

## Introduction to Information Extraction (IE) and ANNIE

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## **About this tutorial**

This tutorial comprises the following topics:

- Introduction to IE
- ANNIE
- Evaluation and Corpus Quality Assurance

Later, you'll learn how to use JAPE, the pattern matching language that is used in ANNIE and useful for many IE tasks.

## **Information Extraction**



- Find relevant parts of texts: e.g. mentions of persons, dates
- Input: unstructured text
- Output: structured information (XML file, Database table, RDF)
- IE is often hard:
  - Has to deal with the ambiguity of language
  - Requires knowledge about the language
  - Requires knowledge about the world
  - Information could be e.g.:
    - Persons, Locations, .. mentioned: Named Entity Recognition
    - Relations between entities
    - Events



## **Named Entity Recognition**

- Named Entity
  - some specific individual named thing in the world
  - often extends to other pieces of information like dates
- Different types of NEs: Person, Organization (companies, government organizations, committees, universities, etc), Location (cities, countries, rivers, etc. bridge? Mall?)
- Various other types are frequently added, depending on the task, e.g. newspapers, ships, species, monetary amounts, percentages.
- Need to find and classify the mentions

## **NER Importance**



- NER provides a foundation from which to build more complex IE systems.
   Once we have found NEs we can:
- Find co-referring mentions: "Dr Smith", "John Smith", "John", "he"
- Find relations: "He (Dr Smith) is CEO of XY"
- Link to Ontologies/Knowledge Bases: "Athens, Georgia" vs "Athens, Greece" (*Named Entity Disambiguation*)

# GATE

## **Typical NER pipeline**

- Pre-processing
  - Tokenisation: characters  $\rightarrow$  tokens/words
  - Sentence splitting
  - POS tagging: nouns, verbs, adjectives, ...
  - Morphological analysis: word root/lemma
- Find entity mentions: Persons, Locations, ...
- Type disambiguation (Person or Location or ..?)
- Coreferencing: (different) mentions of the same entity





#### John lives in London . He works there for Polar Bear Design .



## Preprocessing



#### **Basic NE Recognition**











## Relations





## **Relations (2)**







## [Semantic Annotation / Linking]





# ANNIE <u>A Nearly New Information Extraction System</u>

## **About this tutorial**



- As before, this tutorial will be a hands on session with some explanation as you go.
- We will use a corpus of news texts in your hands-on directory annie-hands-on/news-texts
- Things for you to try yourself are in **red**.
- Start GATE on your computer now (if you haven't already)

#### A Nearly New Information Extraction [System]



- ANNIE is a ready made collection of processing resources (PRs) that performs (parts of) IE on text.
- ANNIE was "nearly new" because
  - It was based on an existing IE system, LaSIE
  - Being >10 years old now, it's not really new any more
- ANNIE performs competitively "out of the box"
- ANNIE can be used as the starting point for your own solutions

## **ANNIE Components**

GATE

- The ANNIE application contains a set of core PRs:
  - Tokeniser
  - Sentence Splitter
  - POS tagger
  - Gazetteers
  - Named entity tagger (JAPE transducer)
  - Orthomatcher (orthographic NE coreference)
- Other PRs available from the plugin (not in the application):
  - Pronominal Coreferencer
  - Additional PR s in the Tools plugin, e.g. Morphological Analyser, VP Chunker

# GATE

## Loading and running ANNIE

- Because ANNIE is a ready-made application, we can just load it directly from the menu
- Click the icon from the top GATE menu OR File →Ready Made Applications →ANNIE →ANNIE OR right-click Applications →Ready Made Applications →ANNIE →ANNIE
- Select "with defaults" if necessary
- Load the hands-on corpus from the "news-texts" directory
- Run ANNIE and inspect the annotations
- You should see a mixture of Named Entity annotations (Person, Location etc) and some other linguistic annotations (Token, Sentence etc)



## **ANNIE Processing Resources**

- View the ANNIE application
   (Double Click the Icon near the name)
- Each PR in the ANNIE pipeline creates some new annotations, or modifies existing ones
- Document Reset: removes any existing annotations
- Tokeniser: create Token, SpaceToken annotations
- Gazetteer: create Lookup annotations
- Sentence Splitter: create Sentence, Split annotations
- POS tagger: adds category feature to Token annotations
- NE transducer: create Date, Person, Location, Organisation, Money, Percent annotations
- Orthomatcher: adds match features to NE annotations

# GATE

## **Document Reset**

- This PR should go at the beginning of (almost) every application you create
- It removes annotations created previously, to prevent duplication if you run an application more than once
- It does not remove the "Original markups" set, by default
- It keeps the "Key" set, by default
- You can configure it to keep any other annotation sets you want, or to remove particular annotation types only
- In more complex applications it is a good idea to:
  - At the start, delete only the annotations created
  - At the end, delete any temporary annotations



#### **Document Reset Parameters**





## Tokenisation and Sentence Splitting

## Tokeniser



- Tokenisation based on Unicode character classes
- Uses declarative token specification language
- Produces Token and SpaceToken annotations with features orthography and kind
- Length and string features are also produced
- Rule for a lowercase word with initial uppercase letter:

"UPPERCASE\_LETTER" LOWERCASE\_LETTER"\* >
 Token; orthography=upperInitial; kind=word

# GATE

### **Document with Tokens**

Annotation Sets Annotations List Annotations Stack Class Co-reference Edite	٥r	Ins	stance Text 🔍 💌
Union Appeals For Talks To End BA Strike			Data
	=		Date
Skip to navigation . Skip to content . Heme I Contact Us I News Search			FirstPerson
HubPage			JobTitle
Airwise News			Location
Airport Guide			Lookup
Airwise Travel			Monoy
Search Upiep Appeals Far Talks To Fod PA Strike			noney
March 22, 2010			Organization
March 22, 2010			Percent
Union leaders on Sunday called for talks with British Airways bosses to end strike action by			Person
cabin crew that has led to the cancellation of hundreds of flights and disrupted travel plans for			Sentence
thousands of passengers,	•		SpaceToken
			Colit
Type Features			Spiit
Token {category=NNP, kind=word, length=5, orth=upperInitial, string=Union}	*		Title
Token {category=NNPS, kind=word, length=7, orth=upperInitial, string=Appeals}	=	1	Token
Token {category=IN, kind=word, length=3, orth=upperInitial, string=For}			Unknown
Token {category=NNS, kind=word, length=5, orth=upperInitial, string=Talks}			UNKIOWI
Token {category=TO, kind=word, length=2, orth=upperInitial, string=To}	•		Original markups



## **ANNIE English Tokeniser**

- The English Tokeniser is a slightly enhanced version of the Unicode tokeniser
- It wraps the Unicode Tokeniser PR and an additional JAPE transducer which adapts the generic tokeniser output for the POS tagger requirements
- It converts constructs involving apostrophes into more sensible combinations
  - don't  $\rightarrow$  do + n't
  - you've  $\rightarrow$  you + 've

## GATE

## **Re-creating ANNIE**

- > Tidy up GATE by removing all resources and applications
- Load the news text hands-on corpus
- Create a new application (Conditional Corpus Pipeline)
- Load a Document Reset and an ANNIE English Tokeniser
- > Add them (in that order) to the application and run on the corpus
- > View the Token and SpaceToken annotations
- > What different values of the "kind" feature do you see?

# GATE

## **Sentence Splitter**

- The default splitter finds sentences based on Tokens
- Creates Sentence annotations and Split annotations on the sentence delimiters
- Uses a gazetteer of abbreviations etc. and a set of JAPE grammars which find sentence delimiters and then annotate sentences and splits
- Load an ANNIE Sentence Splitter PR and add it to your application (at the end)
- Run the application and view the results

## GAT

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### **Document with Sentences**

Apportation Sets Apportations List Apportations Stack Class Co-reference Edit	or Inst	ance Text 🔍 💌
The opposition conservatives, aneau in opinion poils, have been turning up the pressure on		
Labour over its links to Unite, saying the government had failed to take action quickly enough because it did not want to alienate its financial backers		Date
		FirstPerson
"We deplore the strike, and the prime minister and the transport secretary have said that		obTitle
absolutely clearly," Foreign Secretary David Miliband told Sky News.	🕴 🗖 🛽	ocation
"The way to resolve these disputes is through negotiation, it is damaging for the company, it is		ookup
damaging for the crews and it is damaging for the country."		Money
The dispute arose because BA, which has 12,000 cabin crew, wants to save an annual		Organization
GBP£62.5 million pounds (USD\$95 million) to help cope with a fall in demand, volatile fuel		Percent
prices and increased competition from low-cost carriers.		Person
A spokesman said there was no estimate yet as to how much the industrial action would cost		Sentence
the company.	I 🛛 🔤 🖬	SpaceToken
Type Features		Split
Sentence {}	🔺 🛛 🗖	<b>Fitle</b>
Sentence {}		ſoken
Sentence {}		Jnknown
Sentence {}		Original markups



## **Sentence splitter variants**

- By default a new-line character always ends a sentence (better suited when importing HTML or PDF)
- Sometimes plain text files have sentences spanning new line characters. In this case it is better to only end a sentence when two consecutive new-line characters are found (paragraph end).
- To do this, create the sentence splitter using "main.jape" instead of "main-single-nl.jape" as the value of the grammar parameter
- A regular expression Java-based splitter is also available, called RegEx Sentence Splitter, which is sometimes faster
- This handles new lines in the same way as the default sentence splitter



## Shallow lexico-syntactic features



## **POS tagger**

- Adds category feature to Token annotations NNP, VBP, DT, ...
   → Penn Treebank tagset https://www.ling.upenn.edu/courses/Fall\_2003/ling001/penn\_treebank\_pos.html
- ANNIE POS tagger is a Java implementation of Brill's transformation based tagger
- Previously known as **Hepple Tagger** ("heptag")
- Trained on WSJ
- Default ruleset and lexicon can be modified manually (with a little deciphering)
- Requires Tokeniser and Sentence Splitter to be run first

## Morphological analyser



- Not an integral part of ANNIE, but can be found in the *Tools* plugin
- Rule-based: can be modified by the user (instructions in the User Guide)
- Generates "root" feature (lemma) on Token annotations "met" → "meet", "pages" → "page"
- Requires Tokeniser to be run first
- Requires POS tagger to be run first if the considerPOSTag parameter is set to true

## **Find POS tags and roots**



- Add an ANNIE POS Tagger to your application
- Add a GATE Morphological Analyser after the POS Tagger
- If this PR is not available, load the Tools plugin first
- Re-run your application
- Examine the features of the Token annotations
- New features "category" and "root" have been added



## Finding Known Names: Gazetteers

## Gazetteers



- Gazetteers are plain text files containing lists of names (e.g rivers, cities, people, ...)
- Idea is to quickly find all mentions in a text that match a name
- List is usually compiled into a compact representation for fast matching (e.g. Finite State Machines)
- Each gazetteer has a *definition file* listing all the lists, plus features of each list (majorType, minorType and language)
- Lists can be modified either internally using the Gazetteer Editor, or externally in your favourite editor
- Gazetteers generate (by default) Lookup annotations with relevant features corresponding to the list matched
- Lookup annotations are used primarily by the NE transducer



## **Running the ANNIE Gazetteer**

- Various different kinds of gazetteer are available: first we'll look at the default ANNIE gazetteer
- IMPORTANT: there is a copy of the original GATE gazetteer files available in annie-hands-on/gazetteer we use that to avoid modifying the originals!
- Add the ANNIE Gazetteer PR to the end of your pipeline
   !! Change the listsURL parameter to the file annie-hands-on/gazetteer/lists.def
- Re-run the pipeline
- Look for "Lookup" annotations and examine their features


#### **ANNIE gazetteer - contents**

- Double click on the ANNIE Gazetteer PR (under Processing Resources in the left hand pane) to open it
- Select "Gazetteer Editor" from the bottom tab
- In the left hand pane ("List name") you see the definition file containing all the lists
- In the right hand pane you see the contents of the list selected in the left hand pane
- Each entry can be edited by clicking in the box and typing
- Add new list: enter list name in the left pane, click "Add"
- Add new entry: enter entry in the right pane, click "Add"

#### **Gazetteer editor**



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File Options Tools Help							
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G GATE	Messages 🏾 🌺 Corp	us Pipeline	🕼 in-whitbre	ad.	10 🖓 ANNIE Gazetteer		
• 🗱 Applications	airport.lst 👻	Add			Filter Add	+Cols 1989 entries 🗹 Case Ins.	
🗠 🕎 Language Resources	List name	Major	Minor		Value		
- 🏕 Processing Resources	abbreviations.lst	stop		•	Аассга		
	adbc.lst	adbc			Aalborg		
– ≪ ANNIE Gazetteer_0007≡	airports.lst	location	airport	=	Aarhus		
GATE Morphological a	charities.lst	organization			Ababa		
	city.lst	location	city		Abadan		
— 🥦 ANNIE POS Tagger_00	city_cap.lst	location	city		Abakan		
ANNIE Septence Splitt	company.lst	organization	company		Aberdeen		
ANNIE Sentence Spirt	company_cap.lst organization		company		Abha		
🚽 🗕 🔫 Document Reset PR_0	country.lst	location	country		Abi Dhabi		
	country_abbrev.lst	location	country_abbr		Abidjan		
ANNIE English Tokenis	country_adj.lst	country_adj			Abilene		
	country_cap.lst	location	country		Abu		
	currency_prefix.lst	currency_unit	pre_amount		Abu Dhabi		
	cu rency_unit.lst	currency_unit	post_amount		Abuja		
	date_key.lst	date_key			Acapulco		
	ate_unit.lst	date_unit			Acarigua		
	day.lst	date	dav	-	Accra		
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	Gazetteer Editor	itialisation Para	ameters				

lists in the definition file

#### entries for selected list

#### **Modifying the definition file**



add a new list edit an edit the major and minor Types by existing list airport.lst Add Ŧ typing here name by List name Major Minor abbreviations.lst stop ۰ typing here adbc.lst adbc location airport airports.lst charities.lst organization location 🔶 city.lst city city cap.lst location city company company.lst organization delete a list company cap.lst organization company country.lst location country by right country abbrev.lst location country abbr country\_adj.lst country adj clicking on country cap.lst location country an entry and currency\_prefix.lst pre amount currency unit currency\_unit.lst currency\_unit post amount selecting date key.lst date\_key date unit.lst Delete date unit date Ŧ day.lst dav Ш

#### **Modifying a list**







#### **Gazetteer lists**

- The ANNIE gazetteer has about 80,000 entries arranged in >100 lists
- Each list is for a category/type, e.g. airports, cities, first names etc.
- List entries might be names or parts of names, or they may contain contextual information (e.g. job titles often indicate people)
- Click on any list to see the entries
- Note that some lists are not very complete!



#### **Editing gazetteer lists**

- Try adding, deleting and editing existing lists, or the list definition file
- To save an edited gazetteer, right click on the gazetteer name in the tabs at the top or in the resources pane on the right, and select "Save and Reinitialise" before running the gazetteer again.
- "Save as" saves the list under the given name plus all lists in the same directory
- Try adding a word from a document you have loaded (that is not currently recognised as a Lookup) into the gazetteer, re-run the gazetteer and check the results.

## Editing gazetteers outside GATE



- You can also edit both the definition file and the lists outside GATE, in your favourite text editor
- If you choose this option, you will need to reinitialise the gazetteer in GATE before running it again
- To reinitialise any PR, right click on its name in the Resources pane and select "Reinitialise"

# GATE

#### **Per-List Features**

- When something in the text matches a gazetteer entry, a Lookup annotation is created, with various features
- The ANNIE gazetteer has the following default feature names: majorType, minorType, language
- These features are used to organize the lists In the definition file features are separated by ":"
- For example, the "city" list has a majorType "location" and minorType "city", while the "country" list has "location" and "country" as its types
- Later, in the JAPE grammars, we can refer to all Lookups of type location, or we can be more specific and refer just to those of type "city" or type "country"



#### **Per-Entry Features**

- Each entry in a list file can have arbitrary features
- For each row must contain the entry to match, then for each feature: a feature separator character, feature name, equals character and feature value, e.g. with the feature separator character ":"

Paris:country=France:timezone=CET:lang=fr In many cases, a tab character is the best choice of feature separator character.

• For each match, the per-entry features are added to the Lookup annotation in addition to the per-list features



## **NE Transducers**

#### **NE Transducer**

GATE

- Gazetteers can be used to find terms that suggest entities
- However, the entries can often be ambiguous
  - "May Jones" vs "May 2010" vs "May I be excused?"
  - "Mr Parkinson" vs "Parkinson's Disease"
  - "General Motors" vs. "General Smith"
- Hand-crafted grammars are used to define patterns over the Lookups and other annotations
- These patterns can help disambiguate, and they can combine different annotations, e.g. Dates as day name + number + month name
- NE transducer consists of a number of grammars written in the JAPE language
- This afternoon will be devoted to JAPE

#### "Transducer"?



- We start with patterns which we want to match in text The patterns are based on annotations and their features
- Patterns are associated with one or more actions
  e.g. create a new annotation for some part of the pattern,
  add a feature, remove some overlapping annotation
- Patterns and actions are are expressed in the "JAPE" language <Pattern> → <Action> Rule left hand side: pattern to match Rule right hand side: Action to carry out
- The JAPE language gets compiled into a FSM for fast matching and carrying out the actions

# GATE

#### **ANNIE NE Transducer**

- Load an ANNIE NE Transducer PR
- Add it to the end of the application
- Run the application
- Look at the annotations
- You should see some new annotations such as Person, Location, Date etc.
- These will have features showing more specific information (eg what kind of location it is) and the rules that were fired (for ease of debugging)



## **Co-reference**

#### **Using co-reference**



- Different expressions may refer to the same entity
- Orthographic co-reference module (orthomatcher) matches proper names and their variants in a document
- [Mr Smith] and [John Smith] will be matched as the same person
- [International Business Machines Ltd.] will match [IBM]

# GATE

#### **Orthomatcher PR**

- Performs co-reference resolution based on orthographical information of entities
- Produces a list of annotation IDs that form a co-reference "chain"
- List of such lists stored as a *document feature* named "MatchesAnnots"
- Improves results by assigning entity type to previously unclassified names, based on relations with classified entities
- May not reclassify already classified entities
- Classification of unknown entities very useful for surnames which match a full name, or abbreviations, e.g. "Bonfield" <Unknown> will match "Sir Peter Bonfield" <Person>
- A pronominal PR is also available

## Looking at co-reference



- > Add a new PR: ANNIE OrthoMatcher
- Add it to the end of the application
- Run the application
- Look at the features of NE annotations
- Look at the document features in the bottom left pane
- In a document view, open the co-reference editor by clicking the button above the text
- All the documents in the corpus should have some coreference, but some may have more than others



#### **Coreference editor**



Document Editor Initialisation Parameters



### Using the co-reference editor

- Select the annotation set you wish to view (Default)
- A list of all the co-reference chains that are based on annotations in the currently selected set is displayed
- Select an item in the list to highlight all the member annotations of that chain in the text (you can select more than one at one)
- Hovering over a highlighted annotation in the text enables you to Delete an item from the co-reference chain



### Using the co-reference editor

- Deselect all items in the coreference list (right hand pane), then select a type from the "Type" combo box (e.g. "Person") and click "Show" to view all coreferences of a particular annotation type (note that some types may not have coreferences)
- Hovering over a highlighted annotation in the text enables you to add a coreference between this annotation and one of the co-reference chains listed in the right hand pane
- Try it!

#### **ANNIE Alternatives**



 ANNIE is a *rule-based* system: manual rules based on linguistic features (e.g. POS tags) and features from gazetteer lists (e.g. minorType=city) Advantage: an expert can fine-tune, update, extend incrementally

Disadvantage: a lot of work, "diminishing results"

- Other approaches uses machine learning GATE provides:
  - Stanford NER
  - LingPipe NER
  - OpenNLP NER

Advantage: no rule crafting necessary

Disadvantage: need pre-annotated corpus, ML-expert

### **Evaluation**





"We didn't underperform. You overexpected."

### **Evaluation exercises: Preparation**



- Restart GATE, or close all documents and PRs to tidy up
- Load the hands on corpus
- Take a look at the annotations.
- There is a set called "Key". This is a set of annotations against wish we want to evaluate ANNIE. In practice, they could be manual annotations, or annotations from another application.
- Load the ANNIE system with defaults
- Run ANNIE: You should have annotations in the Default set from ANNIE, and in the Key set, against which we can compare them.

#### AnnotationDiff



- Graphical comparison of 2 sets of annotations
- Similar to other visual diff tools (kdiff3, tkdiff)
- Compares one document at a time, one annotation type at a time
- Calculates evaluation measures: Precision, Recall, F-Measure

## **Annotation Diff Exercise**

GATE

- Open the document "ft-airlines-27-jul-2001.xml"
- Open the AnnotationDiff
  (Tools  $\rightarrow$  Annotation Diff or click the icon)
- For the Key set (containing the manual annotations) select Key annotation set
- For the Response set (containing annotations from ANNIE) select Default annotation set
- Select the Organization annotation
- Click on "Compare"
- Scroll down the list, to see correct, partially correct, missing and spurious annotations

#### **Annotation Diff**



😣 🗖 🗊 Annotation Diff Tool																	
Key doc:	ft-airl	ines-27	-jul-200	-	Key set:	Key		-	Type:	Orga	anizatio	n 🔻	w	eight			
Resp. doc:	ft-airl	ines-27	-jul-200	-	Resp. set:	[Defau	lt set]	-	Features:	all	⊖some	none	1.	0	1	Compare	
Start End			Key						Featur	es			=7	Start	End		
1932 1936	Nats					{}							=	1932	1936	Nats	
2456 2460	Nats					{}							=	2456	2460	Nats	
2070 2075	LATCC					{}							=	2070	2075	LATCC	
1354 1362	Barclay	/s				{}							=	1354	1362	Barclays	
1784 1788	Nats					{}							=	1784	1788	Nats	1
1751 1768	The∙Air	rline•Gr	oup			{}							~	1755	1768	Airline · Gro	<u>,</u>
938 955	The•Air	rline•Gr	oup			{}							~	942	955	Airline•Gro	
1669 1686	the•Air	line Gro	oup			{}							~	1673	1686	Airline•Gro	
2412 2429	The• <b>Ai</b> r	rline•Gr	oup			{}							~	2416	2429	Airline Gro	<u>,</u>
1266 1283	The• <b>Ai</b> r	rline•Gr	oup			{}							~	1270	1283	Airline Gro	<u>,</u>
1052 1068	Monard	ch•Airlin	nes			{}							~	1030	1068	Britannia	A
2029 2068	Londor	n•Area•a	and·Termi	nal	Control·Cei	ntre {}							~	2045	2068	Terminal·O	
634 640	Labour	r				{}							-?				
1030 1047	Britann	nia•Airw	ays			{}							-?				
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•																	17
				_		_		10 d	ocuments	loade	ed						
Correct:		19		Rec	all Precisio	on F-me	asure			_							
Partially co	orrect:	7	Strict:	0.68	8 0.68	0.68		G	> 🖄 🍙	)							
Missing:		2	Lenient:	<b>0.9</b> 3	3 0.93	0.93											
False posi	tives:	2	Average:	0.80	0.80	0.80											

Statistics Adjudication

## Kinds of Evaluation and Terminology



- Different communities use different terms when talking about evaluation, because the tasks are a bit different.
- The IE community usually talks about "correct", "spurious" and "missing"
- The IR community usually talks about "true positives", "false positives" and "negatives". They also talk about "false negatives", but you can ignore those.
- Some terminologies assume that one set of annotations is correct ("gold standard")
- Other terminologies do not assume one annotation set is correct
- When measuring inter-annotator agreement, there is no reason to assume one annotator is more correct than the other

## Kinds of Evaluation and Terminology



- In NLP we can easily quantify the locations where something should get annotated (Key, Target annotations) Harder to quantify the locations where we do not want annotations => Tagging Measures: Precision, Recall, F-Measure
- [In other situations we have fixed locations and what an algorithm does for that location is either correct or wrong:
   => Classification] Measure: Accuracy
- In NLP we sometimes compare the tagging output of two algorithms or two human annotators where no output is "the correct" one.

### **Tagging Matches**



- (Strict) Correct = correct type at exact correct position (True Positive, TP)
   e.g. annotating "Hamish Cunningham" as a Person
- Missing = not annoted (False Negative, FN)
  e.g. not annotating "Sheffield" as a Location
- Spurious = wrong type or wrong location (False Positive, FP)
  e.g. annotating "Hamish Cunningham" as a Location
- Partially correct = correct type, location overlap e,g, annotating just "Cunningham" as a Person (too short) or annotating "Unfortunately Hamish Cunningham" as a Person (too long)

## Finding Precision, Recall and F-measure



😸 🖃 🗉 ANNOTATION DIFF TOOL	
Key doc: ft-airlines-27-jul-200 🔻 Key set: Key 👻 Type: Organization 👻 Weight	~
	🔍 Compare
Resp. doc: [ft-airlines-27-jul-200 🔽 Resp. set: [Default set] 🔽 Features: 🔾 all 🔾 some @ none 1.0	74°
Start End Key Features =?Start E	End
1932 1936 Nats {} = 1932 1	.936 Nats 🔺
2456 2460 Nats = 2456 2	2460 Nats
2070 2075 LATCC {} = 2070 2	2075 LATCC
1354 1362 Barclays {} = 1354 1	.362 Barclays
1784 1788 Nats {} = 1784 1	.788 Nats
1751 1768 The Airline Group {} ~ 1755 1	768 Airline · Gro
938 955 The Airline Group {} ~ 942 9	55 Airline-Gro
1669 1686 the Airline Group {} ~ 1673 1	.686 Airline · Gro
2412 2429 The Airline Group {} ~ 2416 2	429 Airline Gro
1266 1283 The Airline Group {} ~ 1270 1	283 Airline · Gro
1052 1068 Monarch Airlines {} ~ 1030 1	.068 Britannia A
2029 2068 London · Area · and · Terminal · Control · Centre {} ~ 2045 2	2068 Terminal C
634 640 Labour {}	
1030 1047 Britannia Airways {} -?	
	040 London Are
· · · · · · · · · · · · · · · · · · ·	395 Hampshire 👻
10 documents loaded	
Correct: 19 Recall Precision F-measure	
Partially correct: 7 Strict: 0.68 0.68 0.68 🔊 🗟 🍙	
Missing: 2 Lenient: 0.93 0.93 0.93 Scores displaye	ed
False positives: 2 Average: 0.80 0.80 0.80	

#### **NLP Precision**



- Fraction of annotations we found that were correct
- Ideally all would be correct, so no spurious!.

All the found annotations, the "response" annotations





- The fraction of the entities that were annotated
- Ideally, all would be correct, so no missing!



All the real entities, the "key" annotations or "target" annotations





- Precision and recall tend to trade off against one another
  - Limiting output to only very specific, high confidence annotations will create fewer annotations and thus is likely to reduce recall
  - Creating more generic annotations in an attempt to improve recall is likely to create more spurious ones and reduce precision
- F-measure combines precision and recall into one measure
- Since both precision and recall are fractions the F-measure is the harmonic mean of precision and recall





$$F=2 \cdot \left( \frac{precision \cdot recall}{precision + recall} \right)$$

- Sometimes precision or recall is given more weight but usually, precision and recall are equally weighted
- This is also known as F1 or F1.0

#### F-Measure Why Harmonic Mean?





### Annotation Diff Defaults to F1



😣 🖻 🗈 Annotation Diff Tool										
Key doc:	ft-airlines-27-jul-200	▼ Key set:	Key	Type:	Organization	- W	oight			
Resp. doc:	ft-airlines-27-jul-200	▼ Resp. set:	[Default set]	▼ Features:	⊖all ⊖some ⊛n	one 1.	0	🦓 Compare		
Start End	Key			Featur	es	=7	Start?	Enc		
1932 1936	Nats		{}			=	1932	1935 Nats		
2456 2460	Nats		{}			=	2456	2460 Nats		
2070 2075	LATCC		{}			=	2070	2075 LATCC		
1354 1362	Barclays		{}			=	1354	1362 Barclays		
1784 1788	Nats		{}			=	1784	1788 Nats		
1751 1768	The•Airline•Group		{}			~	1755	1768 Airline Gro		
938 955	The•Airline•Group		{}			~	942	955 Airline Gro		
1669 1686	the•Airline•Group		{}			~	1673	1686 Airline · Gro		
2412 2429	The•Airline•Group		{}			~	2416	2429 Airline Gro		
1266 1283	The•Airline•Group		{}			~	1270	1283 Airline · Gro		
1052 1068	Monarch·Airlines		{}			~	1030	1068 Britannia A		
2029 2068	London·Area·and·Term	inal·Control·Cer	ntre {}			~	2045	2068 Terminal·C =		
634 640	Labour		{}			-?				
1030 1047	Britannia·Airways		{}			-?				
						?-	2029	2040 London Are		
						?-	2386	2395 Hampshire 👻		
•								•		
Correct:	19	Recall Precisio	on F-measure	10 documents	loaded					
Partially o	orrect: 7 Strict:	0.68 0.68	0.68	la 🛓 🕕	F-mea	isure	e we	ight set to 1		
Missing:	2 Lenient:	0.93 0.93	0.93		-					
False posi	tives: 2 Average:	0.80 0.80	0.80							

Statistics Adjudication


## **Count the partial ones?**

- How we want to deal with partially correct annotations may differ, depending on our goal
- In GATE, there are 3 different ways to measure them
- Strict measures: Only perfectly matching annotations are counted as correct
- Lenient: Partially matching annotations are counted as correct. This usually makes the measures look better
- Average: strict and lenient measures are averaged (this is the roughly the same as counting a half weight for every partially correct annotation)



### Strict, Lenient, Average

	Annotation D	iff Tool												
Key doc:	ft-airlines-2	7-jul-200	🔻 Key	set:	Key	-	Туре:	Organizatio	n 🔻	We	eight			
Resp. doc:	ft-airlines-2	7-jul-200	Res	p. set:	[Default set]	-	Features:	⊖all ⊖some	Image:	1.0	0	0	Compar	e
Start End	L	Kev				Eeatures =7Start End								
1932 1936	Nats	i key			{}		reatur	0.5		=	1932	1936	Nats	
2456 2460	Nats				{}					=	2456	2460	Nats	
2070 2075	LATCC				{}					=	2070	2075	LATCC	
1354 1362 Barclays			{}					=	1354	1362	Barclays			
1784 1788 Nats			{}					=	1784	1788	Nats			
1751 1768	The Airline G	roup			{}					~	1755	1768	Airline•Gro	01
938 955	The Airline G	roup			{}					~	942	955	Airline•Gr	
1669 1686 the Airline Group			{}					~	1673	1686	Airline•Gro			
2412 2429 The Airline Group			{}					~	2416	2429	Airline•Gro	01		
1266 1283 The Airline Group			{}					~	1270	1283	Airline•Gro	01		
1052 1068 Monarch Airlines			{}					~	1030	1068	Britannia	A		
2029 2068 London Area and Terminal Control Centre				tre {}					~	2045	2068	Terminal	ci=	
634 640 Labour			{}					-?						
1030 1047 Britannia Airways			{}					-?						
										?-	2029	2040	London·A	re
										?-	2386	2395	Hampshir	e 🗸
•										· · · ·				•
6	10		D	D		10 o	locuments	loaded						
Correct:	19		Recall	Precisio	n F-measure									
Partially c	orrect: 7	Strict:	0.68	0.68	0.68	G	> 🖄 🚺							
Missing:	2	Lenient:	0.93	0.93	0.93									
False posi	itives: 2	Average:	0.80	0.80	0.80									
Statistic	Statistics Adjudication													

## **Comparing Individual Annotations**



- In the AnnotationDiff window, colour codes indicate whether the annotation pair shown are correct, partially correct, missing (false negative) or spurious (false positive)
- You can sort the columns however you like



#### **Comparing the annotations**

A 888	Annotation Diff Tool			
Key doc:	ft-airlines-27-jul-200 🔻 Key set:	Key	▼ Type: Organization	▼ Weight
Resp. doc:	ft-airlines-27-jul-200 🔻 Resp. set:	[Default set]	▼ Features: ○all ○some	●none 1.0
Start End	Key		Features	=?Start End
1932 1936	Nats	{}		= 1932 1936 Nats
2456 2460	vats	13		= 2456 2460 Nats
2070 2075		{}		= 20/0 20/5 LATCC
1354 1362	Barclays	{}		= 1354 1362 Barclays
1784 1788	Nats	{}		= 1784 1788 Nats
1751 1768	The Airline Group	{}		~ 1755 1768 Airline Gro
938 955	Fhe•Airline•Group	{}		~ 942 955 Airline Gro
1669 1686 t	the•Airline•Group	{}		~ 1673 1686 Airline · Grou
2412 2429	The•Airline•Group	{}	<b>&gt;</b>	~ 2416 2429 Airline Gro
1266 1283	The•Airline•Group	{}		~ 1270 1283 Airline Gro
1052 1068	Monarch-Airlines	1		~ 1030 1068 Britannia 4
2029 2068	London·Area·and·Terminal·Control·Cer	ntre {}		~ 2045 2068 Terminal·C ≡
634 640 I	Labour	{}		-?
1030 1047 I	Britannia·Airways	{}		-?
				?- 2029 2040 London Are
				?- 2386 2395 Hampshire 👻
•				Þ
Correct:	19 Recall Precisio	on F-measure	10 documents loaded	
Partially co	orrect: 7 Strict: 0.68 0.68	0.68	🧟 🚊 🕕	
Missing:	2 Lenient: 0.93 0.93	0.93		$\backslash$
False posi	tives: 2 Average: 0.80 0.80	0.80	Respon	se annotations
Statistics	Adjudication Ney annota	แบกร		



## **Corpus Quality Assurance**

- The Corpus Quality Assurance tool extends the Annotation Diff functionality to an entire corpus
- It produces statistics both for the corpus as a whole (Corpus statistics tab) and for each document separately (Document statistics tab)
- It evaluates several types at once (e.g. Person, Location, Organization)
- It creates **micro averages** and **macro averages** over documents and over types (see next slide)



# **Micro and Macro Averaging**

- Micro averaging treats the entire corpus as one big document, for the purposes of calculating precision, recall and F
- Macro averaging takes the average over each of precision, recall and F over all documents or over all types: generally less useful / not meaningful
- Over documents, micro average is more useful
- Over types:
  - micro gives the overall picture, more frequent types influence the result more
  - [macro gives ~equal weight to each type]

# Try Out Corpus Quality Assurance



Open your hands-on corpus and click the Corpus Quality Assurance tab at the bottom of the Display pane.





## **Select Annotation Sets**

Corpus statistics	Document statistics	Oply B	Overlap		Select the
	Maten	Unity B	ovenap	Annotation Sets A/Key & B/Respons	annotation sets
				Key Original markups	you wish to
				present in every document	compare.
				Annotation Types	
					<ul> <li>Click on the Key</li> </ul>
					annotation set –
				Annotation Features	this will label it
					set A.
				present in every selected type	<ul> <li>Now click on the</li> </ul>
				F-Score Classification	default
				F1.0-score strict	annotation set -
				F1.0-score average F1.0-score strict BDM	this will label it
				🐁 Compare	
		 			Sel D.



## **Select Type**



- Select the annotation type to
   compare (suggestion: select Organisation, Person and Location)
  - Select the features to include (if any – leave unselected for now)
- You can select as many types and features as you want.



#### **Select measure**



 In the "Measures" box, select the kind of F score you want "Strict, Lenient, Average" or any combination of them. Suggestion: select all three

Select Compare



#### **Corpus Statistics Tab**

			b o cam	ene su	ausuts			
	Annotation	Match	Only A	Only I	BOverlap	Rec.B/A	Prec.B/A	F1-strict
Pe	rson	32 <b>8</b>	26	11	7	0.91	0.95	0.93
Ma	lcro summary					0.91	0.95	0.93
Mi	cro summary	32 <b>8</b>	26	11	7	0.91	0.95	0.93

- Each annotation type is listed separately
- Precision, recall and F measure are given for each
- Two summary rows provide micro and macro averages



## **Document Statistics Tab**

Corpus statistics Document statistics								
Document	Match	Only A	Only B	Overlap	Rec.B/A	Prec.B/A	F1-strict	
in-reed-10-aug-2001.xml_00072	10	1	0	0	0.91	1.00	0.95	
in-rover-10-aug-2001.xml_00073	3	0	0	0	1.00	1.00	1.00	
in-scoot-10-aug-2001.xml_00074	1	0	0	0	1.00	1.00	1.00	
in-shell-cirywire-03-aug-2001.xml_00075	7	1	0	0	0.88	1.00	0.93	
in-tesco-citywire-07-aug-2001.xml_00076	1	0	0	0	1.00	1.00	1.00	
in-whitbread-10-aug-2001.xml_00077	1	0	0	0	1.00	1.00	1.00	
Macro summary					0.95	0.95	0.94	
Micro summary	328	26	11	7	0.91	0.95	0.93	-
Corpus editor Initialisation Parameters Corpus Quality Assurance								

- Each document is listed separately
- Precision, recall and F measure are given for each
- Two summary rows provide micro and macro averages

### **Classification Measures**



- present	in every selected type
Measures	
F-Score	Classification
Observed a Cohen's Ka Pi's Kappa	greement ppa
4	Compare

- By default, Corpus Quality Assurance presents the Fmeasures
- However, classification measures are also available
- These are not suitable for entity extraction (tagging) tasks





- This session has been devoted to IE and ANNIE
- You should now have a basic understanding of:
  - what IE is
  - how to load and run ANNIE, what each of the ANNIE components do, how to modify ANNIE components
  - Evaluation using Annotation Diff and Corpus QA



#### **Extra exercises**

If you have some spare time, you can try:

- Load the application
   annie-hands-on/apps/lingpipe-ner.gapp

   Look at the PRs it contains
   Run it on the evaluation corpus, evaluate
- Load the application
   annie-hands-on/apps/stanford-ner.gapp
   Look at the PR s it contains
   Run it on the evaluation corpus, evaluate