Analyzing word: full
Top 10 similar words (custom model):
being: 0.4325
believe: 0.4212
found: 0.4205
feeling: 0.4192
friday: 0.4185
football: 0.3860
mind: 0.3818
classic: 0.3786
club: 0.3754
duck: 0.3734
Top 10 opposite words (custom model):
cakes: 0.0242
always: 0.0201
italian: 0.0187
groceries: 0.0145
tastes: 0.0131
market: 0.0087
seafood: -0.0041
shop: -0.0072
does: -0.0174
enough: -0.0607
Top 10 similar words (pre-trained model):
fulll: 0.6939
FULL: 0.6360
Full: 0.6175
chock: 0.6144
complete: 0.5861
near-full: 0.5773
chocked: 0.5493
partial: 0.5305
-full: 0.5143
fledged: 0.5102
Top 10 opposite words (pre-trained model):
story.Read: 0.2972
wonderful: 0.2972
three-to-four: 0.2972
loaded: 0.2971
satisfyingly: 0.2971
ofs: 0.2971
desciption: 0.2971
full-view: 0.2971
34-page: 0.2970
120-second: 0.2970

Analyzing word: dessert
Top 10 similar words (custom model):
caramel: 0.5589
filet: 0.4860
creamy: 0.4805
mimosas: 0.4704
chip: 0.4700

creme: 0.4686
bottomless: 0.4444
garlic: 0.4298
beans: 0.4233
velvet: 0.4189
Top 10 opposite words (custom model):
wouldn: 0.0242
sure: 0.0226
tuesdays: 0.0147
thank: 0.0089
would: 0.0075
recommend: 0.0016
today: -0.0021
thanks: -0.0026
after: -0.0151
phone: -0.0761
Top 10 similar words (pre-trained model):
desserts: 0.8228
dessert-: 0.7928
Dessert: 0.7358
dessert.This: 0.7300
dessert.The: 0.7240
tiramisu: 0.7161
cheesecake: 0.6951
dessert.I: 0.6939
dessert.: 0.6913
non-dessert: 0.6781
Top 10 opposite words (pre-trained model):
delicious.It: 0.4409
'oevres: 0.4409
food-laden: 0.4408
kugelhopf: 0.4408
TIRAMISU: 0.4408
caramel-covered: 0.4408
Gugelhupf: 0.4408
dourves: 0.4407
kuihs: 0.4405
pancakes.I: 0.4405

Analyzing word: been
Top 10 similar words (custom model):
beignets: 0.4833
nachos: 0.4495
alcohol: 0.4483
seen: 0.4452
date: 0.4310
chance: 0.4285
eaten: 0.4223
believe: 0.4136
beef: 0.4115
building: 0.4059
Top 10 opposite words (custom model):
short: 0.0315

noodles: 0.0315
with: 0.0242
tonight: 0.0225
stop: 0.0196
recommend: 0.0171
post: 0.0152
start: 0.0133
wife: 0.0115
tomato: 0.0109
Top 10 similar words (pre-trained model):
beeen: 0.7583
has: 0.7313
have: 0.7284
had: 0.7033
already: 0.6949
tbeen: 0.6669
beenn: 0.6529
recently: 0.6378
gotten: 0.6339
beent: 0.6317
Top 10 opposite words (pre-trained model):
hadnot: 0.3709
hasalso: 0.3709
supplanted: 0.3708
disussed: 0.3707
succeeded: 0.3707
denied: 0.3707
noticed.I: 0.3706
acutally: 0.3705
viewed: 0.3705
changing.I: 0.3704

Analyzing word: raining
Top 10 similar words (custom model):
dining: 0.7385
using: 0.5444
evening: 0.5427
running: 0.5251
opening: 0.5237
warning: 0.4993
tables: 0.4942
prepared: 0.4891
cleaning: 0.4889
building: 0.4603
Top 10 opposite words (custom model):
shrimp: 0.0425
cook: 0.0419
roll: 0.0353
oysters: 0.0271
seasoned: 0.0260
free: 0.0229
rolls: 0.0183
delicious: 0.0127

lobster: 0.0023
fresh: -0.0146
Top 10 similar words (pre-trained model):
rained: 0.7936
rain: 0.7297
snowing: 0.7207
raining.: 0.6992
downpouring: 0.6929
sleeting: 0.6837
thunderstorming: 0.6800
rains: 0.6793
rain.It: 0.6773
rainning: 0.6736
Top 10 opposite words (pre-trained model):
sunlight: 0.3249
whooshed: 0.3249
thundered: 0.3248
dry.: 0.3248
night: 0.3247
flat-light: 0.3246
skys: 0.3246
saturday: 0.3246
arrived.It: 0.3245
stir-crazy: 0.3244

Analyzing word: quiet
Top 10 similar words (custom model):
wifi: 0.5582
croissants: 0.5282
quick: 0.5185
valet: 0.5058
buffet: 0.4826
bisque: 0.4800
play: 0.4743
actually: 0.4654
taking: 0.4652
venue: 0.4464
Top 10 opposite words (custom model):
sauce: 0.0623
asked: 0.0604
less: 0.0602
shrimp: 0.0599
chili: 0.0590
then: 0.0574
orange: 0.0561
lucky: 0.0458
pepper: 0.0350
seasoned: 0.0310
Top 10 similar words (pre-trained model):
Quiet: 0.7148
peaceful: 0.7078
tranquil: 0.7045
quiet.It: 0.7014

quiet.A: 0.6881
quietness: 0.6828
quiet.The: 0.6676
quiet.This: 0.6658
quiet-: 0.6626
quiet.Very: 0.6609
Top 10 opposite words (pre-trained model):
well-equiped: 0.3633
clamorous: 0.3633
Well-furnished: 0.3633
classy: 0.3630
00ZComfortable: 0.3630
sleep.This: 0.3630
jovial: 0.3629
freindly: 0.3629
noise.A: 0.3629
non-pretentious: 0.3628

Analyzing word: http
Top 10 similar words (custom model):
celebrating: 0.5508
philly: 0.5111
actually: 0.5000
list: 0.4918
interesting: 0.4900
yelp: 0.4878
view: 0.4862
king: 0.4859
2016: 0.4820
coming: 0.4568
Top 10 opposite words (custom model):
meal: 0.0505
place: 0.0502
turkey: 0.0478
does: 0.0467
night: 0.0459
take: 0.0444
fine: 0.0434
snack: 0.0420
soup: 0.0312
that: 0.0102
Top 10 similar words (pre-trained model):
https: 0.8264
hhttp: 0.8230
-http: 0.8002
thttp: 0.7894
.http: 0.7484
htttp: 0.7481
andhttp: 0.7425
ttp: 0.7399
orhttp: 0.7322
ahttp: 0.7319
Top 10 opposite words (pre-trained model):

web.stagram.com: 0.5461
kids.http: 0.5461
www.pennlive.com: 0.5461
www.projectwedding.com: 0.5459
www.nvic.org: 0.5459
www.xfire.com: 0.5458
my.opera.com: 0.5458
www.crainsnewyork.com: 0.5457
worldhttp: 0.5456
www.slideshare.net: 0.5456

Analyzing word: noodles
Top 10 similar words (custom model):
kinda: 0.5011
thai: 0.4955
tuna: 0.4933
pancakes: 0.4925
spaghetti: 0.4813
unless: 0.4805
tender: 0.4752
eggs: 0.4739
value: 0.4677
waffles: 0.4651
Top 10 opposite words (custom model):
longer: 0.0422
open: 0.0391
here: 0.0350
been: 0.0315
long: 0.0281
hope: 0.0240
closed: 0.0225
able: 0.0222
hours: 0.0134
last: 0.0039
Top 10 similar words (pre-trained model):
noodle: 0.8130
noddles: 0.7973
noodles.: 0.7406
noodles-: 0.7289
noodles.I: 0.7084
udon: 0.7034
vermicelli: 0.7031
ramen: 0.6953
pasta: 0.6872
Noodles: 0.6665
Top 10 opposite words (pre-trained model):
c■m: 0.4108
mayonaisse: 0.4108
murukkus: 0.4108
poppadums: 0.4108
spiraliser: 0.4107
Bibimbap: 0.4107
ajinomoto: 0.4107

risoni: 0.4106
ragù: 0.4105
dosais: 0.4104

Analyzing word: baby
Top 10 similar words (custom model):
shopping: 0.5848
girl: 0.5005
purchase: 0.4999
football: 0.4922
juice: 0.4881
doing: 0.4774
bakery: 0.4664
courtyard: 0.4580
second: 0.4474
secret: 0.4434
Top 10 opposite words (custom model):
stopped: 0.0553
everything: 0.0541
fish: 0.0540
chicken: 0.0517
very: 0.0463
lots: 0.0454
must: 0.0393
onion: 0.0359
chips: 0.0205
well: 0.0093
Top 10 similar words (pre-trained model):
babies: 0.7998
newborn: 0.7903
baby-: 0.7148
infant: 0.7081
Baby: 0.7038
baby-to-be: 0.6873
baby.A: 0.6867
baby.: 0.6753
-baby: 0.6735
baby.The: 0.6709
Top 10 opposite words (pre-trained model):
lullabies: 0.4231
surro: 0.4231
girl.So: 0.4230
parenthood: 0.4229
babiesBaby: 0.4229
Post-pregnancy: 0.4228
skin-to-skin: 0.4227
weeks-old: 0.4226
neonates: 0.4224
Mums-to-be: 0.4224

Analyzing word: italian
Top 10 similar words (custom model):
chicago: 0.4434

asian: 0.4372
chocolate: 0.3811
real: 0.3807
fair: 0.3771
caramel: 0.3732
true: 0.3727
meats: 0.3699
blackberry: 0.3681
christmas: 0.3651
Top 10 opposite words (custom model):
smells: 0.0044
tonight: 0.0036
dont: -0.0039
after: -0.0072
free: -0.0073
take: -0.0083
watch: -0.0109
times: -0.0129
thanks: -0.0153
dinner: -0.0473
Top 10 similar words (pre-trained model):
Italian: 0.7742
italians: 0.7200
sicilian: 0.7178
italian.: 0.7070
itallian: 0.7010
-Italian: 0.6847
Italian.: 0.6840
Italain: 0.6811
sardinian: 0.6797
ltalian: 0.6724
Top 10 opposite words (pre-trained model):
egyption: 0.4382
giordano: 0.4380
serb: 0.4380
cocco: 0.4380
Japaneese: 0.4379
malato: 0.4379
cajun: 0.4379
turkmen: 0.4379
tastful: 0.4379
portugueses: 0.4378

Analyzing word: ginger
Top 10 similar words (custom model):
club: 0.5393
turkey: 0.5210
gift: 0.5203
killer: 0.4816
curry: 0.4793
juice: 0.4694
mmmm: 0.4668
girl: 0.4498

bakery: 0.4482
baby: 0.4394
Top 10 opposite words (custom model):
very: 0.0301
right: 0.0248
thru: 0.0244
drink: 0.0193
last: 0.0151
foot: 0.0132
food: 0.0102
people: 0.0099
soon: -0.0034
location: -0.0363
Top 10 similar words (pre-trained model):
ginger-: 0.7143
gingery: 0.7073
gingers: 0.6843
ginge: 0.6748
lemongrass: 0.6533
ginger.: 0.6480
cinnamon: 0.6350
coriander: 0.6256
Ginger: 0.6237
lemon-ginger: 0.6218
Top 10 opposite words (pre-trained model):
almonds-: 0.4010
redheads: 0.4010
malt: 0.4010
tangerines: 0.4009
lemonade.: 0.4009
pepper-crusted: 0.4009
non-caffeine: 0.4009
1teaspoon: 0.4009
zest: 0.4008
shiitake: 0.4008

Analyzing word: lucky
Top 10 similar words (custom model):
finally: 0.5075
gets: 0.4624
forever: 0.4513
week: 0.4411
changes: 0.4408
diner: 0.4399
changed: 0.4354
eaten: 0.4347
chance: 0.4235
website: 0.4220
Top 10 opposite words (custom model):
cheese: 0.0517
burger: 0.0495
quiet: 0.0458
variety: 0.0421

short: 0.0328
pasta: 0.0243
burgers: 0.0161
fresh: 0.0052
spicy: -0.0046
relish: -0.0286
Top 10 similar words (pre-trained model):
fortunate: 0.7904
unlucky: 0.7218
super-lucky: 0.6880
luckier: 0.6871
lucky.I: 0.6446
luckiest: 0.6361
not-so-lucky: 0.6186
lucky-: 0.6148
LUCKY: 0.6084
lucky.So: 0.5880
Top 10 opposite words (pre-trained model):
penniless: 0.3202
crapshoot: 0.3202
non-adventurous: 0.3201
well: 0.3201
sweetie.: 0.3200
lous: 0.3200
toughing: 0.3199
oh-so-very: 0.3199
bad.One: 0.3198
most-talented: 0.3198

Analyzing word: karaoke
Top 10 similar words (custom model):
tomorrow: 0.6300
coke: 0.5783
wednesday: 0.5532
wednesdays: 0.5507
tomato: 0.5282
totally: 0.5097
wrong: 0.5011
vegan: 0.4878
unless: 0.4823
snack: 0.4783
Top 10 opposite words (custom model):
next: 0.0728
place: 0.0694
outside: 0.0690
made: 0.0666
been: 0.0570
about: 0.0501
look: 0.0479
moved: 0.0307
offered: 0.0255
ordered: -0.0354
Top 10 similar words (pre-trained model):

kareoke: 0.8304
karoake: 0.8300
karoke: 0.8145
Karaoke: 0.8070
Kareoke: 0.7333
karaokes: 0.7287
Karoke: 0.7192
videoke: 0.7147
Karoake: 0.6633
karaoke-ing: 0.6623
Top 10 opposite words (pre-trained model):
entertainers: 0.3533
waitressing: 0.3533
musicIn: 0.3533
seven-eleven: 0.3533
retox: 0.3532
entertainments: 0.3532
hoola-hoop: 0.3531
Dangdut: 0.3531
music-writing: 0.3531
konpa: 0.3531

Analyzing word: using
Top 10 similar words (custom model):
raining: 0.5444
serving: 0.5403
closing: 0.5310
warning: 0.4958
building: 0.4883
single: 0.4713
pass: 0.4695
running: 0.4669
walk: 0.4588
dancing: 0.4577
Top 10 opposite words (custom model):
sandwich: 0.0327
favorite: 0.0300
pretty: 0.0297
food: 0.0279
lobster: 0.0214
sandwiches: 0.0200
love: 0.0168
homemade: 0.0146
bread: -0.0073
little: -0.0328
Top 10 similar words (pre-trained model):
utilizing: 0.7087
Using: 0.6996
use: 0.6784
utilising: 0.6635
usind: 0.6362
employing: 0.6152
useing: 0.6101

usiing: 0.5995
usign: 0.5988
usuing: 0.5949
Top 10 opposite words (pre-trained model):
pre-composed: 0.3378
wiith: 0.3377
tranfering: 0.3377
asuming: 0.3377
ie.: 0.3376
liking: 0.3376
anng: 0.3376
thated: 0.3376
best-matching: 0.3375
re-purposed: 0.3375

Analyzing word: creamy
Top 10 similar words (custom model):
cream: 0.7281
creme: 0.6090
fall: 0.5670
crab: 0.5588
blueberry: 0.5367
foods: 0.5250
boba: 0.5228
blue: 0.5146
bisque: 0.5073
buffet: 0.5029
Top 10 opposite words (custom model):
stopped: 0.0537
phone: 0.0488
side: 0.0477
would: 0.0402
when: 0.0402
says: 0.0338
next: 0.0183
that: 0.0182
seen: 0.0155
pleasant: -0.0295
Top 10 similar words (pre-trained model):
velvety: 0.7401
ultra-creamy: 0.7272
Creamy: 0.7239
extra-creamy: 0.7214
creaminess: 0.7207
super-creamy: 0.7191
mousse-like: 0.7148
cream-like: 0.7108
creamier: 0.6969
buttery: 0.6901
Top 10 opposite words (pre-trained model):
golden-brown: 0.4520
yummy-ness: 0.4520
translucent: 0.4519

frosting-: 0.4519
moisty: 0.4519
pink-purple: 0.4518
toffe: 0.4517
sea-salted: 0.4517
noodle-y: 0.4516
supple: 0.4515

Analyzing word: heard
Top 10 similar words (custom model):
card: 0.5780
charging: 0.5288
forward: 0.5032
give: 0.4845
consistent: 0.4811
board: 0.4787
head: 0.4741
anyone: 0.4683
couldn: 0.4577
girl: 0.4445
Top 10 opposite words (custom model):
there: 0.0316
them: 0.0313
tuna: 0.0264
from: 0.0253
worth: 0.0196
hands: 0.0150
monday: 0.0106
tasty: -0.0005
homemade: -0.0052
beautiful: -0.0060
Top 10 similar words (pre-trained model):
hear: 0.7474
heared: 0.7006
heard.I: 0.6804
listened: 0.6577
heard.As: 0.6209
heard.And: 0.6196
seen: 0.6129
heard-: 0.6046
heard.If: 0.5992
wondered: 0.5984
Top 10 opposite words (pre-trained model):
idea: 0.3594
mis-hearing: 0.3594
seen.The: 0.3593
shrieking: 0.3593
was.Just: 0.3593
hearing.But: 0.3592
excitedly: 0.3592
wrote.I: 0.3592
complaints: 0.3592
neevr: 0.3592

Analyzing word: mediocre
Top 10 similar words (custom model):
pass: 0.5445
nachos: 0.4991
forever: 0.4886
beware: 0.4569
beer: 0.4445
taken: 0.4404
solid: 0.4392
hair: 0.4356
bucks: 0.4266
boba: 0.4223
Top 10 opposite words (custom model):
shop: 0.0410
many: 0.0393
move: 0.0391
work: 0.0372
lots: 0.0342
made: 0.0326
nice: 0.0134
next: 0.0121
stop: 0.0075
this: -0.0461
Top 10 similar words (pre-trained model):
sub-par: 0.8510
so-so: 0.8045
subpar: 0.7994
medicore: 0.7634
lackluster: 0.7353
middling: 0.7236
lousy: 0.7192
mediocre-at-best: 0.7180
sub-mediocre: 0.7152
unimpressive: 0.7124
Top 10 opposite words (pre-trained model):
shaky: 0.4057
thin-sounding: 0.4055
Pathetically: 0.4055
retread: 0.4054
gaudy: 0.4054
unfocused: 0.4054
overrall: 0.4053
under-selling: 0.4053
Substandard: 0.4052
third-worst: 0.4051

Analyzing word: under
Top 10 similar words (custom model):
tender: 0.5244
ownership: 0.5123
bartender: 0.4947
warm: 0.4803

thursdays: 0.4332
unfortunately: 0.4325
sister: 0.4211
nachos: 0.4165
please: 0.4074
bartenders: 0.4044
Top 10 opposite words (custom model):
black: 0.0249
dogs: 0.0239
glad: 0.0216
between: 0.0213
awesome: 0.0209
done: 0.0168
beautiful: -0.0035
first: -0.0060
eggs: -0.0080
enjoy: -0.0182
Top 10 similar words (pre-trained model):
Under: 0.7227
uder: 0.6465
underthe: 0.6444
uner: 0.6437
udner: 0.6349
.under: 0.5576
auspices: 0.5573
nnder: 0.5527
undere: 0.5480
it.Under: 0.5307
Top 10 opposite words (pre-trained model):
witthe: 0.2605
thher: 0.2604
downward: 0.2604
wriggle: 0.2604
relinquish: 0.2604
consideration.The: 0.2604
whithin: 0.2603
any: 0.2603
instituted: 0.2603
control.I: 0.2603

Analyzing word: world
Top 10 similar words (custom model):
worst: 0.5809
work: 0.5211
words: 0.5187
wife: 0.4688
products: 0.4615
worth: 0.4582
caramel: 0.4527
play: 0.4503
would: 0.4413
lasagna: 0.4383
Top 10 opposite words (custom model):

little: -0.0028
drinks: -0.0039
because: -0.0049
coffee: -0.0092
beers: -0.0092
ends: -0.0112
find: -0.0199
bowling: -0.0209
fabulous: -0.0225
packed: -0.0393
Top 10 similar words (pre-trained model):
world.The: 0.6803
world.It: 0.6761
world.But: 0.6738
wolrd: 0.6681
world.What: 0.6665
world-: 0.6632
world--the: 0.6610
world.That: 0.6579
world.This: 0.6566
world.In: 0.6565
Top 10 opposite words (pre-trained model):
time-the: 0.3585
post-petroleum: 0.3585
bizarro: 0.3585
premier: 0.3585
region: 0.3584
internetland: 0.3584
word-wide: 0.3583
globalization: 0.3582
maddest: 0.3582
world.https: 0.3582

Analyzing word: mini
Top 10 similar words (custom model):
groceries: 0.5834
sliders: 0.4867
variety: 0.4700
truffle: 0.4622
cider: 0.4594
gotta: 0.4552
most: 0.4395
margarita: 0.4360
minute: 0.4341
wifi: 0.4276
Top 10 opposite words (custom model):
starbucks: 0.0390
between: 0.0362
start: 0.0340
full: 0.0325
arrived: 0.0301
gave: 0.0247
nail: 0.0197

customer: 0.0123
rolls: 0.0033
state: -0.0052
Top 10 similar words (pre-trained model):
mini-: 0.7273
mini-mini: 0.6930
Mini: 0.6450
miniture: 0.6056
miniature: 0.6011
minis: 0.5971
micro-mini: 0.5827
super-mini: 0.5766
mini-sized: 0.5667
minature: 0.5597
Top 10 opposite words (pre-trained model):
lunchbags: 0.3347
rainbow-themed: 0.3346
car-sized: 0.3346
chibi: 0.3346
macbook: 0.3345
pro: 0.3345
n97: 0.3345
28-piece: 0.3344
quesadillas: 0.3344
playskool: 0.3344

Analyzing word: shop
Top 10 similar words (custom model):
shopping: 0.5892
shoe: 0.5851
shot: 0.5198
show: 0.5198
secret: 0.4877
short: 0.4360
ginger: 0.4152
locally: 0.4141
however: 0.4079
hope: 0.4057
Top 10 opposite words (custom model):
that: 0.0353
vodka: 0.0350
took: 0.0324
cash: 0.0294
when: 0.0283
beer: 0.0161
family: -0.0050
full: -0.0072
were: -0.0289
checked: -0.0487
Top 10 similar words (pre-trained model):
shops: 0.7265
shop.Now: 0.7217
shop.When: 0.7124

shop.We: 0.7104
shop.But: 0.7057
shop.: 0.7037
shop.I: 0.7005
shop.This: 0.6928
shop.It: 0.6910
store: 0.6897
Top 10 opposite words (pre-trained model):
buy.A: 0.3756
boutique-y: 0.3756
job-lot: 0.3756
dry-cleaning: 0.3755
baristas: 0.3755
coffeshops: 0.3755
cabinet-making: 0.3754
car-boot: 0.3754
beautician: 0.3753
caffe: 0.3752

Conclusion:
In this analysis, we explored the similarity and dissimilarity of words using both a custom FastText model and a pre-trained FastText model.
We found that at some words the pretrained model works better than the custom model as it really give very accurate results but at some other words the pretrained model just gives different forms of the same given word but the custom model gives diffrent words most of them are close to thegiven word meaning .
Overall, the results indicate that two models works pretty good but its word dependent if the word is rare or not in the pre-trained model's vocabulary, it may not perform well.