BI 1	query	BI / read / 1				
BI 2	title	Posting summary				
BI 3 BI 4 BI 5 BI 6 BI 7	pattern	message: Message  creationDate < \$dateTime  length year(creationDate)				
BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14 BI 15 BI 16 BI 17 BI 18	desc.	Given a datetime, find all Messages created before that moment. Group them by a 3-level grouping:  1. by year of creation 2. for each year, group into Message types: is Comment or not 3. for each year-type group, split into four groups based on length of their content  • $\emptyset$ : $0 \le \text{length} < 40 \text{ (short)}$ • $1$ : $40 \le \text{length} < 80 \text{ (one liner)}$ • $2$ : $80 \le \text{length} < 160 \text{ (tweet)}$ • $3$ : $160 \le \text{length} \text{ (long)}$				
BI 19 BI 20	params	1 datetime DateTime For later microbatches, later datetime parameters are selected keep the variance low (<0.5%)				
	result	1 year 32-bit Integer R year(message.creationDate) 2 isComment Boolean M True for Comments, False for Posts 3 lengthCategory 32-bit Integer C long 4 messageCount 32-bit Integer A Total number of Messages in that group 5 averageMessageLength 32-bit Integer A Sum of all Message content in that group 6 sumMessageLength 32-bit Integer A Sum of all Message content lengths 7 percentageOfMessages 32-bit Float A percentage of all messages created before the given date				
	sort CPs	1 year ↓ ↓ 2 isComment ↑ False < True, i.e. Posts come first and Comments second 3 lengthCategory ↑ order based on the lengthCategory value  1.2, 3.2, 4.1, 4.2, 8.5				

BI 1	query	BI / read / 2								
BI 2	title	Tag evolution								
BI 3										
BI 4		TagClass  name = \$tagClass								
BI 5		<u> </u>								
BI 6		hasType								
BI 7		tag: Tag								
BI 8	pattern	hasTag hasTag								
BI 9		countWindow1 = count(message) countwindow2 = count(message)								
BI 10										
BI 11		message: Message message: Message								
BI 12 BI 13		creationDate in [\$date, \$date+100 days)   creationDate in [\$date+100 days, \$date+200 days)								
BI 13		Find the Tags under a given TagClass that were used in Messages during in the 100-day period								
BI 14 BI 15	desc.	starting at date and compare it with the 100-day period that follows. For the Tags and for both								
BI 16	uesc.	months, compute the count of Messages.								
BI 17		months, compute the count of wiessages.								
BI 18		1 date Date								
BI 19	params	2 tagClass Long String TagClasses with a similar amount of Messages are selected								
BI 20										
		1 tag.name Long String R								
		2 countWindow1 32-bit Integer A Occurrences of the tagClass during the first time window								
	result	3 countWindow2 32-bit Integer A Occurrences of the tagClass during the second time window								
		4 diff 32-bit Integer A Absolute difference of countWindow1 and countWindow2								
	sort	1 diff $\downarrow$								
	3011	2 tag.name ↑								
	limit	100								
	CPs	2.4, 3.1, 3.2, 4.1, 4.2, 4.3, 5.3, 6.1, 8.2, 8.5								

BI 1	query	BI / read / 3						
BI 2	title	Popular topics in a country						
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14	pattern	Country  name = \$country  isPartOf  City  Tag  hasType  City  Tag  hasTag  count(message)  id  message: Message  replyOf*O  id title creationDate  TagClass  name = \$tagClass  name = \$tagClas						
BI 14 BI 15 BI 16 BI 17 BI 18 BI 19 BI 20	desc.	Given a TagClass and a Country, find all the Forums created in the given Country, containing at least one Message with Tags belonging directly to the given TagClass, and count the Messages by the Person who created it and by the Forum which contains them.  The location of a Forum is identified by the location of the Forum's moderator.  1 tagClass Long String TagClasses with a similar amount of Messages are selected						
D1 20	params	2 country Long String Big Countries are selected						
	result	1 forum.id ID R 2 forum.title Long String R 3 forum.creationDate DateTime R 4 person.id ID R 5 messageCount 32-bit Integer A						
	sort	1 messageCount ↓ 2 forum.id ↑						
	limit	20						
	1.1, 1.2, 1.3, 2.1, 2.2, 2.4, 3.3, 8.2							

BI 1	query	BI / read / 4					
BI 2	title	Top message creators by country					
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14 BI 15	pattern	1. select top 100 forums based on memberCount in country  Country  Iname  IsPartOf  City  IsLocatedIn  MemberCount = count(member)  Message  Message  Iname  Is in top 100 forum  Message  Iname  Is in top 100 forum  In in the top posters  In in top 100 forum  Is in top 100 forum  In in the top posters  In in top 100 forum  In in top 100 forum  Is in top 100 forum  In in top 100 forum  In in top 100 forum  Is in top 100 forum  In in top 100 forum  Is in top 100 forum  In in top 100 forum					
BI 16 BI 17 BI 18 BI 19 BI 20	desc.	Find the most popular Forums by Country, where the popularity of a Forum is measured by the number of members that Forum has from a given Country.  Calculate the top 100 most popular Forums. In case of a tie, the Forum(s) with the smaller id value(s) should be selected.  For each member Person of the 100 most popular Forums, count the number of Messages (messageCount) they made in any of those (most popular) Forums. Also include those member Persons who have not posted any Messages (have a messageCount of 0).					
	params	1 date Date Selected from the first 30 days of the network					
	result	1         person.id         ID         R           2         person.firstName         String         R           3         person.lastName         String         R           4         person.creationDate         DateTime         R           5         messageCount         32-bit Integer         A					
	sort	1 messageCount ↓ 2 person.id ↑					
	limit	100					
	CPs	s 1.2, 1.3, 2.1, 2.2, 2.3, 2.4, 3.3, 5.3, 6.1, 8.2, 8.4					

BI 1	query	BI / read / 5			
BI 2	title	Most active posters of a given topic			
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11	pattern	Tag person: Person id hasCreator  hasTag hasCreator  person.score = 1×messageCount + 2×replyCount + 10×likeCount  replyCount = count(comment)  wopt replyOf comment: Comment			
BI 12 BI 13 BI 14 BI 15 BI 16 BI 17 BI 18 BI 19	desc.	Get each Person (person) who has created a Message (message) with a given Tag (direct relation, not transitive). Considering only these Messages, for each Person node:  • Count its messages (messageCount). • Count likes (likeCount) to its messages. • Count Comments (replyCount) in reply to it messages.  The score is calculated according to the following formula: 1 × messageCount + 2 × replyCount + 10 × likeCount.			
BI 20	params	1 tag Long String Tags with a similar amount of Messages are selected			
	result	1 person.id ID R 2 replyCount 32-bit Integer A 3 likeCount 32-bit Integer A 4 messageCount 32-bit Integer A 5 score 32-bit Integer A			
	sort	1 score ↓ 2 person.id ↑			
	limit	100			
	CPs	1.2, 2.3, 8.2			

BI 1	query	BI / read / 6				
BI 2	title	Most authoritative users on a given topic				
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11	pattern	Tag person: Person id p2: Person hasCreator message1:Message message2:Message message2:Messa				
BI 12 BI 13 BI 14 BI 15 BI 16 BI 17 BI 18 BI 19	desc.	<ul> <li>Given a Tag (tag), find all Persons (person) that ever created a Message with the Tag. For each of these Persons (person) compute their "authority score" as follows:</li> <li>The "authority score" is the sum of "popularity scores" of the Persons (p2) that liked any of that Person's Messages with the given Tag (same criterion as for message1).</li> <li>A Person's (p2) "popularity score" is defined as the total number of likes on all of their Messages (message2).</li> </ul>				
BI 20	params	1 tag Long String Tags with a similar amount of Messages are selected				
	result	1 person.id ID R 2 authorityScore 32-bit Integer A				
	sort	1 authorityScore ↓ 2 person1.id ↑				
	limit	100				
	CPs	1.2, 2.3, 3.3, 6.1, 8.2				
	relevance	Computing the authority scores might involve computing the popularity score for the same Person multiple times. Implementations are advised to avoid such redundant computations.				

BI 1	query	BI / read / 7								
BI 2	title	Related topics								
BI 3										
BI 4		tag: Tag								
BI 5	nattorn	name = \$tag								
BI 6	pattern	hasTag count name								
BI 7		Message comment: Comment								
BI 8										
BI 9		Find all Messages that have a given Tag. Find the related Tags attached to (direct) reply Comments								
BI 10	desc.	of these Messages, but only of those reply Comments that do not have the given Tag.								
BI 11		Group the Tags by name, and get the count of replies in each group.								
BI 12	narame	1 tag Long String Tags with a similar amount of Messages are selected								
BI 13	params	Tago with a similar amount of messages are selected								
BI 14		1 relatedTag.name Long String R								
BI 15	result	2 count 32-bit Integer A								
BI 16		22 Count 32-bit integer A								
BI 17		1 count								
BI 18	sort									
BI 19		relatedTag.name								
BI 20	limit	100								
	CPs	1.4, 3.3, 5.2, 8.1								

BI 1	query	BI / read / 8				
BI 2	title	Central person for a tag				
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12 BI 13	pattern	For each person with a matching hasInterest and/or hasCreator edge, compute person.score = (if hasInterest edge exists then 100 else 0) + count(message)  Tag				
BI 14 BI 15 BI 16 BI 17 BI 18 BI 19 BI 20	desc.	Given a Tag, find all Persons that are interested in the Tag and/or have written a Message (Post or Comment) with a creationDate after a given date and that has a given Tag. For each Person, compute the score as the sum of the following two aspects:  • 100, if the Person has this Tag as their interest, or 0 otherwise  • number of Messages by this Person with the given Tag  Also, for each Person, compute the sum of the score of the Person's friends (friendsScore).				
	params	Dates from around the same day are selected. (TODO - how exactly? what distribution?)				
	result	1 person.id ID R 2 score 32-bit Integer A 3 friendsScore 32-bit Integer A The sum of the score of the person's friends				
	sort	1 score + friendsScore ↓ 2 person.id ↑				
	limit	100				
	CPs	1.2, 2.1, 2.3, 3.2, 5.3, 8.2, 8.4, 8.5				
	relevance	Similarly to BI 16, there are two major ways to compute this query: (1) creating an induced subgraph of the interested Persons and their friends and performing the scoring on this graph or (2) performing the scoring without creating an induced subgraph and scoring the friends of a Person on-the-fly. The first approach is more efficient as it avoids redundant computations, however, specifying it needs support for composable graph queries.				

BI 1	query	BI / read / 9					
BI 2	title	Top thread initiators					
BI 3 BI 4 BI 5 BI 6 BI 7	pattern	threadCount = count  Person  AbaSCreator  Post  StartDate ≤ creationDate and creationDate and creationDate ≤ \$endDate  \$startDate ≤ \$endDate					
BI 8 BI 9 BI 10 BI 11 BI 12 BI 13	desc.	For each Person, count the number of Posts they created in the time interval [startDate, endDate] (equivalent to the number of threads they initiated) and the number of Messages in each of their (transitive) reply trees, including the root Post of each tree. When calculating Message counts only consider Messages created within the given time interval.  Return each Person, number of Posts they created, and the count of all Messages that appeared in the reply trees (including the Post at the root of tree) they created.					
BI 14 BI 15 BI 16	params	1 startDate Date TODO 2 endDate Date 8-10 days after the startDate					
BI 17 BI 18 BI 19 BI 20	result	1 person.id ID R 2 person.firstName String R 3 person.lastName String R 4 threadCount 32-bit Integer A The number of Posts created by that Person (the number of threads initiated) 5 messageCount 32-bit Integer A The number of Messages created in all the threads this Person initiated					
	sort	1 messageCount ↓ 2 person.id ↑					
	limit CPs	100 1.2, 2.2, 2.3, 3.2, 7.2, 7.3, 7.4, 8.1, 8.5					

BI 1	query	BI / read / 10									
BI 2	title	Experts in social circle									
BI 3					Count	rv					
BI 4 BI 5					name = \$country						
BI 6				L	<b>↑</b> i	sPartOt	f				
BI 7					City	,					
BI 8				_	<b>↑</b> i	sLocate	edIn				
BI 9	pattern	startPerson: Person	knows*	ance	expertCandidateF	Person:	Person		Та	agClass	ı
BI 10		id = \$personId	\$maxPathDis	tance	id				name = \$	tagClass	
BI 11					<b>^</b>	hasCre	ator			hasTuna	
BI 12					count for e	ach (ta	g, person)			hasType	
BI 13		tag: Tag	← hasTag	-	Messa	ge		hasTag	>	Tag	
BI 14		name						J			
BI 15 BI 16		Given a Person (sta	rtPerson), fir	nd all	other Perso	ns (e	expert(	CandidatePer	son) that	t live in a	given
BI 17		Country and are con	nected to giv	en Pe	rson by a s	hort	est pai	th with leng	th in ran	ge [minPat	thDis-
BI 18		tance, maxPathDista	ance] through	n the k	nows relation	on.					
BI 19	desc.	For each of these ex							_		
BI 20	ucse.	one Tag belonging t	o a given Tag	gClass	(direct rela	ation	not tr	ransitive). F	or each I	Message, re	trieve
		all of its Tags.	_ 1.		1 .			1	_	1 .	
		Group the results by	Persons and	Tags, t	hen count	the M	Message	es by a certai	n Person	having a c	ertain
-		Tag.									
		1 personId ID			The ID of the startPerson. Persons with a similar						r
		2 country String			degree of knows edges are selected						
					Countries with a similar number of Persons are						
	params				selected						
	params	3 tagClass	Long St	ring	TagClasses with a similar degree of hasType edges are selected						are
			22 11 1								
		4 minPathDistar			1 or 2						
		5 maxPathDistar	nce 32-bit Ir	nteger	2 or 3						
		1 expertCandida	atePerson.id	ID		R					
		2 tag.name			String	R					
	result						Num	ber of Messa	ges create	ed by that	
		3 messageCount		32-bi	it Integer	Α		n containing	_	•	
-											
		1 messageCount		<b>↓</b>							
	sort	2 tag.name		1							
		3 expertCandida	atePerson.id	1							
	limit	100									
	CPs		3. 5.3. 7 1 7	2.73	. 8.1. 8.6						
	C1 3	1.2, 1.3, 2.3, 2.4, 3.3, 5.3, 7.1, 7.2, 7.3, 8.1, 8.6									

BI 1	query	BI / read / 11
BI 2	title	Friend triangles
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12	pattern	Country  name = \$country  isPartOf  isPartOf  isPartOf  City  City  City  City  City  City  isLocatedIn  isLocatedIn  isLocatedIn  knows  \$startDate < knows.creationDate  knows.creationDate  c: Person  knows.creationDate  c: Person
BI 13 BI 14 BI 15 BI 16 BI 17 BI 18 BI 19 BI 20	desc.	For a given country, count all the distinct triples of Persons such that:  • a is friend of b,  • b is friend of c,  • c is friend of a,  and these friendships were created after a given startDate.  Distinct means that given a triple $t_1$ in the result set $R$ of all qualified triples, there is no triple $t_2$ in $R$ such that $t_1$ and $t_2$ have the same set of elements.
	params	1 country Long String 2 startDate Date
	result	1 count 32-bit Integer A
	CPs	1.1, 2.3, 2.5

BI 1	query	BI / read / 12			
BI 2	title	How many persons have a given number of messages			
BI 3 BI 4 BI 5 BI 6 BI 7	pattern	2. personCount = count  hasCreator  Message  count Persons grouped by messageCount value  1. messageCount = count  content not empty and length < \$lengthThreshold and \$date < creationDate  1. messageCount = count  Acopt> Fost  language in \$languages			
BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14 BI 15 BI 16 BI 17 BI 18 BI 19 BI 20	For each Person, count the number of Messages they made (messageCount). Only count with the following attributes:  • Its content is not empty (and consequently, the imageFile attribute is empty for • Its length is below the lengthThreshold (exclusive, equality is not allowed).  • Its creationDate is after date (exclusive, equality is not allowed).  • It is written in any of the given languages.  — The language of a Post is defined by its language attribute.  — The language of a Comment is that of the Post that initiates the thread where ment replies to.  The Post and Comments in the reply tree's path (from the Message to the Post) do resatisfy the constraints for content, length and creationDate.  For each messageCount value, count the number of Persons with exactly messageCount (with the required attributes).				
		1 date Date			
	params	Selected as balanced against date to filter around 30% of the Messages within a language and keep the variance low			
		Only the most frequently used languages are selected			
	result	1 messageCount 32-bit Integer A Number of Messages created 2 personCount 32-bit Integer A Number of Persons with messageCount Messages			
	sort	1 personCount ↓ 2 messageCount ↓			
	CPs	1.1, 1.2, 1.4, 3.2, 4.2, 4.3, 8.1, 8.2, 8.3, 8.4, 8.5			

BI 1	query	BI / read / 13
BI 2	title	Zombies in a country
BI 3	LILIE	Zonibles in a country
BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14 BI 15 BI 16 BI 17 BI 18	pattern	Country  name = \$country  isPartOf  City  zombie: Person  creationDate < \$endDate and (messageCount / months) < 1  2. For each zombie IN zombies, calculate: zombieScore = zombieLikeCount / totalLikeCount  zombie: Person  totalLikeCount = count(likerPerson)  likes  Message  creationDate < \$endDate  and likerZombie: Person  creationDate < \$endDate  likes  Message  likerZombie: Person  creationDate < \$endDate and likerZombie: Person  creationDate < \$endDate and likerZombie IN zombies
Find zombies within the given created before the given endDatd during the time range between months spans the time range from months on both end counting as 1 result in 3 months).  For each zombie, calculate the form total LikeCount: the num total LikeCount: the total result in 2 combieScore: zombieLikeCount and total LikeCount and LikeCount		Find zombies within the given country, and return their zombie scores. A zombie is a Person created before the given endDate, which has created an average of [0, 1) Messages per month, during the time range between profile's creationDate and the given endDate. The number of months spans the time range from the creationDate of the profile to the endDate with partial months on both end counting as one month (e.g. a creationDate of Jan 31 and an endDate of Mar 1 result in 3 months).  For each zombie, calculate the following:  • zombieLikeCount: the number of likes received from other zombies.  • totalLikeCount: the total number of likes received.  • zombieScore: zombieLikeCount / totalLikeCount. If the value of totalLikeCount is 0, the zombieScore of the zombie should be 0.0.  For both zombieLikeCount and totalLikeCount, only consider likes received from profiles that were created before the given endDate.
	params	1 country Long String Only the largest Countries are selected 2 endDate Date Selected from the last days of the initial data set
	result	1 zombie.id ID R 2 zombieLikeCount 32-bit Integer A 3 totalLikeCount 32-bit Integer A 4 zombieScore 64-bit Float A Determined as zombieLikeCount / totalLikeCount
	sort	1 zombieScore ↓ 2 zombie.id ↑
	limit	100
	CPs	1.2, 2.1, 2.3, 2.4, 3.2, 3.3, 4.2, 5.1, 5.3, 8.2, 8.4, 8.5

BI 1		DI / mood / 14
BI 2	query	BI / read / 14
	title	International dialog
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14	pattern	For each pair of countries, calculate the cost as a sum of cases #1-5. Cases that have a match add to the final score with the specified value. Each case only counts once, multiple matches do not increase to the score.  Country
BI 15 BI 16 BI 17 BI 18 BI 19		Case 3: score += 15  person1: Person knows person2: Person likes  Message hasCreator  Case 4: score += 10 person1: Person person2: Person likes hasCreator  Message
BI 20	desc.	Consider all pairs of people (person1, person2) such that one is located in a City of Country country1 and the other is located in a City of Country country2. For each City of Country country1, return the highest scoring pair. The score of a pair is defined as the sum of the subscores awarded for the following kinds of interaction. The initial value is score = 0.  1. person1 has created a reply Comment to at least one Message by person2: score += 4 2. person1 has created at least one Message that person2 has created a reply to: score += 1 3. person1 and person2 know each other: score += 15 4. person1 liked at least one Message by person2: score += 10 5. person1 has created at least one Message that was liked by person2: score += 1  Consequently, the maximum score a pair can obtain is: 4 + 1 + 15 + 10 + 1 = 31.  This query has two variants based on whether the input parameters are selected as correlated (close countries) or uncorrelated (far countries).
	params	A: correlated with parameter country2, i.e. the countries are close and there are many Persons visiting both Countries.  B: uncorrelated with parameter country2, i.e. the countries are afar and there are few Persons visiting both Countries.  Country2 Long String
	result	1 person1.id ID R 2 person2.id ID R 3 city1.name Long String R 4 score 32-bit Integer C
	sort	1 score ↓ 2 person1.id ↑ 3 person2.id ↑
	CPs	1.3, 1.4, 2.1, 3.1, 3.3, 5.1, 5.2, 5.3, 8.3, 8.4

BI 1	query	BI / read / 15
BI 2	title	Trusted connection paths through forums created in a given timeframe
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14 BI 15 BI 16 BI 17	pattern	Enumerate all unweighted shortest paths on knows edges between person1 to person2.    person1: Person
BI 18 BI 19 BI 20	desc.	Given two Persons, find all (unweighted) shortest paths between these two Persons, in the subgraph induced by the knows relationship.  Then, for each path calculate a weight. The nodes in the path are Persons, and the weight of a path is the sum of weights between every pair of consecutive Person nodes in the path.  The weight for a pair of Persons is calculated based on their interactions:  • Every direct reply (by one of the Persons) to a Post (by the other Person) contributes 1.0.  • Every direct reply (by one of the Persons) to a Comment (by the other Person) contributes 0.5.  Only consider Messages that were created in a Forum that was created within the timeframe (interval) [startDate, endDate]. Note that for Comments, the containing Forum is that of the Post that the comment (transitively) replies to. Also note that interactions are counted both ways.  Return all paths with the Person IDs ordered by their weights descending.
	params	1 person1Id ID 2 person2Id ID 3 startDate Date 4 endDate Date
	result	1 person.id [ID] C Ordered sequence of the Person IDs in the path 2 weight 64-bit Float C
	sort	1 weight ↓ The order of paths with the same weight is unspecified 2 personIds ↑ The IDs in the paths are used for lexicographical sorting
	CPs	1.2, 2.1, 2.2, 2.4, 3.3, 5.1, 5.3, 7.2, 7.3, 7.5, 7.7, 8.1, 8.2, 8.3, 8.4, 8.5, 8.6

BI 1	query	BI / read / 16
BI 2	title	Fake news detection
BI 3		For \$tagX/\$dayX in [tagA/dateA, tagB/dateB], compute scoreX = count(messageX)
BI 5		Create an induced subgraph of Persons who created a Message with Tag \$tagX on \$dateX
BI 6		tag: Tag Message hasCreator person: Person
BI 7		name = \$tagX   day(creationDate) = \$dateX
BI 8		2. In the subgraph, count the Messages (using the same conditions) from People with ≤ \$maxKnowsLimit friends
BI 9 BI 10	pattern	count(messageX)
BI 11		tag: Tag messageX: Message hasCreator hasCreator hasCreator
BI 12		count ≤ \$maxKnowsLimit «opt» knows
BI 13		Person
BI 14		
BI 15 BI 16		Given two Tag/date pairs (tagA/dateA and tagB/dateB), for each pair tagX/dateX:
BI 17		
BI 18		• Create an induced subgraph between Persons where for each pair of Persons person1/person2, both have created a Message on the day of dateX with Tag tagX.
BI 19		• In the induced subgraph, only keep pairs of Persons who have at most maxKnowsLimit friends
BI 20	desc.	(in the induced subgraph).
		• For these Persons, count the number of Messages created on dateX with Tag tagX.
		Return Persons who had at least one Messages for both tagA/dateA and tagB/dateB ranked by their
		total number of Messages (descending).
		1 tagA Long String
		2 dateA Date
	params	3 tagB Long String
		4 dateB Date
		5 maxKnowsLimit 32-bit Integer Selected between 3 and 6
		1 person.id ID R
	l+	2 messageCountA 32-bit Integer A Message count for tagA/dateA
	result	3 messageCountB 32-bit Integer A Message count for tagB/dateB
		or an integer with the stage countries tagged and the stage and the stag
		messageCountA +
	sort	messageCountB '
		2 person.id ↑
	limit	20
	CPs	5.3, 8.4, 8.5
		There are two major ways to compute this query: (1) create the induced subgraph as suggested by the specification (either as a view or in materialized form), or (2) skip greating the induced subgraph and perform on the fly check.
	relevance	(either as a view or in materialized form), or (2) skip creating the induced subgraph and perform on-the-fly check for the number of friends (who also posted at least one Message with the given Tag on the given date). The latter
		approach is easier to express in systems which do not provide graph views but might result in redundant computations
		(the query engine will might repeatedly check whether a Person has at least one Message that satisfies the conditions).

BI 1	query	BI / read / 17
BI 2	title	Information propagation analysis
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14 BI 15 BI 16	pattern	person1: Person id  hasCreator  message1: Message  creationDate  replyOf*0  replyOf*0  post1: Post  name = \$tag  hasTag  message2: Message  message1.creationDate  forum1: Forum  hasMember  hasMember  hasMember  hasMember  hasCreator  hasCreator  hasCreator  containerOf  message2: Message  message1.creationDate  + \$delta < creationDate  replyOf  comment: Comment
BI 17 BI 18 BI 19 BI 20	desc.	This query aims to identify instances of "information propagation" when a Person (person1) submits a Message (message1) with a given Tag (tag) to a Forum (forum1). This is read by other members of forum1, Persons person2 and person3. Some time later (specified by the delta parameter), these persons have a discussion with the same tag in a different Forum (forum2) where person1 is not a member. The discussion consists of a Message (message2) by person2 and a direct reply Comment (comment) by person3.  Return IDs of person1 with the number of interactions their Messages (might have) caused.
	params	1 tag Long String Tags with a similar amount of Messages are selected 2 delta 32-bit Integer Measured in hours, selected to be between 8 and 16 hours.
	result	1 person1.id ID R 2 messageCount 32-bit Integer A
	sort	1 person1.id ↑
	CPs	2.1, 2.3, 8.1

BI 1	query	BI / read / 18
BI 2	title	Friend recommendation
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11	pattern	For each person1 compute top-k(person2) based on mutualFriendCount  tag: Tag  name = \$tag  hasInterest  person1: Person  id = \$person1Id  «neg» knows
BI 12 BI 13 BI 14 BI 15 BI 16 BI 17 BI 18	desc.	For a given Person (person1) and a Tag (tag), recommend new friends (person2) who  • do not yet know person1  • have many mutual friends with person1  • are interested in tag.  Rank Persons person2 based on the number of mutual friends.
BI 19 BI 20	params	person1Id ID Persons with a similar amount of friends are selected Long String Tags with a similar amount of Messages are selected
	result	1 person2.id ID R 2 mutualFriendCount 32-bit Integer A
	sort	<pre>1 mutualFriendCount ↓ 2 person2.id ↑</pre>
	limit	20
	CPs	2.5, 8.1

BI 1	query	BI / read / 19
BI 2	title	Interaction path between cities
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9 BI 10 BI 11 BI 12 BI 13 BI 14 BI 15 BI 16 BI 17 BI 18 BI 19 BI 20	pattern	Find the shortest paths between all pairs of Persons in city1 and city2  city1: City  id = \$city1id  isLocatedIn  shortest paths on  person1: Person  knows.weight  person2: Person  Case i1: Reply from personA to Person B's Message  personA: Person  hasCreator  c: Comment  replyOf  m: Message  The weight of a knows edge is based on the number of interactions between its Persons: knows.weight = 1 / (count(i1)+count(i2))  p1 knows pX knows pY  knows pX knows pY  m: Message  replyOf  c: Comment
	desc.	Given two Cities city1, city2, find Persons person1, person2 living in these Cities (respectively) with the shortest <i>interaction path</i> between them. If there are multiple pairs of people with shortest paths having the same total weight, return all of them.  The shortest path is computed using a weight between two Persons defined as the reciprocal of the number of interactions (direct reply Comments to a Message by the other Person). Therefore, more interactions imply a smaller weight. <i>Note:</i> Interactions are counted both ways, i.e. if Alice writes 2 reply Comments to Bob's Messages and Bob writes 3 reply Comments to Alice's Messages, their total number of interactions is 5.
	params	1 city1Id ID Small Cities within the same Country are selected 2 city2Id ID
	result	1 person1.id ID R 2 person2.id ID R 3 totalWeight 64-bit Float C
	sort	1 totalWeight ↓ 2 person1.id ↑ 3 person2.id ↑
	limit	20
	CPs	3.3, 7.6, 7.7, 8.4, 8.6
	relevance	Finding shortest paths between pairs of Persons in Cities can be implemented in theory with an <i>all-pairs shortest paths</i> algorithm. However, this needs to be executed on the whole Person-knows-Person graph (with edge weights derived from the number of interactions) so it is expected to be prohibitively expensive. A better approach is using multiple <i>single-source shortest path algorithms</i> (e.g. from the City with fewer inhibitants).

BI 1	query	BI / read / 20
BI 2	title	Recruitment
BI 3 BI 4 BI 5 BI 6 BI 7 BI 8 BI 9	pattern	company: Company  name = \$company  workAt  person1: Person  shortest path on knows.weight  id = \$person2!d
BI 10 BI 11 BI 12 BI 13 BI 14 BI 15	desc.	Given a Company company and a Person person2 (who is known to be working at another Company), find a Person (person1) working the in the company who have the top-20 shortest path to person2 through people who have studied together. On this path, we only consider edges between Persons who know each other and attended the same university and set the weight of the edge to the absolute difference between the year of enrolment plus 1 (studyAt.classYear + 1). If there are multiple Person person1 nodes with the same shortest path, return all of them.
BI 16 BI 17 BI 18 BI 19	params	1 company Long String Companies with a similar number of employees (former or current) are selected 2 person2Id ID
BI 20	result	1 person1.id ID R 2 totalWeight 64-bit Integer C
	sort	1 person1.id ↑
	limit	20
	CPs	3.3, 7.6, 7.7, 8.4, 8.6
	relevance	Implementations can either pre-compute edge weights or compute them on-the-fly. To find a weighted shortest path efficiently, implementations can use e.g. a bidirectional Dijkstra algorithm.