The COCA corpus (new version released March 2020)

The corpora from English-Corpora.org are the world's **most widely-used corpora**. The Corpus of Contemporary American English (COCA) is by far the most widely-used of these corpora. In early 2020, we dramatically expanded the scope and size and features of COCA to make it even more useful for researchers, teachers, and learners.

The corpus contains more than **one billion words** of data, including 20 million words each year from **1990-2019** (with the same genre balance year by year). This makes COCA the only corpus of English that is 1) large 2) recent and 3) has a wide range of genres. The following table shows the genres in the corpus.

Genre	# texts	# words	Explanation
Spoken	44,803	127,396,932	Transcripts of unscripted conversation from more than 150 different TV and radio programs (examples: All Things Considered (NPR), Newshour (PBS), Good Morning America (ABC), Oprah)
Fiction	25,992	119,505,305	Short stories and plays from literary magazines, children's magazines, popular magazines, first chapters of first edition books 1990-present, and fan fiction.
Magazines	86,292	127,352,030	Nearly 100 different magazines, with a good mix between specific domains like news, health, home and gardening, women, financial, religion, sports, etc.
Newspapers	90,243	122,958,016	Newspapers from across the US, including: USA Today, New York Times, Atlanta Journal Constitution, San Francisco Chronicle, etc. Good mix between different sections of the newspaper, such as local news, opinion, sports, financial, etc.
Academic	26,137	120,988,361	More than 200 different peer-reviewed journals. These cover the full range of academic disciplines, with a good balance among education, social sciences, history, humanities, law, medicine, philosophy/religion, science/technology, and business
Web (Genl)	88,989	129,899,427	Classified into the web genres of academic, argument, fiction, info, instruction, legal, news, personal, promotion, review web pages (by Serge Sharoff). Taken from the US portion of the GloWbE corpus.
Web (Blog)	98,748	125,496,216	Texts that were classified by Google as being blogs. Further classified into the web genres of academic, argument, fiction, info, instruction, legal, news, personal, promotion, review web pages. Taken from the US portion of the GloWbE corpus.
TV/Movies	23,975	129,293,467	Subtitles from OpenSubtitles.org, and later the TV and Movies corpora. Studies have shown that the language from these shows and movies is even more colloquial / core than the data in actual "spoken corpora".
	485,179	1,002,889,754	

GENRES Because COCA has so much data from each of these eight genres, it provides useful information about the frequency of words, phrases, and grammatical constructions across the genres – whether they are very informal (e.g. TV and movie subtitles or in spoken transcripts), more formal (e.g. academic articles), or somewhere in between (e.g. magazines and newspapers). For example, the following two charts show the frequency of the word *lucky* in the different genres, as well as the "get passive" (e.g. he got promoted).

	lucky								get p	assive						
SECTION	BLOG	WEB	TV/M	SPOK	FIC	MAG	NEWS	ACAD	BLOG	WEB	TV/M	SPOK	FIC	MAG	NEWS	ACAD
FREQ	6370	4877	16697	5094	9249	4825	3916	741	34353	26869	50926	33009	21756	19638	17673	3867
WORDS (M)	128.6	124.3	128.1	126.1	118.3	126.1	121.7	119.8	128.6	124.3	128.1	126.1	118.3	126.1	121.7	119.8
PER MIL	49.53	39.25	130.37	40.39	78.17	38.27	32.17	6.19	267.10	216.24	397.63	261.69	183.87	155.74	145.17	32.28
SEE ALL SUB-SECTIONS AT ONCE																

The great general balance (and large size) of COCA also allows you to see the frequency of related phrases across genres. For example, the following table shows "soft NOUN".

	CONTEXT	ALL	BLOG	WEB-GENL	TV/MOVIES	SPOKEN	FICTION	MAGAZINE	NEWSPAPER	ACADEMIC
1	SOFT TISSUE	1120	62	67	74	39	36	100	35	707
2	SOFT DRINKS	1109	123	123	42	90	83	304	296	48
3	SOFT MONEY	790	21	34	12	446	8	78	153	38
4	SOFT SPOT	867	133	110	166	63	159	135	86	15
5	SOFT DRINK	721	43	60	48	77	68	191	199	35
6	SOFT VOICE	546	12	39	10	11	351	53	52	18
7	SOFT POWER	421	49	53	1	64	1	51	32	170
8	SOFT TISSUES	328	13	23	5	6	10	41	6	224
9	SOFT LANDING	274	42	19	18	38	17	56	63	21
10	SOFT LIGHT	247	22	25	9	1	121	38	21	10
11	SOFT PEAKS	216	2	4		1	2	144	62	1
12	SOFT SKIN	207	9	24	28	6	114	19	4	3
13	SOFT TOUCH	213	22	26	28	15	43	51	25	3
14	SOFT SKILLS	219	41	41		7		13	13	104

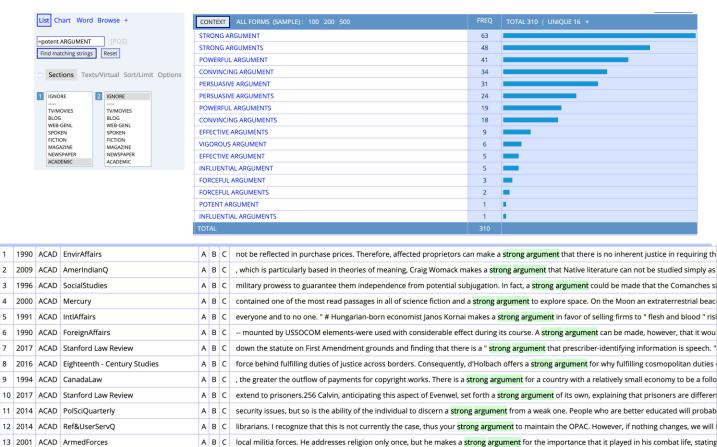
You can also compare two genres (or sets of genres). For example, the following table shows collocates (nearby words) of *chair* in ACADEMIC (left) and FICTION (right).

SEC 1 (ACADEMIC): 119,790,456 W	ORDS					SEC 2 (FICTION): 118,322,084 WC	RDS				
	WORD/PHRASE	TOKENS 1	TOKENS 2	PM 1	PM 2	RATIO		WORD/PHRASE	TOKENS 2	TOKENS 1	PM 2	PM 1	RATIO
1	DIVISION	56	1	0.5	0.0	55.3	1	KITCHEN	293	3	2.5	0.0	98.9
2	ENGINEERING	53	1	0.4	0.0	52.4	2	WING	111	0	0.9	0.0	93.8
3	ABA	55	0	0.5	0.0	45.9	3	WAY	87	1	0.7	0.0	88.1
4	PROGRAM	46	1	0.4	0.0	45.4	4	FEET	220	3	1.9	0.0	74.2
5	EDUCATION	42	0	0.4	0.0	35.1	5	FOOT	72	1	0.6	0.0	72.9
6	RESEARCH	34	1	0.3	0.0	33.6	6	MOTHER	134	2	1.1	0.0	67.8
7	FACULTY	38	0	0.3	0.0	31.7	7	COMMAND	65	1	0.5	0.0	65.8
8	ASME	36	0	0.3	0.0	30.1	8	LEATHER	409	7	3.5	0.1	59.2
9	COMMISSION	30	1	0.3	0.0	29.6	9	CANVAS	58	1	0.5	0.0	58.7
10	ROLE	33	0	0.3	0.0	27.5	10	BEACH	67	0	0.6	0.0	56.6
11	SUBCOMMITTEE	24	1	0.2	0.0	23.7	11	FINGERS	53	1	0.4	0.0	53.7
12	DESIGN	28	0	0.2	0.0	23.4	12	LAWN	318	6	2.7	0.1	53.7
13	MEMBER	45	2	0.4	0.0	22.2	13	VISITOR	51	1	0.4	0.0	51.6
14	RISE	44	2	0.4	0.0	21.7	14	CAPTAIN	61	0	0.5	0.0	51.6
15	DEAN	26	0	0.2	0.0	21.7	15	DAY	50	1	0.4	0.0	50.6

You can even compare synonyms of a given word in different genres. For example, this table shows the synonyms of *strong* in TV/MOVIES (left) and ACADEMIC (right).

SEC 1	(TV/MOVIES): 128,074,534	WORDS					SEC 2	(ACADEMIC): 119,790,456 \	WORDS				
	WORD/PHRASE	TOKENS 1	TOKENS 2	PM 1	PM 2	RATIO		WORD/PHRASE	TOKENS 2	TOKENS 1	PM 2	PM 1	RATIO
1	SPICY	581	53	4.5	0.4	10.3	1	ROBUST	2829	108	23.6	0.8	28.0
2	BEEFY	64	7	0.5	0.1	8.6	2	EFFECTIVE	28807	1596	240.5	12.5	19.3
3	TOUGH	11190	1579	87.4	13.2	6.6	3	DURABLE	683	65	5.7	0.5	11.2
4	STRAPPING	143	21	1.1	0.2	6.4	4	CONCENTRATED	2750	298	23.0	2.3	9.9
5	нот	23639	3877	184.6	32.4	5.7	5	STAUNCH	252	32	2.1	0.2	8.4
6	BRILLIANT	4418	1156	34.5	9.7	3.6	6	ZEALOUS	217	31	1.8	0.2	7.5
7	GREAT	107836	34522	842.0	288.2	2.9	7	ARDENT	375	54	3.1	0.4	7.4
8	BITING	658	230	5.1	1.9	2.7	8	COMPELLING	2845	452	23.7	3.5	6.7
9	BURLY	52	27	0.4	0.2	1.8	9	FERVENT	210	34	1.8	0.3	6.6
10	BRIGHT	4561	2542	35.6	21.2	1.7	10	CLEAR-CUT	405	83	3.4	0.6	5.2
11	SOUND	18257	11719	142.5	97.8	1.5	11	PERSUASIVE	1360	308	11.4	2.4	4.7
12	WELL-BUILT	20	14	0.2	0.1	1.3	12	DEEP-SEATED	260	61	2.2	0.5	4.6

The ability to focus in on specific genres means that you can find "just the right word" for a particular concept in a particular genre. For example, suppose that you want to know what word related to *potent* is the most frequent with a form of *argument* in academic English. As the following image shows, you simply search for *=potent ARGUMENT* and limit the search to academic, and you would see the following results, and then (as with any search results) can see the matching phrases in context,



In addition to comparing across genres, you can also compare words, to "tease out" differences between related words. For example, the following chart shows words that occur with *deep* and *profound*, showing that (for example) *deep breath* is common but *profound breath* never occurs, or that *profound effect* is common but *deep effect* is not. You can also compare other related words, such as adjectives used with the words *men* and *women*, or verbs occurring near *Obama* or *Trump*.

WORD 2 (W2): PROFOLIND (0.14)

WORD 1 (W1): DEEP (7 33)

WORD I	(VVI). DEEP (7.33)					WORD 2	(WZ). PROFOUND (0.14)				
	WORD	W1	W2	W1/W2	SCORE		WORD	W2	W1	W2/W1	SCORE
1	BREATH	7687	0	15,374.0	2,096.7	1	HEARING	54	0	108.0	791.9
2	SPACE	1292	0	2,584.0	352.4	2	DISABILITIES	25	0	50.0	366.6
3	BREATHS	1285	0	2,570.0	350.5	3	AFFECT	23	0	46.0	337.3
4	POCKETS	764	0	1,528.0	208.4	4	EFFECT	695	16	43.4	318.5
5	WATER	1224	1	1,224.0	166.9	5	EFFECTS	214	7	30.6	224.2
6	THROAT	513	0	1,026.0	139.9	6	DISABILITY	14	0	28.0	205.3
7	OCEAN	414	0	828.0	112.9	7	REVELATION	14	0	28.0	205.3
8	VOICE	799	1	799.0	109.0	8	LESSON	27	1	27.0	198.0
9	SNOW	360	0	720.0	98.2	9	MANNER	11	0	22.0	161.3
10	RED	337	0	674.0	91.9	10	RETARDATION	11	0	22.0	161.3
11	SEA	671	1	671.0	91.5	11	SHIFT	100	5	20.0	146.7
12	FREEZE	327	0	654.0	89.2	12	GIFT	10	0	20.0	146.7
13	EYES	318	0	636.0	86.7	13	IMPAIRMENT	10	0	20.0	146.7
14	ROOTS	615	1	615.0	83.9	14	STATEMENT	57	3	19.0	139.3

HISTORICAL COCA is the only large corpus of English that has extensive data from the entire period of the last 30 years – 20 million words per year from 1990-2019 (with the same genre balance year by year). This means that in addition to seeing variation by genre, you can also map out recent changes in English in ways that are not possible with any other corpus – such as with the frequency of awesome from 1990-2019,.

SECTION	ALL	BLOG	WEB	TV/M	SPOK	FIC	MAG	NEWS	ACAD	1990-94	1995-99	2000-04	2005-09	2010-14	2015-19
FREQ	24063	9902	6800	8494	2740	957	1819	1352	350	940	1127	1688	3075	4523	4359
WORDS (M)	993	128.6	124.3	128.1	126.1	118.3	126.1	121.7	119.8	139.1	147.8	146.6	144.9	145.3	144.7
PER MIL	24.23	76.99	54.73	66.32	21.72	8.09	14.43	11.11	2.92	6.76	7.63	11.52	21.21	31.14	30.12
SEE ALL SUB-SECTIONS AT ONCE															

And of course you can look at much more than just simple words or phrases. COCA is the only corpus that allows you to map out changes in syntactic constructions over the past 30 years, as with the "like construction" (and I'm like, no way) or the "end up V-ing" construction (you'll end up paying way too much) – both of which have increased in each five year period since the early 1990s.

SECTION	1990-94	1995-99	2000-04	2005-09	2010-14	2015-19	1990-94	1995-99	2000-04	2005-09	2010-14	2015-19
FREQ	140	393	639	1145	1780	2581	1826	2340	2489	2849	2949	3293
WORDS (M)	139.1	147.8	146.6	144.9	145.3	144.7	139.1	147.8	146.6	144.9	145.3	144.7
PER MIL	1.01	2.66	4.36	7.90	12.25	17.83	13.13	15.83	16.98	19.66	20.30	22.75
SEE ALL	like co	nstructio	n (CONJ I	PRON BE	like ' ,)		END u	v V-ing				
SUB-SECTIONS AT ONCE												

Just as we compared collocates (nearby words) in different genres above (for chair in ACAD and FIC), we can also compare collocates in different periods, to look at semantic change. For example, the following table shows collocates of green in the 1990s (left) and the 2010s (right). SEC 1 (1995-1999, 1990-1994): 286,833,557 WORDS

SEC 2 (2010-2014, 2015-2019): 290,003,115 WORDS

	WORD/ITHVIDE	TORLINGT	TOREINS		11112			WORD/THIVIDE	TORLING	TOREITS	1 111 2		101110
1	GREEN RANGER	187	0	0.7	0.0	65.2	1	GREEN GAZETTE	96	0	0.3	0.0	33.1
2	GREEN PEPPER	215	35	0.7	0.1	6.2	2	GREEN JOBS	87	0	0.3	0.0	30.0
3	GREEN CROSS	52	9	0.2	0.0	5.8	3	GREEN PRACTICE	60	2	0.2	0.0	29.7
4	GREEN VEGETABLES	73	32	0.3	0.1	2.3	4	GREEN ENERGY	170	7	0.6	0.0	24.0
5	GREEN PEPPERS	94	47	0.3	0.2	2.0	5	GREEN ARROW	192	8	0.7	0.0	23.7
6	GREEN MAN	58	31	0.2	0.1	1.9	6	GREEN BUILDING	130	18	0.4	0.1	7.1
7	GREEN ACRES	52	31	0.2	0.1	1.7	7	GREEN SCREEN	85	12	0.3	0.0	7.0
8	GREEN BELL	154	92	0.5	0.3	1.7	8	GREEN ZONE	118	21	0.4	0.1	5.6
9	GREEN PLANTS	61	37	0.2	0.1	1.7	9	GREEN LANTERN	96	21	0.3	0.1	4.5
10	GREEN GLASS	61	37	0.2	0.1	1.7	10	GREEN SPACES	97	23	0.3	0.1	4.2

8	GREEN BELL			154	92	0.	5 0.3	1.7	8 GRI	EN ZONE		118	2	1 (0.1	5.6
9	GREEN PLANTS			61	37	0.	2 0.1	1.7	9 GRI	EN LANTERN		96	2	1 (0.1	4.5
10	GREEN GLASS			61	37	0.	2 0.1	1.7	10 GRI	EN SPACES		97	2	3 (0.1	4.2
in e	We can also see what is being said about a topic over time. For example, the following shows the collocates of crisis in each five year period (and genre) from 1990-2019, which shows what we were worrying about in these different periods. ALL BLOG WEB-GENL TY/MOVIES SPOKEN FICTION MAGAZINE NEWSPAPER ACADEMIC 1990-1994 1995-1999 2000-2004 2005-2009 2010-2014 2015-2019															
CONT	EXT	ALL	BLOG						NEWSPAPER		1990-1994	1995-1999			2010-2014	
DEDT		4540	270	110		254		7.6	202	475	445	60	0.5	20	120	70
DEBT		1513	370	419	4	264	2	76	203	175	115	63	25	29	420	72
CRISIS	5	1088	142	146	64	268	24	140	128	176	194	128	76	116	162	124
HEAL1	гн	917	65	92	36	209	7	214	177	117	149	84	84	133	115	195
ENER	GY	771	87	96	32	129	10	135	196	86	94	43	245	123	48	35
GULF		656		5	3	267	1	122	194	64	629	11	3	2	4	2
OIL		437		38	8	76	8	89	69	91	132	46	64	53	28	18
REFU	GEE	381	10	16	6	122	6	57	65	99	47	48	19	15	27	199
SOLU	TION	381		49	4	97	7	42	67	66	116	31	33	36	40	27
MIDL	IFE	348	32	33	115	62	24	40	21	21	25	31	45	26	104	52
CRED	IT	385	86	82	2	46	1	50	94	24	3	9	2	138	50	15
CLIM	ATE	396	114	80	2	65	4	86	14	31	2	1	12	48	33	106

As with all of the other corpora from English-Corpora.org, there is a very wide range of searches, including: words, phrases, substrings, lemmas, part of speech, synonyms, and customized wordlists. For example, the search WEAR * ADJ @CLOTHES takes just about one second to search through the billion words to find strings like the following (and it doesn't require learning unnecessarily convoluted search syntax).

CONTEXT ALL FORMS (SAMPLE): 100, 200, 500

2016 FIC

4

6

7

8

9

10

11

12

Dialogue: Mormon

FIND COMMON GROUND

TAKE FULL ADVANTAGE

GRATED PARMESAN CHEESE

BEGIN AUDIO CLIP

SAY GOOD NIGHT

COME FULL CIRCLE

COME RIGHT BACK

WORK FULL TIME

TRY NEW THINGS

				CONTEXT ALL FORMS	(SAI	VIP	LC)	. 100 200 500	11112	TOTAL 1,292 UNIQUE 900 +
	1			WEARING A TOP HAT					22	
	2			WEARING A SHORT SKIRT					18	
	3			WEARING A HARD HAT					14	
	4			WEARING A WHITE BLOUS	SE				14	
	5			WEAR A SHORT SKIRT					13	
	6			WEARING A WHITE HAT					13	
	7			WORE A SHORT SKIRT					13	
	8			WEARING A RED TIE					12	
1	2019	FIC	Penr	nLitJournal	Α	В	c	know Jessica, the salesgirl with the husky voice who as usual	wore cuffed	hot pants. I discovered much to my surprise that she lived in southeast, to
2	2019	NEWS	Chica	ago Tribune	Α	В	c	the 2400 block of Partridge Lane told police Feb. 19 that some	neone weari r	ng a dark hat and jacket was looking into a neighbor's vehicle. # A resident i
3	2019	TV	The	Blacklist	Α	В	С	to the defendant? It was the only suit in the area wearing a for	edora-style h	nat. I see. And did you arrest the defendant at that point? No
4	2018	NEWS	Oreg	gonLive.com	Α	В	c	Security disability benefits, punched the guard after he was a	accused of w	rearing a stolen hat on May 12, 2017. # Nathaniel M. Williams, 25Columbia
5	2017	FIC	Bk:O	ne-hundred-knuckledFist	Α	В	С	because he looks toward the sound, too. # A woman wearing	g a bright-ye	llow blouse and black pants hums past his stand, armed with a small vacuu
6	2017	MAG	Fox I	News	Α	В	С	evacuated Tuesday after an explosion was reported and poli	ice shot a su	spect <mark>wearing an explosive belt</mark> who shouted " Allahu Akbar, " in what offici
7	2017	NEWS	USA	TODAY	Α	В	c	in Jerusalem's Old City on May 22, 2017. She wore a white sk	irt suit by Mi	ichael Kors and her customary sky-high heels in a brightly colored striped p
8	2016	FIC	Bk:Le	oveMaps	Α	В	С	my apple! " Maya yelled from the porch. # She wore a white	blouse tied a	at the waist and a blue-and-black plaid skirt. She skipped toward them,

Because of COCA's advanced architecture, even searches for very general searches like **NOUN** + **NOUN** or **VERB ADJ NOUN** take just 1-2 seconds to search through the billion word corpus:

A B C . Approaching her, he saw it was Jennie O'Brien. She wore a drab skirt, and her hair was wound into an untidy bun upon her head. #

683

641

558

553

491

445

	USE LARGE NGRAMS [?]	FREQ	TOTAL 8,108,778 UNIQUE 17,446 +
1	HEALTH CARE	64877	
2	WORLD WAR	32709	
3	LIVING ROOM	23171	
4	LAW ENFORCEMENT	23057	
5	WEB SITE	22382	
6	CLIMATE CHANGE	21967	
7	AIR FORCE	19768	
8	REPUBLICAN PARTY	16928	
9	STATE UNIVERSITY	16926	
10	HEALTH INSURANCE	16378	
11	FAMILY MEMBERS	16031	
	USE LARGE NGRAMS [?]	FREQ	TOTAL 178,237 UNIQUE 6,395 +
1	TAKE GOOD CARE	1226	
2	MAKES PERFECT SENSE	1120	
3	PAY CLOSE ATTENTION	706	

A unique feature of COCA, which makes it very useful for language learners and teachers, is the ability to browse through a list of the top 60,000 words (lemmas) in the corpus, and then to see an extremely wide range of information on each of these words.

Word form	
Part of speech	☑ NOUN ☑ VERB ☑ ADJ ☑ ADV ☑ OTHER □ ALL
Range	5145
Pronunciation	Rhymes with Type EXACT
Syllables / stress	00000000 ×

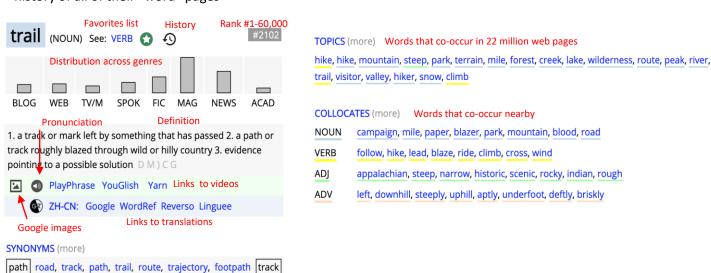
For example, the following are just a few examples of high frequency words (about word #5000 in the 60,000 word list), medium frequency (~25,000), and low frequency (~45,000) words. For each word in the list, users can hear the word **pronounced**, see **videos** with that word in the text, find related **images** from Google Images, an see a **translation** for their preferred target language.

	RANK	FREQ	Word	PoS	Audio	Video	Image	ZH-CN
1	5200	11377	blogger	NOUN	•	(b)		€ ₹
2	5201	11374	utterly	ADV	•	⊚		G _V
3	5202	11372	trouble	VERB	•	(b)	A	€ ₹
4	5203	11368	texture	NOUN	•	(b)		GR)
5	5204	11365	head	ADJ	3	(b)	A	•
	RANK	FREQ	Word	PoS	Audio	Video	Image	ZH-CN
1	25201	577	unappealing	ADJ	•	(D)	_	•
2	25202	577	tryst	NOUN	•	(D)	_	•
3	25203	577	urologist	NOUN	•	(b)	_	©
4	25204	577	scrabble	NOUN	•	(b)		©
5	25205	577	demarcate	VERB	•	ٱ	'	•
	RANK	FREQ	Word	PoS	Audio	Video	Image	ZH-CN
1	45107	112	incubus	NOUN	•	⊳		©
2	45108	112	unquantifiable	ADJ	•		_	©
3	45110	112	husk	VERB	•	(D)	_	©
4	45111	112	hunky-dory	ADJ	•	(D)	_	•
5	45112	112	jacked	ADJ	•	(b)		•

It is even possible to search the 60,000 word list by **pronunciation** (very helpful, because of difficult English spelling). For example, the following is a partial list of two syllable words (accented on the second syllable) that rhyme with *stay*:

	Pronunciation		Rhymes with stay Type EXACT						
	Syllables / stress		000000000 ×						
11	3958	17233	array	NOUN	•	D		•	
12	4122	16221	convey	VERB		⊳		6	
13	4589	13766	survey	VERB	•	D		©	
14	4751	13055	cafe	NOUN	•	D	_	•	
15	4779	12934	betray	VERB	•	D	🔼	©	
16	5536	10325	obey	VERB	•	D	🔼	€ ₹	
17	5933	9149	ballet	NOUN	•	D	_	•	

Each of the top 60,000 words in the corpus has a "home page" such as the following, with links to other pages with more information. Users can save words to a "favorites" list for later review, and go back through a history of all of their "word" pages



2, 3, 4 word strings **CLUSTERS** (more)

trail, trace

• trail	campaign trail • paper trail • appalachian trail • hiking trails • long trail • blood trail • on trails • money trail
trail • •	trail of tears a trail of blood a trail of destruction a trail that led a trail leads to a trail of smoke a trail of bread a trail of broken

- on the trail along the trail up the trail off the trail down the trail leaving a trail left a trail to the trail
- • trail on the campaign trail out on the trail from the campaign trail end of the trail hot on the trail side of the trail on the appalachian trail miles of hiking trails

trail • • • trail of bread crumbs • trail that led to • trail to the top • trail through the woods • trails in the sky • trail of destruction across • trail that leads to • trail of broken

Texts that have this as a "keyword" TEXTS / VIRTUAL CORPORA (more)

trail in • trail to • trail blazers • trail for • trail with • trails in • trail from • trail on

WEB:nps.gov • WEB:ponderosaca.com • NEWS:AssocPress • MAG:Backpacker • ACAD:InstrPsych • MAG:Bicycling • NEWS:USA TODAY • MAG:Backpacker • FIC:FantasySciFi • MAG:Backpacker • MAG:NatlParks • NEWS:USAToday • MAG:Backpacker • MAG:Backpacker • WEB:...documentaryfilms.com • MAG:NatlParks • MAG:Backpacker • MAG:Backpacker • MAG:NatlParks • MAG:NatlPar MAG:Backpacker • MAG:Ba

CONC	ORDANCE LINES (more)	The word in context (to see patterns of use)
1 \//	IED: 2012: amazon com	Along the way, they confront Army Pangare along the Langering Trail and conduct a late night stealth mission to an abandoned NASA

2	BLOG: 2012:heother.blogspot	me , a direct definition of ultra trail <mark>running</mark> . The	trails	are mazing DCNR does an excellent job keeping the trails and
3	MAG: 1991: NatlParks	to protect a continuous corridor are ongoing . The North Country	Trail	Association , a 500-member organization established in 1980 ,
4	MAC: 1001: NatiParks	# For more information, contact the Overmountain Victory	Trail	Association GO Poy 622 Managers Park VA 22111

🔟 📈 <mark>c/o</mark> P.O. Box 632 , Manassas Park , VA 22111

MAG: 2003: Bicycling and anti-bike website operator John Parulis, describing the at China Camp , east of San Rafael " Look at this

at the top , and taking the ski lift back down . WEB: 2012: yelp.com try ... # Perks of coming here is hiking the vari trails

before approving its completion; and the fact that an that the city did not "re-interview" businesses along the BLOG: 2012: seattlemet.com trail bikes out the back if they have n't been taken . Awesome TV: 2016: Tomorrow, When the W... OK? Rain gear 's there too . Should be some

Blazers (owned by billionaire Paul Allen) , the New York NEWS: 1998: Houston because only a few of the wealthiest teams - the Portland

butts a 50-foot cliff off the creek bed . With my blissful 10 MAG: 2010: PopMech , I also fail to notice that the side of the

camera that was placed on the stand farthest from camp, the 11 MAG: 2009: OutdoorLife to find one . He showed me some images from a

can sometimes make another . I told you I 'd back your 12 TV: 2014: Avengers Assemble , arizona . How did ... Technopathic static . Covering one trail climbed steadily up the drainage through thick timber above the 13 MAG: 2007: AmSpect first four miles or so were quite steep, as the

country , the huge and austere Kaiparowits Plateau , and the slot 14 MAG: 1997: Backpacker and its remote eastern tributaries, the vast and Trail downriver . Later , we recognize what 's left of an elk 15 MAG: 2009: Backpacker , so we divert onto a slippery embankment that rejoins the

for 10 or more days at a time , Mike values a 16 MAG: 1997: Backpacker and describes himself as a warm sleeper. Often on the trail

from his kitchen window . But the Tims also ride . Barnack 17 MAG: 2008: Bicycling paths . Barnack can see riders heading up the Tollgate Trail

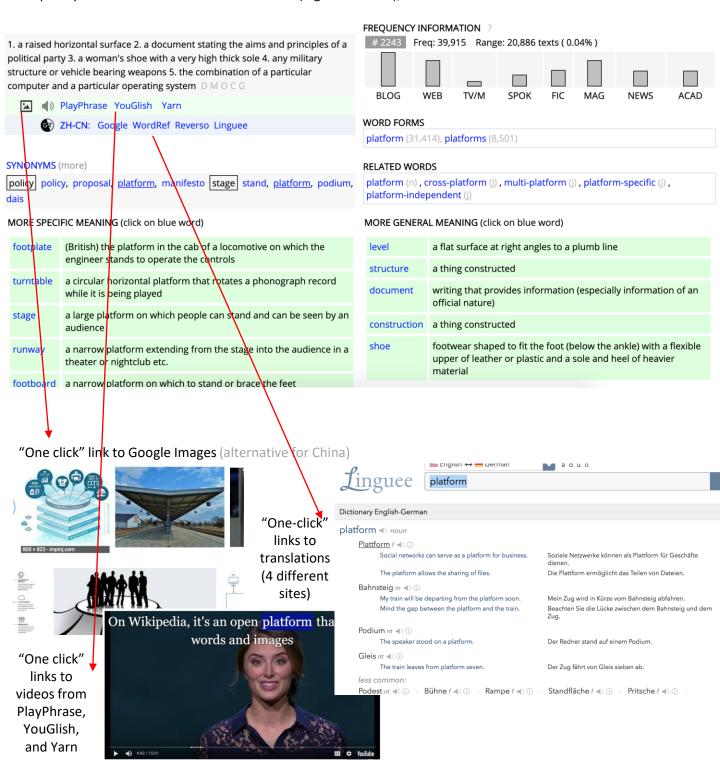
18 NEWS: 2018: OregonLive.com TriMet and PGE are working a section of the Spring from Southeast Tacoma to 45th Place and the widely used trail is

head and picks them up at the bottom . A historian and 19 NEWS: 1992: WashPost tour, the bus drops hikers off at the Skyline Drive 20 MAG: 1997: Backpacker a free backcountry trip planner and general map, or call Illustrated , (800) 962-1643 for its map to the park

Each of the top 60,000 words also has more detailed pages, including a "dictionary" page, related topics, collocates, clusters, websites, and concordance lines. Samples of each of these pages are given below.

DICTIONARY page

Includes a **definition**, links to Google **Images**, **pronunciation**, **videos**, and **translations** (to desired language) at (up to) four different sites. Also includes **synonyms** and words with more specific meanings (**hyponyms**) and more general meanings (**hypernyms**) (both from WordNet). It also includes **frequency information** (including rank order, number of tokens, and two "range" measures of how well the word is spread throughout the ~500,000 texts, as well as its distribution across the eight main genres in the corpus. Finally, it also includes the frequency of the different **forms** of the lemma (e.g. verb forms), and **related words**.



TOPICS page

.....

Shows words that tend to co-occur in the 22 million webpages in the corpus. In many cases these provide better insight into meaning and usage than collocates (the standard tool for finding textually related words), and yet we're not aware of any other corpus that has these. Users can click on any of these words to follow a "chain" of related words.

TOPICS	HORMONE	NOUN		TOPICS	COSMOLOG	ICAL ADJ		TOPICS	VIBRATE	VERB	
	Same text	Word	PoS		Same text	Word	PoS		Same text	Word	PoS
1	253	estrogen	n	1	108	universe	n	1	40	vibration	n
2	245	supplement	n	2	82	bang	n	2	30	frequency	n
3	227	stimulate	v	3	73	physics	n	3	25	physics	n
4	224	therapy	n	4	71	cosmic	j	4	23	atom	n
5	209	cancer	n	5	70	cosmology	n	5	22	particle	n
6	203	disease	n	6	69	galaxy	n	6	21	quantum	n
7	202	prescribe	v	7	68	quantum	n	7	19	faint	j
8	199	symptom	n	8	62	physicist	n	8	19	string	n
9	192	diet	n	9	59	particle	n	9	17	molecule	n
10	187	breast	n	10	57	astronomer	n	10	17	sound	n
11	174	stress	n	11	55	gravity	n	11	17	wave	n
12	171	nutrient	n	12	54	expansion	n	12	16	electron	n
13	168	obesity	n	13	53	theory	n	13	16	gravity	n
14	161	protein	n	14	50	relativity	n	14	16	magnetic	j
15	161	researcher	n	15	48	atom	n	15	16	mathematical	j
16	160	induce	v	16	45	constant	j	16	16	motion	n
17	160	injection	n	17	45	correspond	v	17	16	physicist	n
18	160	tissue	n	18	44	gravitational	j	18	16	vacuum	n
19	158	medicine	n	19	44	mathematical	j	19	15	grin	V

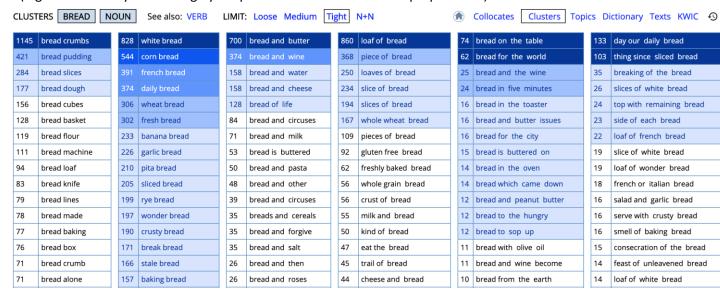
COLLOCATES page

Words that occur nearby. Click on any word to see the collocates for the new word, or click on the "text" icon to see the word in context. The gray boxes show the most normal placement of the node word (e.g. *Christmas tree*, but *tree trunk*). You can also sort by Mutual Information score and set frequency thresholds (Advanced Options).



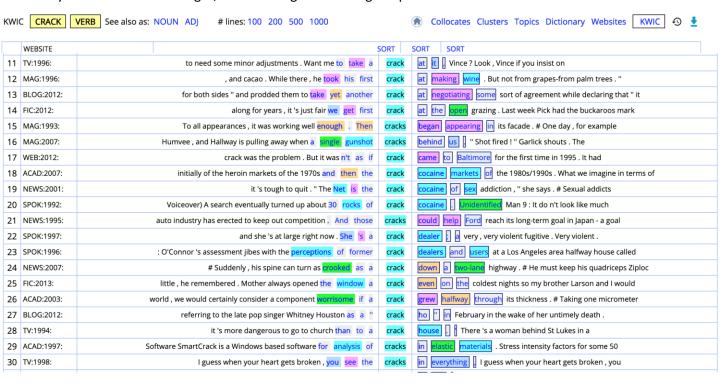
CLUSTERS page

The most common 2, 3, and 4 word strings. You can decide how "wide" or "tight" you want the clusters to be (e.g. whether they include highly frequent words like articles and prepositions).



KWIC / concordance page

See 100 - 1000 random lines, with the surrounding words highlighted for part of speech. You can re-sort the words by words to the left or right, all with the goal of seeing the patterns in which the word occurs.



Texts page

Create Virtual Corpus

Freq	Text (click for topics)	Words
160	SPOK: NPR_Science SCIENCE, HISTORY AND CHEMISTRY OF CHOCOLATE M	chocolate, cocoa, tree, bean, bird, coffee, flavor, migratory, crop, interesting, farmer, forest, tropical, rain, plantation, growing, wine, texture, ferment, cacao,
119	WEB:eyourgrandmother.com How To Make Perfect Brownies How To Cook Li	brownie, recipe, chocolate, butter, chip, add, cocoa, ingredient, sugar, bake, perfect, cup, egg, texture, flour, powder, salt, pan, oven, batch,
69	NEWS: WashPost OOZE AND AAHS;	chocolate, cake, dessert, gm, mixture, egg, chef, pan, sugar, cream, serve, melt, pastry, butter, fat, mg, inn, ounce, cup, warm,
64	NEWS: WashPost Beyond Bittersweet; Once, any kid with a few	chocolate, cocoa, percent, bean, dark, bar, manufacturer, flavor, bitter, percentage, sugar, milk, plantation, origin, dark, product, specialty, growing, prefer, association,
61	MAG: CountryLiving Cook Book: A fine Romance	chocolate, cocoa, set, tablespoon, cup, sugar, beat, teaspoon, butter, bowl, minute, aside, vanilla, heat, pan, flour, baking, add, min, salt,
60	NEWS: USAToday A Valentine to Chocolate	chocolate, specialty, cocoa, praline, shop, information, belgian, butter, cream, bean, sweet, fill, sugar, filling, liquor, passion, box, supplier, demonstration, nearly,

Move between page showing the texts in which the word is a keyword (see above), and the page showing all keywords from a specific texts (right).

WORDS FROM: NEWS: USAToday: A Valentine to Chocolate

	Frequency	Word	PoS	Topics	Collocates
1	60	chocolate	n	Topics	Collocates
2	18	specialty	n	Topics	Collocates
3	16	cocoa	n	Topics	Collocates
4	15	praline	n	Topics	Collocates
5	13	information	n	Topics	Collocates
6	13	shop	n	Topics	Collocates
7	10	belgian	j	Topics	Collocates
8	10	liquor	n	Topics	Collocates
9	9	butter	n	Topics	Collocates
10	8	cream	n	Topics	Collocates
11	7	bean	n	Topics	Collocates
12	7	sweet	j	Topics	Collocates
13	6	fill	V	Topics	Collocates

Summary

COCA has a number of features that sets it apart from any other corpus. These include its **size** (1.0 billion words), how up-to-date it is (texts through Dec 2019), the wide range of **genres** (TV/Movie subtitles, spoken, blogs, web, fiction, magazine, newspaper, academic), and its **searches** (range of query types, and the ease and speed of its searches), including the ability to limit to and compare across genres and time periods.

In addition, it is different from most of the other corpora from English-Corpora.org in the attention that it gives to the **top 60,000 words** in the corpus, and the wide range of information for each word, including frequency information, definitions, synonyms, WordNet entries, related topics, concordances (new display in COCA), clusters, websites that have the word as a "keyword", and KWIC/concordance lines.

All of these features make COCA the ideal corpus for researchers, teachers, and learners.