 3 days or more Used last week Yahool News 36% 51% ETtoday online 29% 49% TVBS News online 20% 31% EBS News online 16% 28% Apple Daily online 15% 28% ChungTien News online 15% 25% Sanlih E-Television News online 13% 24% United Daily online 11% 22% Storm Media 5% 16% Liberty Times online 9% 16% Common Wealth Magazine online 5% 13% China TV News online 6% 11% Next Magazine online 4% 11% Business Weekly online 5% 11%







**Trust** 







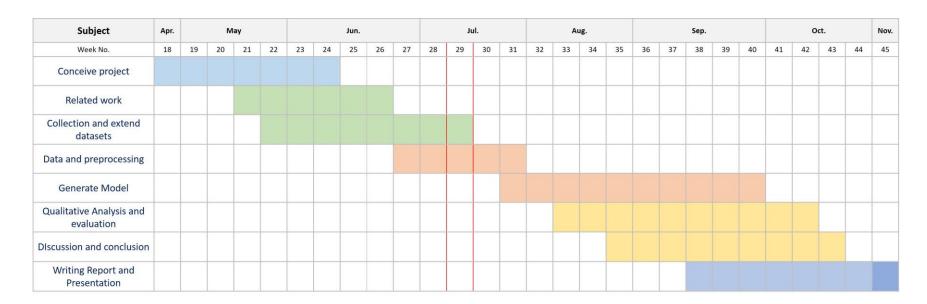
#### Introduction - 動機

- 現有技術與應用 (媒體):
  - NLP媒體研究:假新聞分析、情感分析等
  - 針對假新聞 台灣現有查核機構:「台灣事實查核中心」、「Cofacts 真的假的」、「MyGoPen」與「蘭姆酒吐司」等
  - 資策會將AI自然語言處理模型BERT 與IFCN認證機構結合,提前過濾假訊 息,提升人工查證的效率
  - 但現在並沒有框架立場報導的分析系統



#### Introduction - 目的與規劃

- 改善問題:
  - AI模型預測與分析報導框架 / 立場
  - 輔助受眾提升媒體識讀,促進民眾對於報導議題的全面了解
  - 進而讓受眾可以有效選擇媒體。



3步看懂媒體風向

▲ | 說了什麼? | 沒說什麼?

2 服務了誰?

3 | 事買優先

#### 過去研究參考

- Kristen Johnson, Di Jin, & Dan Goldwasser (2017). Leveraging Behavioral and Social Information for Weakly Supervised Collective Classification of Political Discourse on Twitter. Proceedings of the 55th Annual Meeting of the Association for Computational Linguistics, pages 741–752. https://doi.org/10.18653/v1/P17-1069
- 研究中使用184,914則美國國會議員的推文 (tweets)作為文本基礎
- 建立弱監督global PSL models可用於英文文本的框架(frames)分析
- 最終模型成功提升unsupervised F1 score by 21.52 points.

ISSUE	Frames																
1330E	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17
Abortion	4	7	23	55	40	0	2	32	10	0	4	46	20	0	1	13	8
ACA	65	9	6	28	24	0	3	128	21	3	18	116	174	2	21	100	15
Guns	2	2	37	16	30	21	93	8	36	14	49	166	65	0	5	55	147
Immigration	16	7	6	6	42	3	15	0	29	19	7	81	52	1	1	32	2
LGBTQ	0	0	9	99	23	2	2	3	10	17	7	39	14	1	2	11	48
Terrorism	6	4	46	3	11	10	115	1	6	13	14	69	68	35	6	99	57

Table 7: Count of Each Type of Frame Per Issue in Labeled Dataset.

#### ISSUE AND KEYWORDS OR PHRASES

ABORTION: abortion, pro-life, pro-choice, Planned Parenthood, StandWithPP, Hobby Lobby, birth control, women's choice, women's rights, women's health

ACA: patient protection, affordable care act, ACA, obamacare, health care, healthcare, Burwell, Medicare, Medicaid, repeal and replace

GUNS: Charleston, gun, shooting, Emanuel, Second Amendment, Oregon, San Bernadino, gun violence, gun control, 2A, NRA, Orlando, Pulse

IMMIGRATION: immigration, immigrants, illegal immigrants, border, amnesty, wall, Dreamers, Dream Act

LGBTQ: equality, marriage, gay, transgender, marriage equality, same sex, gay marriage, religious freedom, RFRA, bathroom bill

TERRORISM: terrorism, terrorists, terror network, ISIS, ISIL, Al Qaeda, Boko Haram, extremist

Table 8: Keywords or Phrases Used to Filter Tweets for Issue.

#### FRAME NUMBER, FRAME, AND ADAPTED UNIGRAMS

- 1. ECONOMIC: premium(s), small, business(es), tax(es), economy, economic, cost(s), employment, market, spending, billion(s), million(s), company, companies, funding, regulation, benefit(s), health
- CAPACITY & RESOURCES: resource(s), housing, infrastructure, IRS, national, provide(s), providing, fund(s), funding, natural, enforcement
- 3. MORALITY & ETHICS: moral, religion(s), religious, honor(able), responsible, responsibility, illegal, protect, god(s), sanctity, Islam, Muslim, Christian, radical, violence, victim(s), church
- 4. FAIRNESS & EQUALITY: fair(ness), equal(ity), inequality, law(s), right(s), race, gender, class, access, poor, civil, justice, social, women(s), LGBT, LGBTQ, discrimination, decision(s)
- 5. LEGALITY, CONSTITUTIONALITY, & JURISDICTION: right(s), law(s), executive, ruling, constitution(al), amnesty, decision(s), reproductive, legal, legality, court, SCOTUS, immigration, amendment(s), judge, authority, precedent, legislation
- 6. CRIME & PUNISHMENT: crime(s), criminal(s), gun(s), violate(s), enforce(s), enforced, enforcement, civil, tribunals, justice, victim(s), civilian(s), kill, murder, hate, genocide, consequences
- 7. SECURITY & DEFENSE: security, secure, defense, defend, threat(s), terror, terrorism, terrorist(s), gun(s), attack(s), wall, border, safe, safety, violent, violence, ISIS, ISIL, suspect(s), domestic, prevent, protect
- HEALTH & SAFETY: health(y), care, healthcare, obamacare, access, disease(s), mental, physical, affordable, coverage, quality, (un)insured, disaster, relief, unsafe, cancer, abortion
- 9. QUALITY OF LIFE: quality, happy, social, community, life, benefit(s), adopt, fear, deportation, living, job(s), activities, family, families, health, support
- 10. CULTURAL IDENTITY: identity, social, value(s), Reagan, Lincoln, conservative(s), liberal(s), nation, America, American(s), community, communities, country, dreamers, immigrants, refugees, history, historical
- 11. Public Sentiment: public, sentiment, opinion, poll(s), turning, survey, support, American(s), reform, action, want, need, vote
- 12. POLITICAL FACTORS & IMPLICATIONS: politic(s), political, stance, view, (bi)partisan, filibuster, lobby, Republican(s), Democrat(s), House, Senate, Congress, committee, party, POTUS, SCOTUS, administration, GOP
- 13. POLICY DESCRIPTION, PRESCRIPTION, & EVALUATION: policy, fix(ing), work(s), working, propose(d), proposing, proposal, solution, solve, outcome(s), bill, law, amendment, plan, support, repeal, reform
- 14. EXTERNAL REGULATION AND REPUTATION: regulation, US, ISIS, ISIL, relations, international, national, trade, foreign, state, border, visa, ally, allies, united, refugees, leadership, issues, Iraq, Iran, Syria, Russia, Europe, Mexico, Canada 15. FACTUAL: health, insurance, affordable, deadline, enroll, sign, signed, program, coverage
- 16. (SELF) PROMOTION: statement, watch, discuss, hearing, today, tonight, live, read, floor, talk, tune, opinion, TV, oped
- 17. PERSONAL SYMPATHY & SUPPORT: victims, thoughts, prayer(s), pray(ing), family, stand, support, tragedy, senseless, heartbroken, people, condolences, love, remember, forgive(ness), saddened

Table 9: Frame and Corresponding Unigrams Used for Initial Supervision.

- 文本蒐集與建立
  - Web Crawler
  - 針對台灣六家中文媒體的政治新聞:

中央社、華視新聞網、中國時報、自由時報、民視新聞網、東森ET Today

• 劃分Datasets: (初期將使用約30,000筆data)

<Train> <Test> <Inference>

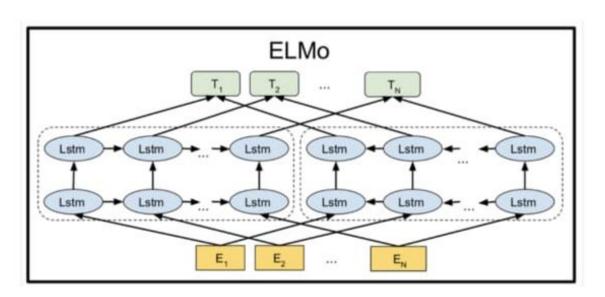


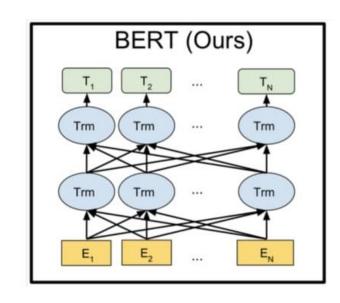
- 資料標註
  - 依政治立場:
  - <DPP> <KMT> <無>
  - 依媒體框架:
  - <經濟> <民生> <國防> <政治與政策> <道德>
  - <法律與平等> <健康與安全> <置入行銷> <外交>

(需事先設定標籤的關鍵字詞)



- 建立模型
  - 期望能在BERT、ELMo架構下做到遷移學習
  - 主要先嘗試建立Linear SVC、Naive Bayes等模型
  - 甚至RFC、SVM等多種方法





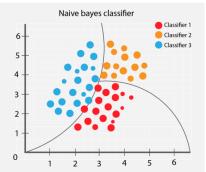
#### Naive Bayes

**thatware.co** 

In machine learning, naive Bayes classifiers are a family of simple "probabilistic classifiers" based on applying Bayes' theorem with strong (naive) independence assumptions between the features.

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

using Bayesian probability terminology, the above equation can be written as

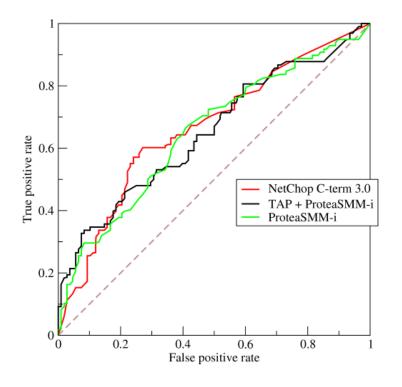


$$F_1 = 2 * \frac{precision * recall}{precision + recall}$$

		Actual				
		Positive	Negative			
cted	Positive	True Positive	False Positive			
Predicted	Negative	False Negative	True Negative			

- 模型評估
  - F1-measure
  - 使用 Scikit-learn 架構來定型和評估模型
  - ROC curve圖形,

運用AUC的方式提供更客觀、公正的評估方法



• 分析與應用



#### 政治立場

- DPP
- KMT
- none

#### 框架報導

- 經濟
- 民生
- 國防
- 政治等

#### 應用分析

- 總結歸納
- 分析報導 媒體

# 預期成果與展望

- 使機器具備預測報導立場框架能力
- 避免讀者被框架報導所影響誤導



- ACCESS × ANALYZE × EVALUATE
- 營造互利共榮的健全媒體生態



#### References

- Kristen Johnson, Di Jin, & Dan Goldwasser (2017). *Leveraging Behavioral and Social Information for Weakly Supervised Collective Classification of Political Discourse on Twitter*. Proceedings of the 55th Annual Meeting of the Association for Computational Linguistics, pages 741–752. https://doi.org/10.18653/v1/P17-1069
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