Privacy threat analysis

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Objectives

- Understanding the nature of privacy requirements and their relationship with antirequirements
- Method to elicit privacy anti-requirements (LINDDUN)
- Documenting privacy threats

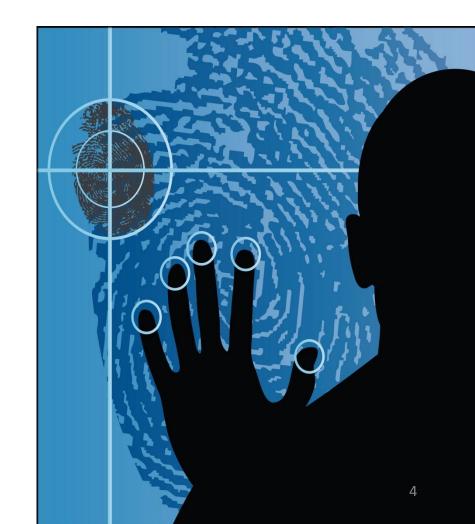
Overview

- - Privacy
 - What?
 - Properties
 - Privacy methodology
 - Example case study
 - Project information

Privacy

- what is privacy?
 - Confidentiality
 - Data minimization
 - User empowerment

— ...



What is privacy?

- The right to be let alone (Warren & Brandeis, 1890)
- The right of the individual to decide what information about himself should be communicated to others and under what circumstances (Westin, 1970)
- Freedom from unreasonable constraints on the construction of one's own identity (Agre & Rotenberg, 2001)

People don't care about online privacy?

- In the "real world": concerned about information we share
 - Who they tell what
 - You might be willing to tell your best friend that you had an argument with your girlfriend, but you don't want everybody to know about it
 - Concerns over information taken out of context
 - A picture taken at a crazy party being available to a potential employer
 - We value friends who are discreet and keep our secrets
 - We give more information to people we trust
 - The cost of gathering and analyzing information without advanced technologies has guaranteed that we had a rather high level of privacy protection

People don't care about online privacy?

- Online:
 - less concerned or unaware of privacy violations
- This information is not necessarily secret, but would you want to broadcast it?
 - Identity attributes (Name, age, gender, race, IQ, marital status, place of birth, address, phone number, ID number...)
 - Location (Where you are at a certain point in time, movement patterns)
 - Interests / preferences (Books you read, music you listen, films you like, sports you practice, political affiliation, religious beliefs, sexual orientation)
 - Behavior (Personality type, what you eat, what you shop, how you behave and interact with others)
 - Social network (Who your friends are, who you meet when, your different social circles)
 - Health data (Medical issues, treatments you follow, DNA, health risk factors)
 - Financial data (How much you earn, how you spend your money, credit card number, bank account)
- Combination of them all is even more troublesome

Privacy problems

- Identity theft
 - Getting a credit card on your name
- Stalking
- Profiling
 - Find compulsive buyers, ...
- Sensitive information being shared
- Information taken out of context

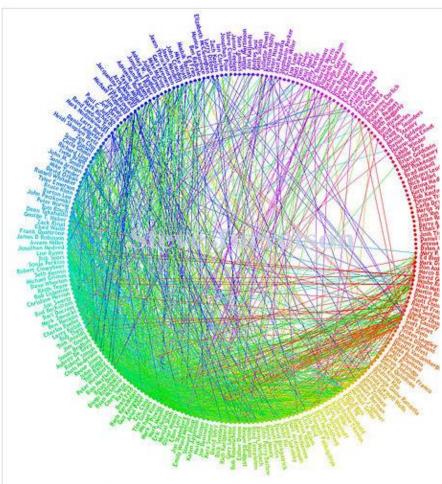


Signs of the social networking times.

Gaydar Algorithm Outs Facebook Users

By Susannah F, Locke Posted 09.21.2009 at 12:27 pm 📃 12 Comments





What are your friends saying about you? Online social networks like this Facebook one mi sharing things you'd rather keep to yourself. Even if more about you than you think jurvetson (CC licensed) you're only declaring to the world that someone's you

A pair of MIT students claim that they have created an algorithm that outs gay members of Facebook by analyzing the sexual orientations of their networks of friends.

The students first analyzed the networks of people who publicized their sexual orientation on Facebook. Turns out that statistically speaking, gay men have more gay friends than straight guys do. So then, they used an algorithm to run the stats on men who kept mum about their sexual orientation on the site. Their computer program was able to correctly identify 10 men whom the students personally knew to be gay in the real world but who hadn't shared that fact on Facebook. (The algorithm didn't work as well with women or with bisexual Facebookers.)

The students completed the project for a class on ethics and the Internet and hope to publish it in a scientific journal.

Their project is far from the first study showing that a simple computer program can sleuth out details you might prefer to keep private by looking at your social network on the Internet. Earlier this year, computer scientists correctly linked 30 percent of anonymous Twitter and Flickr accounts with a simple algorithm that compares who's following who on each site. And other researchers have used Internet social networks to correctly identify peoples' political affiliations or where they live.

It's a good reminder to take a look at your privacy settings. Because you might inadvertently be sharing things you'd rather keep to yourself. Even if you're only declaring to the world that someone's your friend.

'I'm going to destroy America and dig up Marilyn Monroe': British pair arrested in U.S. on terror charges over Twitter jokes

By RICHARD HARTLEY-PARKINSON

Last updated at 1:08 PM on 31st January 2012



Two British tourists were barred from entering America after joking on Twitter that they were going to 'destroy America' and 'dig up Marilyn Monroe'.

Leigh Van Bryan, 26, was handcuffed and kept under armed guard in a cell with Mexican drug dealers for 12 hours after landing in Los Angeles with pal Emily Bunting.

The Department of Homeland Security flagged him as a potential threat when he posted an excited tweet to his pals about his forthcoming trip to Hollywood which read: 'Free this week, for quick gossip/prep before I go and destroy America?'





Spear-phishing

Using personal information to make phishing more successful



Dear 'cs.kuleuven.be' E-mail User,

We are currently upgrading our database and all account need to be verified. To complete your account activation with us, you are required to reply to this message and enter your password in the space provided (*******) you are required to do this before the next 48 hours of the receipt of this email or your database will be de-activated from our database.

Full Name:

username:

Password:

Thank you for using cs.kuleuven.be

Copyright 2012 © cs.kuleuven.be web Team.

Using Facebook data



Freddi Staur

- 41% agreed to be friends with Freddi which (often) led to access to
 - Email address
 - Full date of birth
 - Details on education and workplace
 - Current address
 - Pictures of family and friends
 - Name of their partner / relatives



Privacy properties

- Unlinkability
- Anonymity/ pseudonymity
- Plausible deniability
- Undetectability
- Confidentiality
- Content awareness
- Policy and consent compliance

Hard privacy

Security

Soft privacy

Hard privacy

- Data minimization
 - Subject provides as little data as possible
 - Reduce as much as possible the need to "trust" other entities



Soft privacy

- Data subject has already lost control of her data
 - In practice, very difficult for data subject to verify how her data are collected and processed



Soft privacy

Need to trust data controllers (honesty, competence)



Privacy

Anonymity & Pseudonymity

Anonymity

An attacker cannot sufficiently identify the subject within a set of subjects, the anonymity set



Anonymity

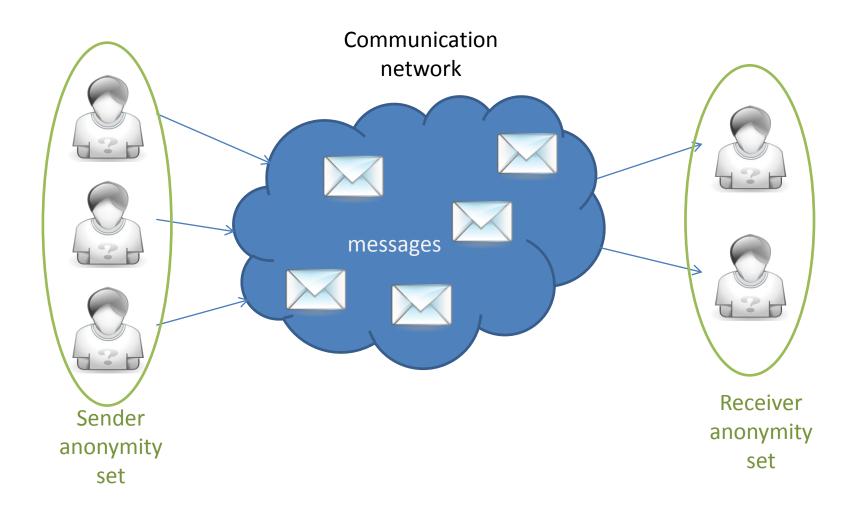
- An attacker cannot sufficiently identify the subject within a set of subjects, the anonymity set (Pfitzmann)
- Hiding link between identity and action / piece of information.

Examples:

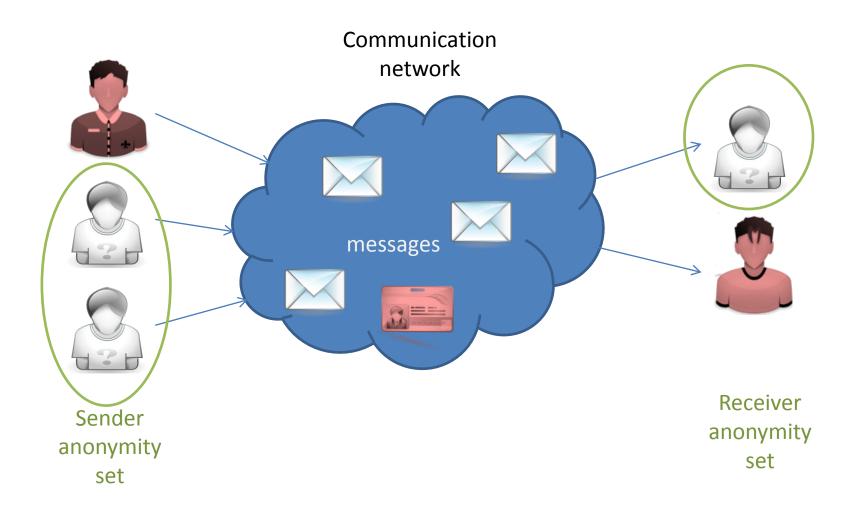
- Reader of a web page, person accessing a service
- Sender of an email, writer of a text
- Person to whom an entry in a database relates
- Person present in a physical location



Anonymity set



Anonymity set wrt attacker



Identifiability

- The attacker can sufficiently identify the subject within a set of subjects, the identifiability set (pfitzmann)
- A identity is any subset of attribute values of an individual personal which sufficiently identifies this individual person with any set of persons.
 - There can thus be many "identities"

Identifiability example

Browser uniqueness



Your browser fingerprint appears to be unique among the 2,123,272 tested so far.

Currently, we estimate that your browser has a fingerprint that conveys at least 21.02 bits of identifying information.

The measurements we used to obtain this result are listed below. You can read more about our methodology, statistical results, and some defenses against fingerprinting in this article.

http://panopticlick.eff.org/

Possible to track "anonymous" visitors

Pseudonymity

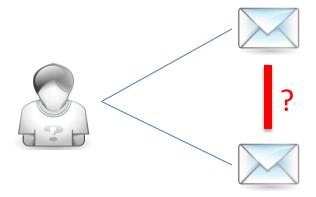
- A pseudonym is an identifier of a subject other than one of the subjects real names.
 Pseudonymity is the use of pseudonyms as identifiers. (Pfitzmann)
- Pseudonymity is the entire field between anonymity and identifiability

Privacy

Anonymity & Pseudonymity Unlinkability

Unlinkability

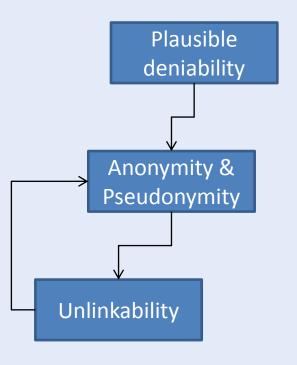
Within a system, the attacker cannot sufficiently distinguish whether two or more items of interest (IOI) are related or not



Unlinkability

- Within a system, the attacker cannot sufficiently distinguish whether two or more items of interest (IOI) are related or not (Pfitzman)
- Hiding link between two or more actions / identities /pieces of information
- Examples:
 - Two anonymous letters written by the same person
 - Two web page visits by the same user
 - Entries in two databases related to the same person
 - Two people related by a friendship link
 - Same person spotted in two locations at different points in time

Privacy



Plausible deniability

Not possible to prove user knows, has done or has said something

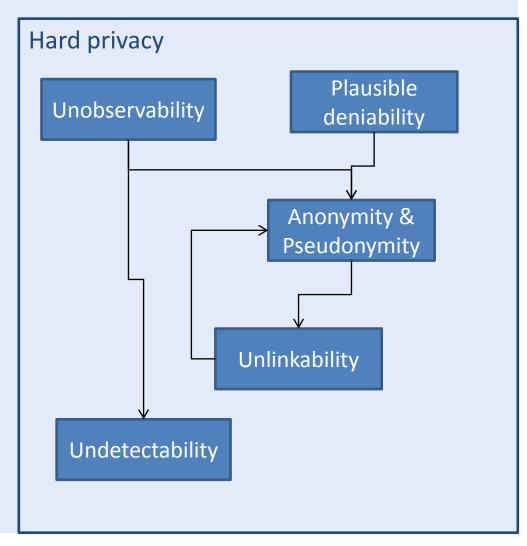




Plausible deniability

- Not possible to prove user knows, has done or has said something
- Examples:
 - Resistance to coercion:
 - Not possible to prove that a person has hidden information in a computer
 - Not possible to know that someone has the combination of a safe
 - Possibility to deny having been in a place at a certain point in time
 - Possibility to deny that a database record belongs to a person
 - Off-the-record conversations

Privacy



→ requires

Undetectability

The attacker cannot sufficiently distinguish whether it exists or not



Unobservability

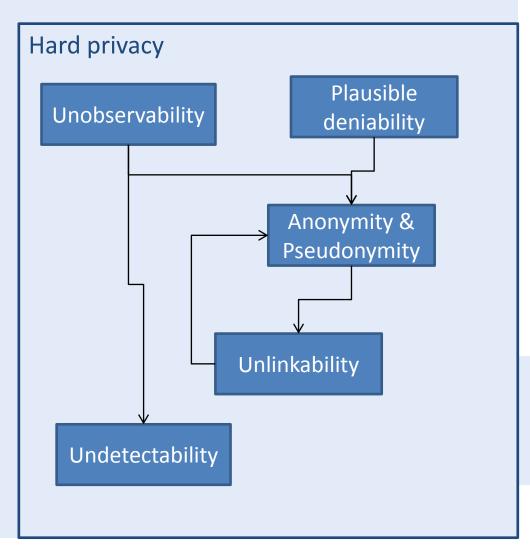
undetectability + anonymity of subjects involved in the IOI even against the other subjects involved in that IOI

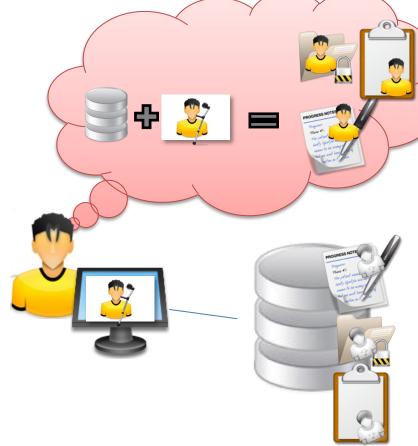


Undetectability & Unobservability

- Undetectability: The attacker cannot sufficiently distinguish whether it exists or not (Pfitzmann)
- Unobservability: undetectability + anonymity of subjects involved in the IOI even against the other subjects involved in that IOI (Pfitzmann)
- Hiding user activity
- Examples:
 - Impossible to see whether someone is accessing a web page
 - Impossible to know whether an entry in a database corresponds to a real person
 - Impossible to distinguish whether someone or no one is in a given location

Privacy





User awareness

User awareness

Users are aware of the consequences of sharing information 30

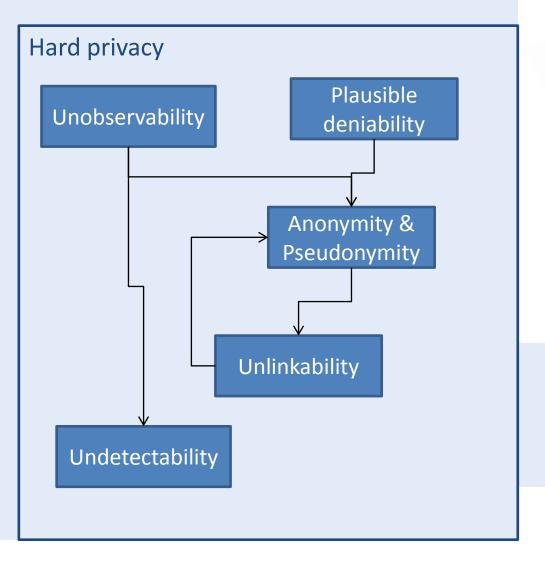
Content Awareness

 Users should be made aware of the consequences of sharing information

 Suggested solution: Feedback & awareness tools



Privacy



Directive 95/46/EC of the Furonce of Directive on Directive on Directive on Directive of the Furonce of Directive 95/46/EC of the Furonce of Directive 95/46/EC

Directive 95/46/EC of the European Parliament and of the Council of 24 October 1995 on the protection of individuals with regard to the processing of personal data and on the free movement of such data



User awareness

Compliance

32

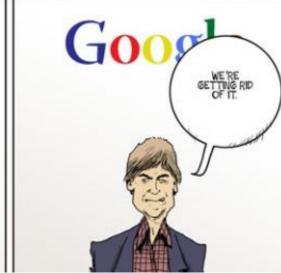
Compliance

Legal compliance is obligated. e.g. consents

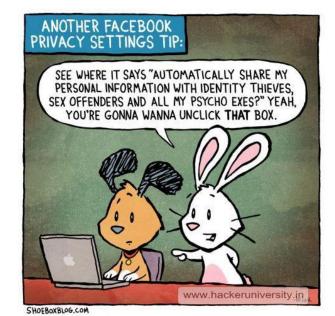
Policy & Consent compliance

- Policies
 - Coorporate
 - Privacy





- Openness to users
 - + control



Policy & Consent compliance

- Legal compliance is obligated
 - E.g. European Data Protection Directive
 - fair and lawful processing
 - Consent
 - purpose specification
 - minimality
 - minimal disclosure

- information quality
- data subject control
- sensitivity
- information security



Compliance example: User consent

- *personal data* = any information relating to an identified or identifiable natural person ("data subject")
- Sensitive data = personal data revealing racial or ethnic origin, political opinions, religious or philosophical beliefs, trade-union membership, and the processing of data concerning health or sex life
- Processing of sensitive data prohibited unless
 - the processing is necessary for the protection of the vital interests of the data subject,
 - the processing is necessary for purposes of preventive medicine, medical diagnosis, provision of care or treatment or
 - the data subject has given his explicit, written consent to the processing of the data
 - ...(art. 7)

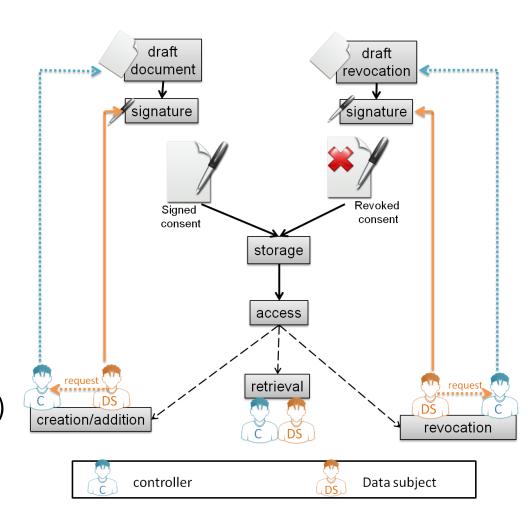
User consent

Legal requirements

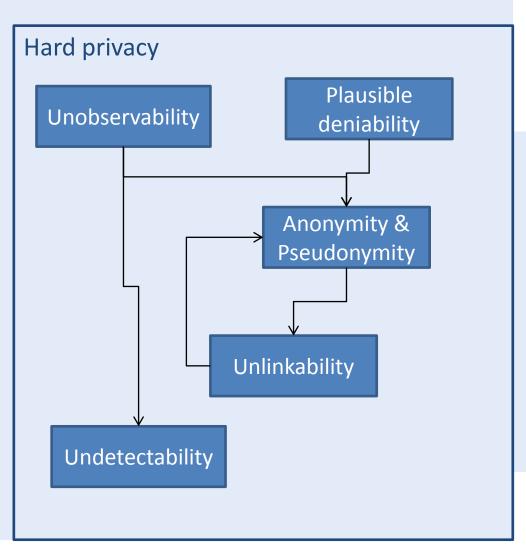
- Informed
- Freely given
- Specific

Consent structure

- Data subject
- Controller
- Receiver
- Types of data
- Action (Upload or share)
- Purpose of sharing
- •Type of consent (opt-in/opt-out)
- Revoked
- •(Context (e.g., "emergency"))
- (Location)



Privacy





Confidentiality

User awareness

Compliance

Confidentiality

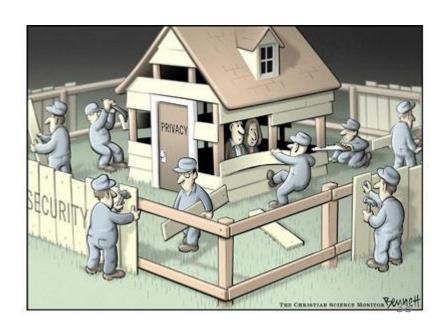
authorized restrictions on information access and disclosure,7

----> requires

Confidentiality

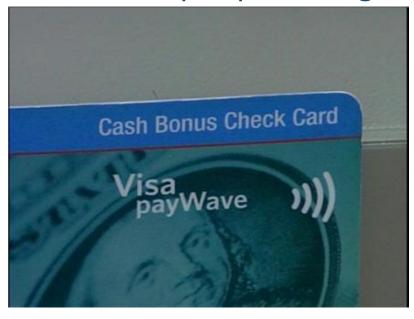
 Preserving authorized restrictions on information access and disclosure, including means for protecting personal privacy and proprietary information (NIST)

Security property



Confidentiality: example

Problem: Electronic pickpocketing

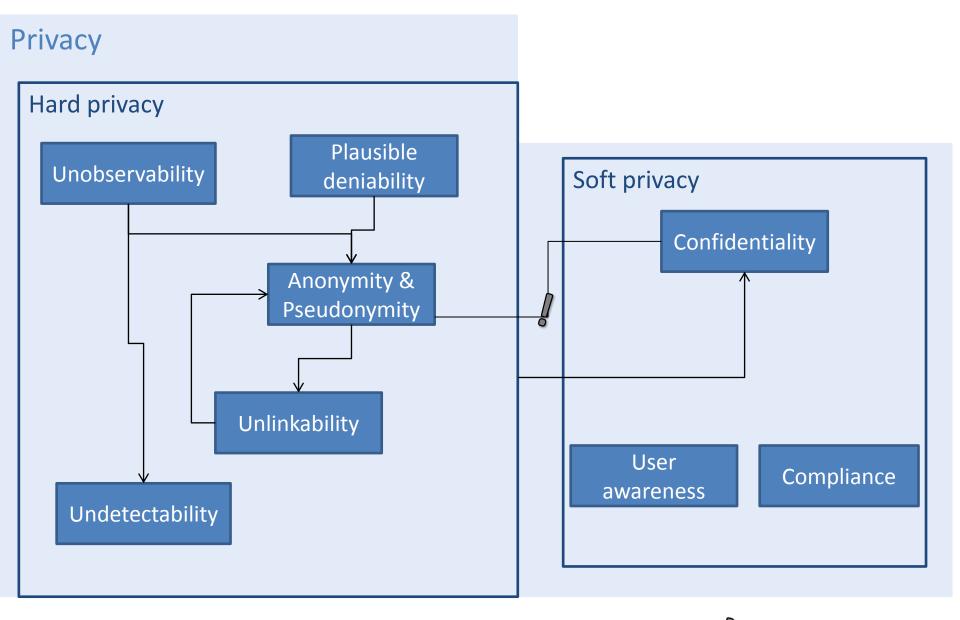


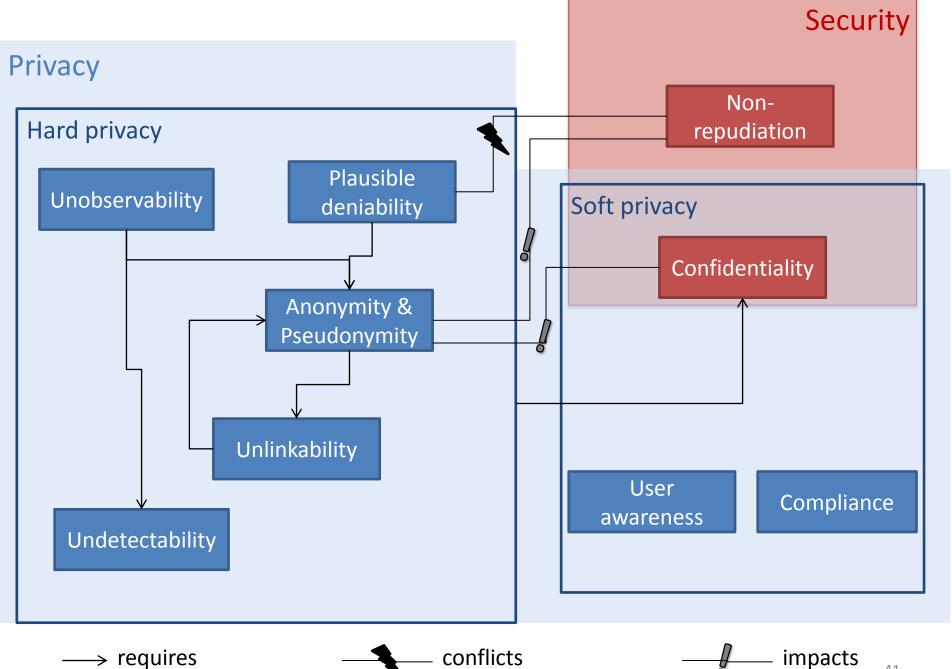




Solution: Confidentiality





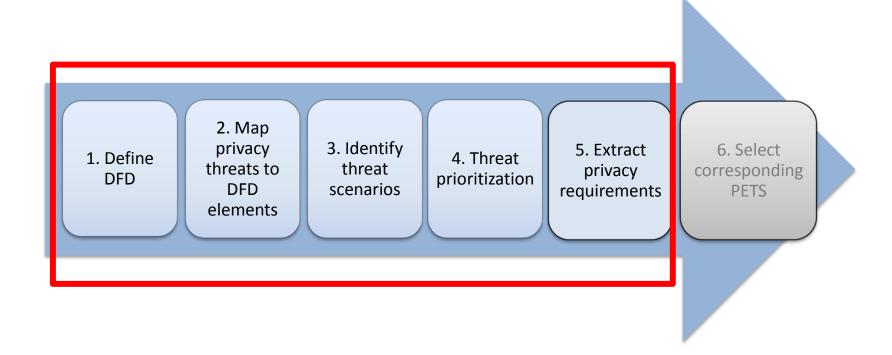


LINDDUN - Privacy threat analysis

PRIVACY METHODOLOGY

Integrating privacy in the system

- Not straight-forward
- Should be part of Software development lifecycle
- Methodology based on STRIDE

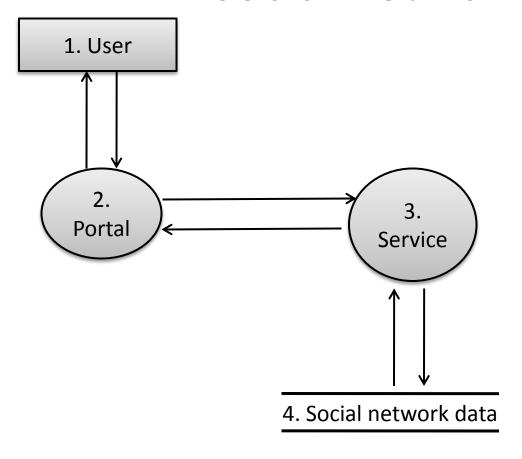


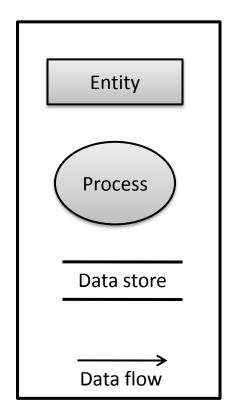


- Step 1
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DFD: social network scenario







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LINDDUN privacy threats

Linkability

Sufficiently distinguish whether 2 IOI are linked or not

Identifiability

Possible to identify the subject within a set of subjects

Non-repudiation

Possible to gather evidence to counter the claims of the repudiating party

Detectability

sufficiently distinguish whether IOI exists or not

Disclosure of Information

Exposal of information to individuals who are not suppose to have access to it

Unawareness of the content

user is unaware of the information he is supplying to the system

Noncompliance of policy/consent

System is not compliant with its advertised policies/consents

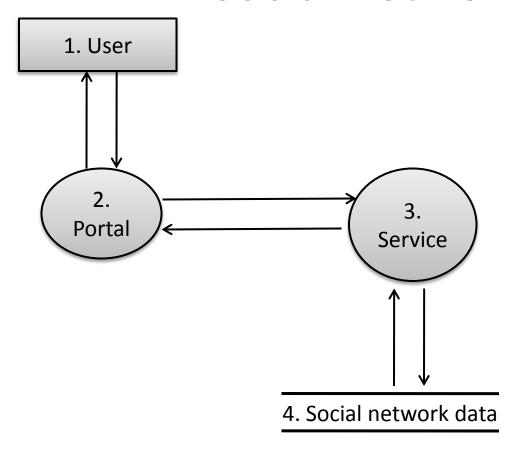


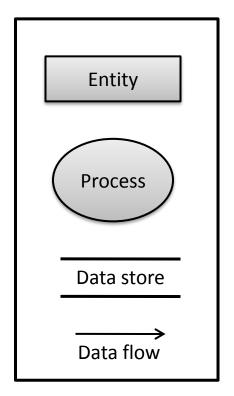
Mapping threats to DFD

	Linkability	Identifiability	Non-repudiation	Detectability	Information Disclosure	Content Unawareness	Policy & Consent Non-compliance
Data store	Х	X	X	X	X		X
Data flow	Х	Х	X	Х	X		X
Process	Х	Х	X	Х	X		X
Entity	Х	Х				Х	



DFD: social network scenario







Mapping Example scenario

	Threat target	L	I	N	D	D	U	N
Data store	Social network db	X	Х	Х	X	X		Х
Data flow	User data stream (user- portal)	Х	Х	Х	Х	X		X
	Service data stream (portal-service)	Х	Х	X	Х	X		Х
	DB data stream (service – DB)	Х	Х	Х	Х	X		Χ
Process	Portal	Х	Х	Х	Х	Х		Х
	Social network service	Х	Х	Х	Х	Х		Х
Entity	User	Х	Х				Х	



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Assumptions

- Assumptions are explicit or implicit choices to trust an element of the system (e.g., human, piece of software) to behave as expected
- The privacy analyst trusts the assumption to be true
- These assumed properties or assertions act as domain restrictions, i.e., they restrict the domain in some way

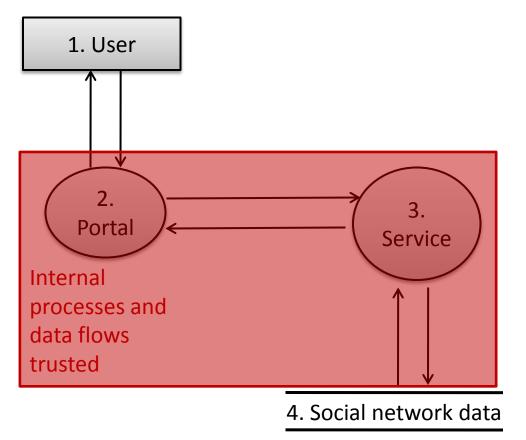


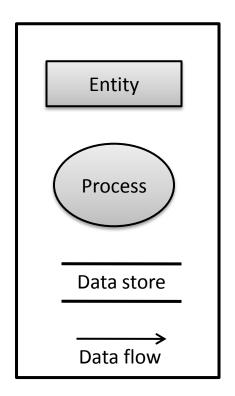
Assumptions

- When adding DFD elements, the number of threats grows exponentially
 - Limit by making assumptions
- Example: assumptions for Social network 2.0:
 - 1. Internal DFD elements are trustworthy.
 - A. trust the processes and data flows in the back-end system.
 - B. do not trust the user and its communication with the portal or the data store containing all the user's information.
 - non-repudiation and detectability threats are considered irrelevant for social networks. (based on threat trees)
 - non-compliance threats are not specific to a specific DFD element, but are applicable to the entire system



DFD: social network scenario





Impact assumptions on example scenario

5. privacy

Priorities

Mapping

	Threat target	L	I	N	D	D	U	N
Data store	Social network db	1	4	Х	X	7		10*
Data flow	User data stream (user- portal)	2	5	Χ	X	8		10*
	Service data stream (portal-service)	Χ	X	Х		X		10*
	DB data stream (service – DB)	X	X	Х	X	Χ		10*
Process	Portal	X	X	Х		Х		10*
	Social network service	X	Χ	Х	Χ	Χ		10*
Entity	User	3	6				9	56



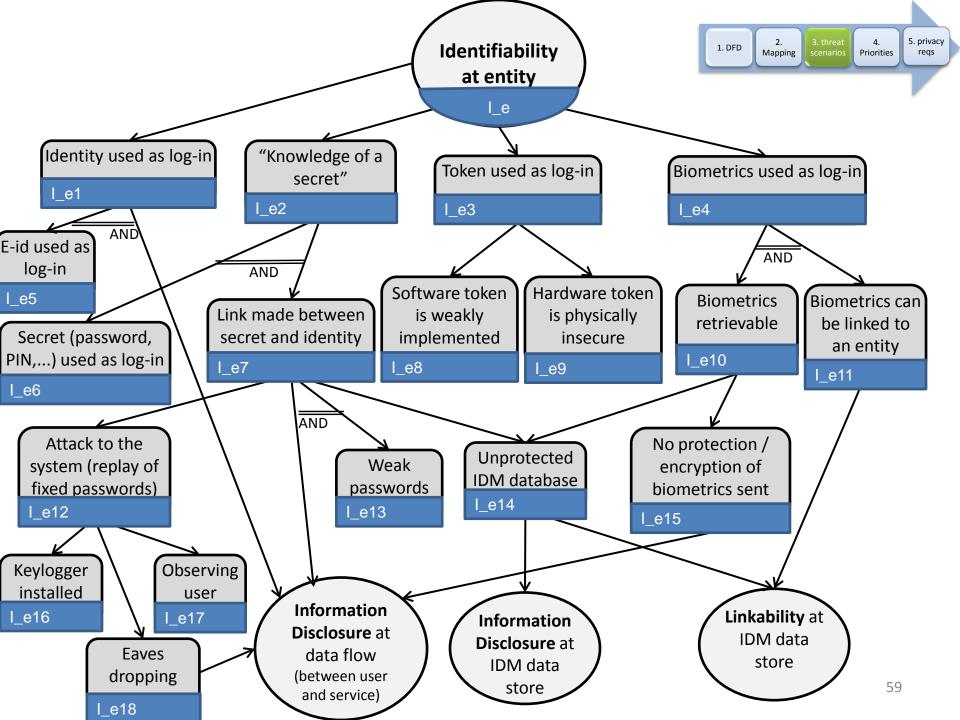
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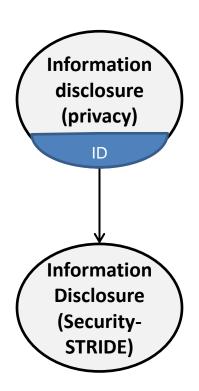
Privacy threat tree patterns

- Illustrate the most common attack patterns
- Used to determine threat applies to system

- Note:
 - Do not limit your analysis to these trees

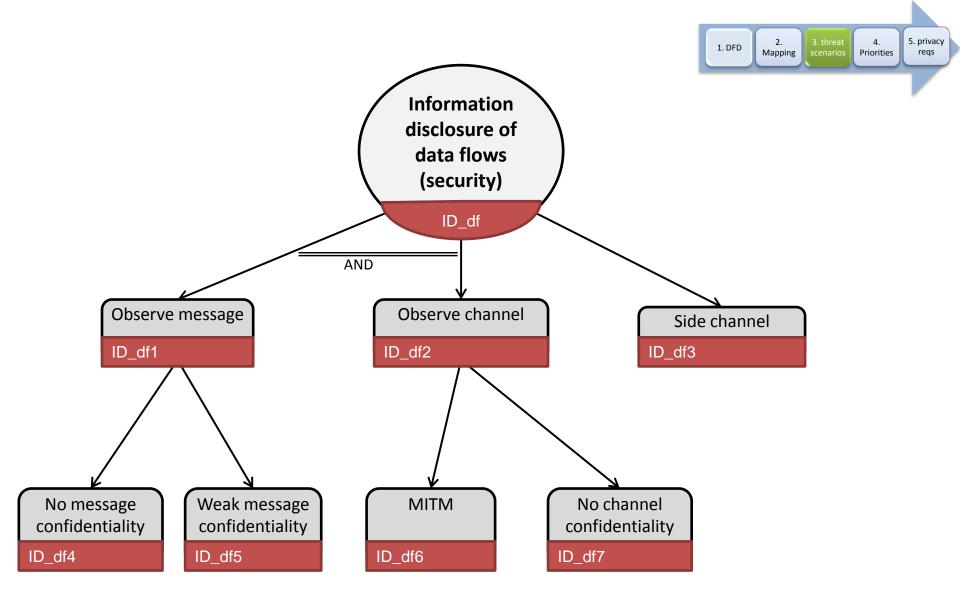






STRIDE revisited

- Systematic approach for security threat identification
- Spoofing, Tampering, Repudiation, Information disclosure, Denial of service, Elevation of privilege





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1. DFD

2. Mapping



Threat description

Inspired by Misuse Cases template

- ID & Title
- Summary
- Misactor profile
- Basic path
- Consequence

- Leaf node(s)
- Root node(s)
- DFD element(s)

Remarks

In your report

✓ Mention the threats in the order you found them

Threat description Example (naive) 1/3

ID & Title

- T01. Identify users of the social network system

Summary

 A misactor gains access to the "secret" sent by the user to log-in and deduces the user's identity from it

Misactor profile

skilled outsider

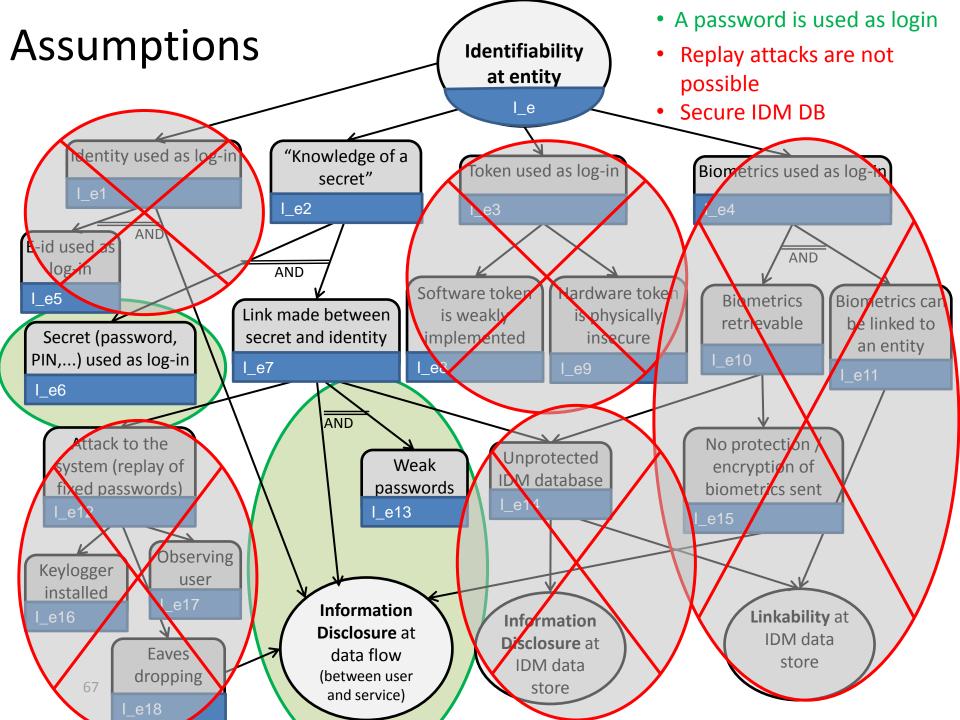
Threat description Example (naive) 2/3

Basic path

- 1. The misactor gains access to the data flow between the user and the portal
- 2. The data contains the user's password
- The misactor can directly link the password to the user due to weak password use (e.g. initials + birthdate)

Consequence

The user's identity is compromised



Threat description Example (naive) 3/3

- Reference to leaf node(s): |_e6, |_e13
- Reference to root node: |_e
- DFD element: User
- Remarks:
 - the data flow between the user and the portal is susceptible to information disclosure threats (assumption 1B). This threat is described in T06.
 - A password is used as log-in (Assumption 4)
 - Replay attacks are not considered a threat (Assumption 5)
 - The IDM database is considered secure (Assumption 6)

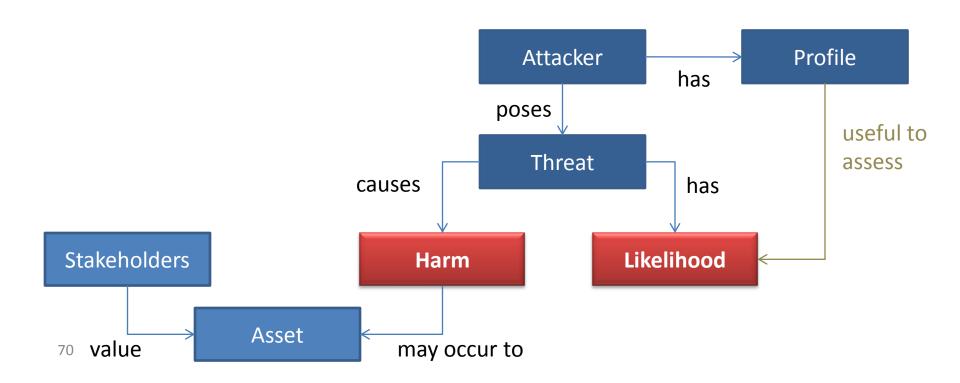
If these assumptions do not hold, the threat tree leaf nodes will result in additional threats



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The role of risk

- Risk is a function of the likelihood of a threat and the severity of its impact on the organization
 - R = f(likelihood , impact)





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Privacy requirements

- Possible mitigation techniques ¹
 - Do nothing
 - Remove feature
 - Turn off feature
 - Warn user
 - Counter threats
 - with preventive or reactive technology

1. DFD

2. Mapping scenarios Priorities F. privacy reqs

From threats to requirements

Threat	Requirement
Linkability	Unlinkability
Identifiability	Anonymity(/pseudonymity)
Non-repudiation	Plausible deniability
Detectability	Undetectability
Disclosure of information	Confidentiality
Unawareness of content	Content awareness
Noncompliance of policy/consent	Consent/policy compliance of the system
	system Straight-form mapping

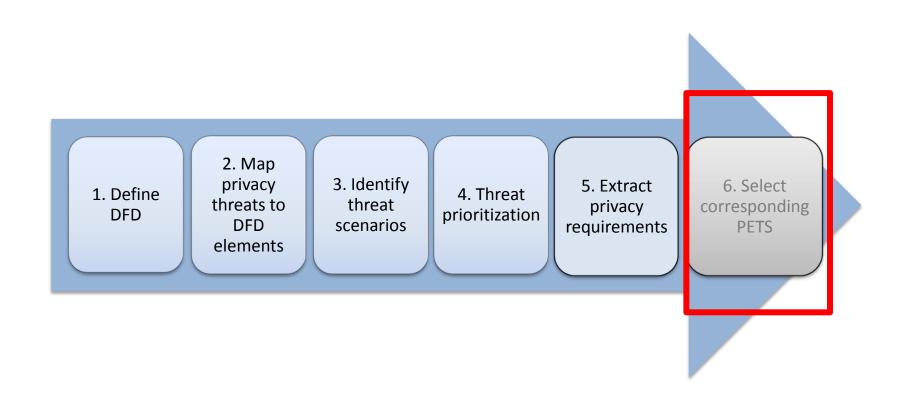


From threats to requirements

- Requirements should not be limited to straight-forward mapping
 - Look at each leaf node that causes threat
 - Determine for each node the proper mitigation

Threats (misuse cases)	Caused by (leaf nodes)	Mitigated by (requirements)
	Information disclosure of data flow	Ensure confidential communication channel (encryption)
Deducing identity from		System should reject weak passwords (at registration) OR
password	Weak passwords	Users should be made aware of consequences of weak passwords (e.g. feedback given at registration)

LINDDUN Methodology



Anonymous communication Mix-networks (1981) [29], DC-networks (1985) [30,31],									
ISDN-mixes 32], Onion Routing (1996) 33], Crowds (1998) 34], Single proxy (908) (Penet pseudonymous remailer (1993-1996), Anonymizer, SafeWeb), anonymous Remailer (Cipherpunk Type 0, Type 1 35], Mixmaster Type 2 (1994) 36], Mixminion Type 3 (2003) 37]), and Low-latency communication (Freedom Network (1999-2001) 138], Java Anon Proxy (JAP) (2000) 39], Tor (2004) 40]) DC-net & MIX-net + dummy traffic, ISDN-mixes 32]		Mitigation techniques: PETs	U	Α	P	D	C	W	O
Privacy preserving authentication [43,44]	Anonymous communication	ISDN-mixes [32], Onion Routing (1996) [33], Crowds (1998) [34], Single proxy (90s) (Penet pseudonymous remailer (1993-1996), Anonymizer, SafeWeb), anonymous Remailer (Cipherpunk Type 0, Type 1 [35], Mixmaster Type 2 (1994) [36], Mixminion Type 3 (2003) [37]), and Low-latency communication (Freedom Network (1999-2001) [38], Java Anon Proxy (JAP) (2000) [39], Tor (2004)	×	×			×		
Privacy preserving authentication Anonymous credentials (single show [45], multishow [46])			×	×		×	×		
Anonymous credentials (single show [45], multishow [46]) × × × × × × × × × × × × × × × × × × ×		Broadcast systems [41,42] + dummy traffic	×	×		×			
Deniable authentication [47]		Private authentication [43,44]	×	×					
Off-the-record messaging [48]									
Privacy preserving crypto- graphic protocols Multi-party computation (Secure function evaluation) [49,									
graphic protocols 50] Anonymous buyer-seller watermarking protocol [51]			×	×	×		×		
Information retrieval Private information retrieval [52] + dummy traffic Oblivious transfer [53,54]) Privacy preserving data mining [55,56] Searchable encryption [57] / Private search [58] Data anonymization K-anonymity model [23,59], I-Diversity [60] X Information hiding Steganography [61] Covert communication [62] Spread spectrum [63] Pseudonymity systems Privacy-enhancing identity management system [64] User-controlled identity management system [65] Privacy-preserving biometrics [66] X X Encryption techniques Symmetric key & public key encryption [67] Deniable encryption Homomorphic encryption [68] Verifiable encryption [69] Access control techniques Context-based access control [70] Privacy-aware access control [71,72] Policy and feedback tools Policy communication (P3P [19]) Policy enforcement (XACML [73], EPAL [74]) Feedback tools for user privacy awareness [12,13,75] Data removal tools (spyware detection and removal, browser cleaning tools, activity traces eraser, harddisk data		50]	×				×		
Oblivious transfer [53,54])		Anonymous buyer-seller watermarking protocol [51]	×	×			×		
Privacy preserving data mining [55,56]	Information retrieval		×	×		×			
Searchable encryption [57] / Private search [58]									
Information hiding Steganography [61] Covert communication [62] Spread spectrum [63] Pseudonymity systems Privacy-enhancing identity management system [64] User-controlled identity management system [65] Privacy-preserving biometrics [66] Encryption techniques Symmetric key & public key encryption [67] Deniable encryption Homomorphic encryption [68] Verifiable encryption [69] Access control techniques Context-based access control [70] Privacy-aware access control [71,72] Policy and feedback tools Policy communication (P3P [19]) Policy enforcement (XACML [73], EPAL [74]) Feedback tools for user privacy awareness [12,13,75] Data removal tools (spyware detection and removal, browser cleaning tools, activity traces eraser, harddisk data			×						
Covert communication [62]	Data anonymization	K-anonymity model [23,59], 1-Diversity [60]	×	×					
Spread spectrum [63]	Information hiding	Steganography [61]	×	×		×			
Pseudonymity systems Privacy-enhancing identity management system [64]									
User-controlled identity management system [65]		Spread spectrum [63]	×	×		×			
Privacy-preserving biometrics [66]	Pseudonymity systems								
Encryption techniques Symmetric key & public key encryption [67] Deniable encryption Homomorphic encryption [68] Verifiable encryption [69] Access control techniques Context-based access control [70] Privacy-aware access control [71,72] Policy and feedback tools Policy communication (P3P [19]) Policy enforcement (XACML [73], EPAL [74]) Feedback tools for user privacy awareness [12,13,75] Data removal tools (spyware detection and removal, browser cleaning tools, activity traces eraser, harddisk data									
Deniable encryption Homomorphic encryption [68] Verifiable encryption [69] Access control techniques Context-based access control [70] Privacy-aware access control [71,72] Policy and feedback tools Policy communication (P3P [19]) Policy enforcement (XACML [73], EPAL [74]) Peedback tools for user privacy awareness [12,13,75] Data removal tools (spyware detection and removal, browser cleaning tools, activity traces eraser, harddisk data		Privacy-preserving biometrics [66]	×	×			×		
Homomorphic encryption [68] Verifiable encryption [69] Access control techniques Context-based access control [70] Privacy-aware access control [71,72] Policy and feedback tools Policy communication (P3P [19]) Policy enforcement (XACML [73], EPAL [74]) Feedback tools for user privacy awareness [12,13,75] Data removal tools (spyware detection and removal, browser cleaning tools, activity traces eraser, harddisk data	Encryption techniques								
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Access control techniques Context-based access control [70] Privacy-aware access control [71,72] Policy and feedback tools Policy communication (P3P [19]) Policy enforcement (XACML [73], EPAL [74]) Feedback tools for user privacy awareness [12,13,75] Data removal tools (spyware detection and removal, browser cleaning tools, activity traces eraser, harddisk data									
Privacy-aware access control [71,72] × Policy and feedback tools Policy communication (P3P [19]) × Policy enforcement (XACML [73], EPAL [74]) × Feedback tools for user privacy awareness [12,13,75] × Data removal tools (spyware detection and removal, browser cleaning tools, activity traces eraser, harddisk data	A								
Policy enforcement (XACML [73], EPAL [74]) × Feedback tools for user privacy awareness [12,13,75] × Data removal tools (spyware detection and removal, browser cleaning tools, activity traces eraser, harddisk data	Access control techniques								
	Policy and feedback tools	Policy enforcement (XACML [73], EPAL [74]) Feedback tools for user privacy awareness [12,13,75] Data removal tools (spyware detection and removal,							×

U: unlinkability

A: anonymity

P: plausible deniability

D: undetectability

C: confidentiality

W: content awareness

O: consent/policy compliance

Social network example

Nr	мис	Requirements	Solutions
1	Linkability data store	Unlinkability of data entries within the social network database	Data anonymization techniques: K-anonymity
		Protection of data store	Access control: relationship- based
2	Linkability data flow	Unlinkability of messages	Anonymous communication
		Channel confidentiality	system: TOR
•••			
9	Content unawareness of user	Make users aware that they only need to provide minimal set of information	Use feedback tools to raise awareness
10	Policy and consent non-compliance	Design system in compliance with legal guidelines	Assign policy compliance responsibility to employee
		Ensure user aware of legitimate actions to perform	User can sue organization
		Employee contracts specify internal rules	Employees disclosing information are penalized

Suggested reading

Privacy

- Pfitzmann & Hansen (2010): A terminology for talking about privacy by data minimization: Anonymity, Unlinkability, Undetectability, Unobservability, Pseudonymity, and Identity Management
 - defines anonymity, pseudonymity, undetectability, unobservability, unlinkability
- Guerses (2010): Multilateral Privacy Requirements Analysis in Online Social Network Services (PhD thesis)
 - section 2.2 (pg. 22-32) provide an interesting overview of privacy. Especially interesting for the following concepts: confidentiality, feedback and awareness
- Guarda and Zannone (2008): Towards the development of privacy-aware systems
 - for those interested in the legal aspects of privacy
 - summarize the privacy principles from a legislation perspective, as it is clearly also important that a system is compliant with law, policies, and user consent (policy and consent compliance)

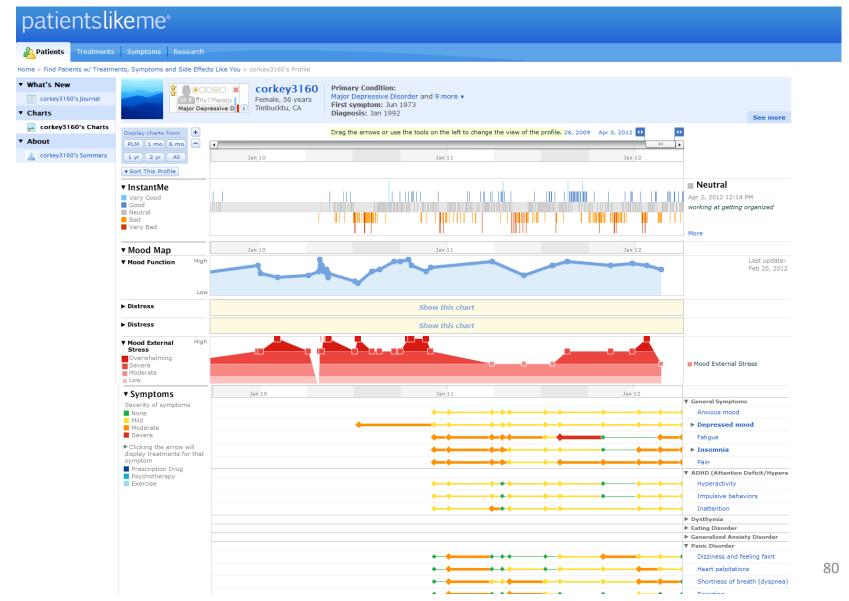
Methodology

- LINDDUN: a privacy threat analysis framework
 - http://people.cs.kuleuven.be/~kim.wuyts/ERISE/LINDDUN.pdf

LINDDUN - Privacy threat analysis

EXAMPLE – PATIENT COMMUNITY SYSTEM

Existing patient communities



Patient communities case study

Patient

- Store personal health data (PHR)
- Retrieve (pseudonymized) PHR from other patients (group members) with same condition
- Retrieve trustworthy information on diseases and treatments (from external service)

Nurse

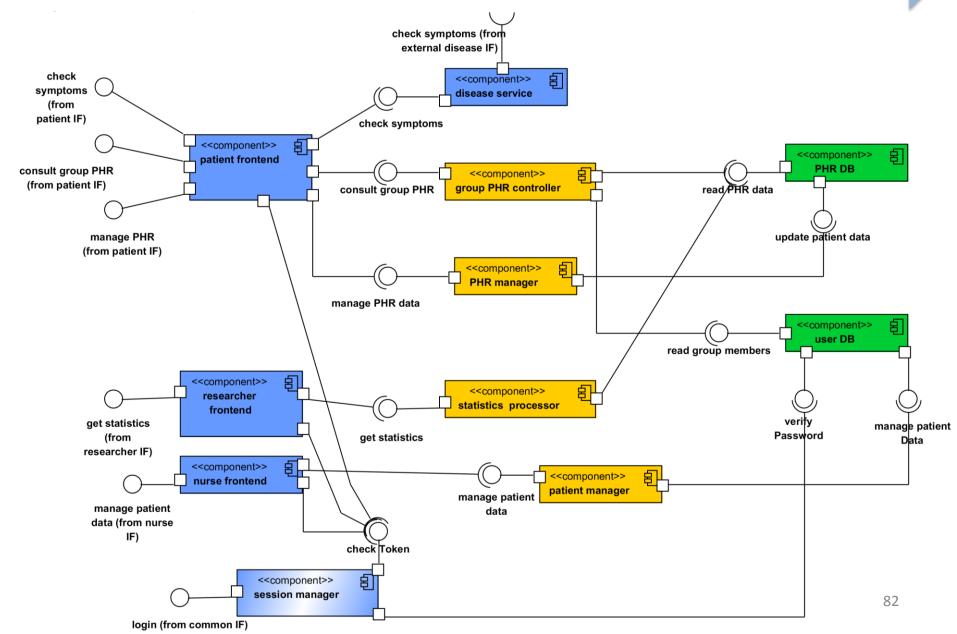
Add users and manage groups

Researcher

Retrieve (anonymized) PHR data to use in analysis

Client-server view





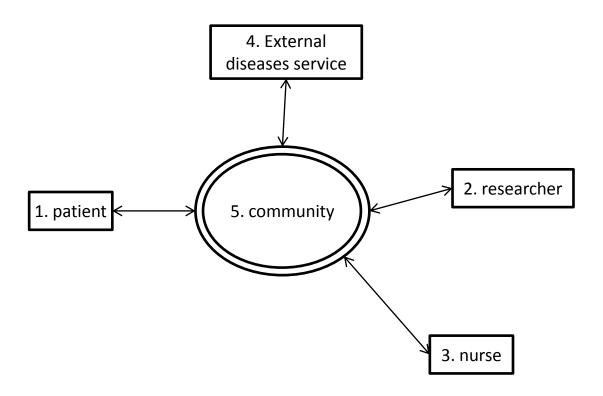


LINDDUN Methodology

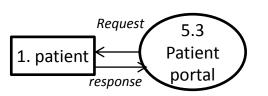
- Step 1
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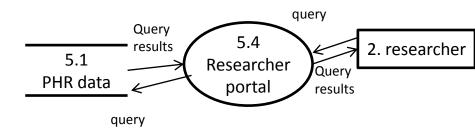
DFD level 0

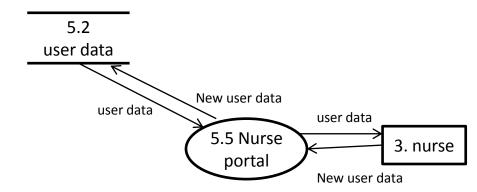


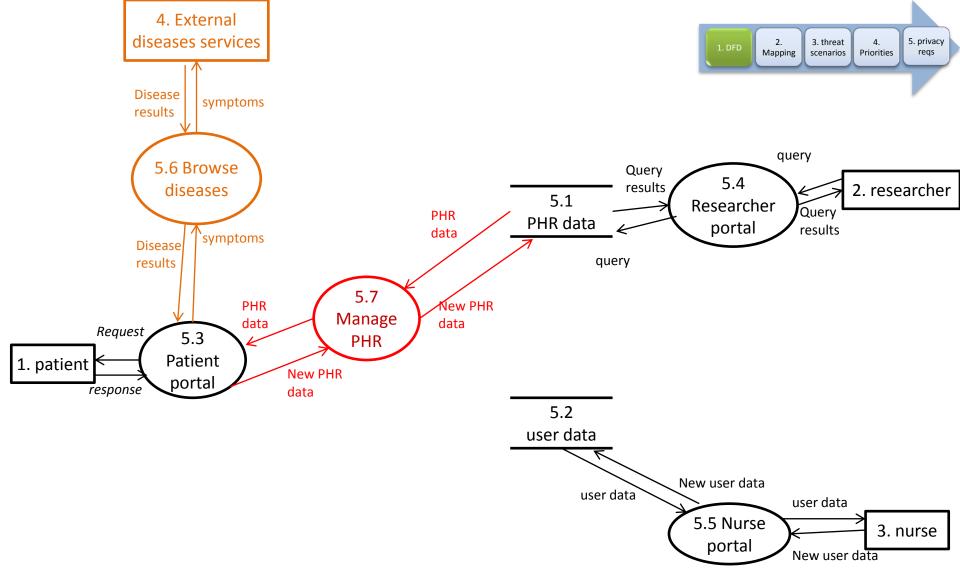
4. External diseases services

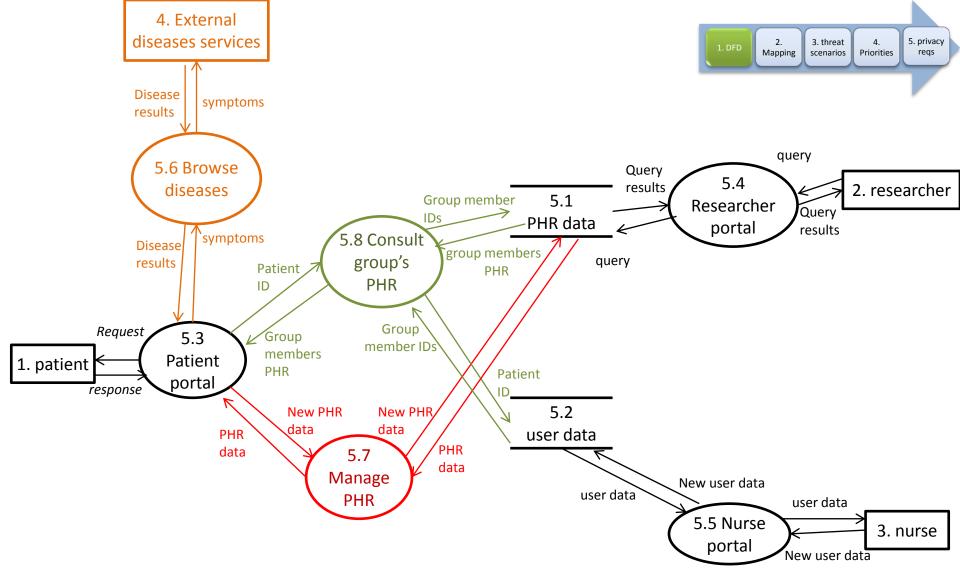


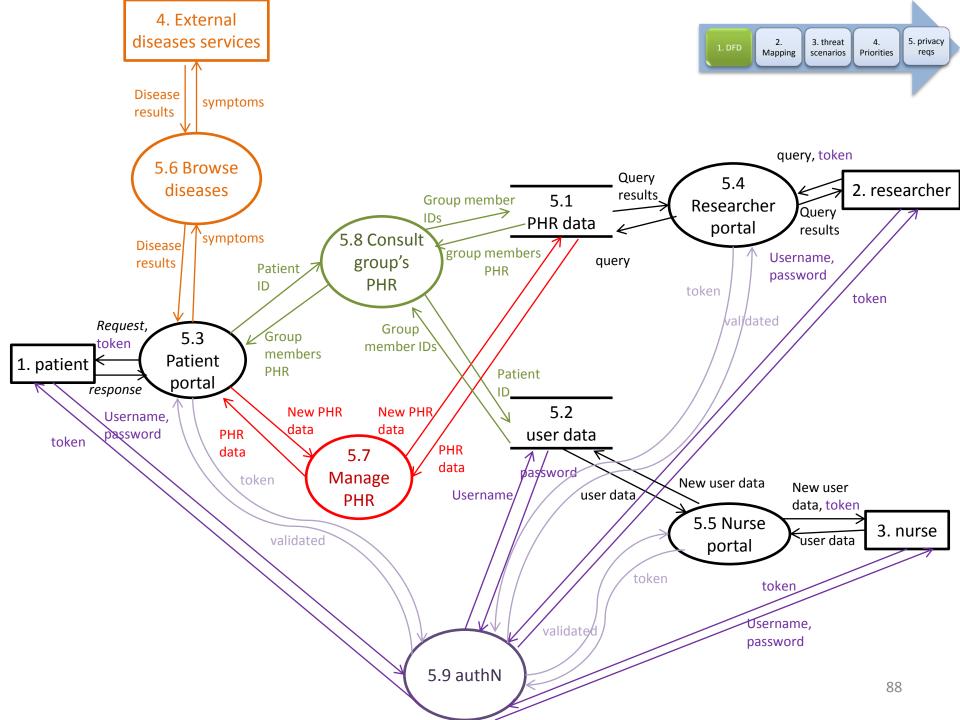














LINDDUN Methodology

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Mapping threats to DFD

	Linkability	Identifiability	Non-repudiation	Detectability	Information Disclosure	Content Unawareness	Policy & Consent Non-compliance
Data store	X	Х	X	X	X		X
Data flow	X	Х	X	X	X		X
Process	Х	Х	X	X	Х		X
Entity	Х	Х				X	

		L	1	N	1. DFD		threat 4. 5. privacy reqs
Data store	PHR data (5.1)	X	X	X		viapping 3cc	THORIES 1845
	User data (5.2)	Х	Х	Х	Х	Х	x
flow	Patient – portal flow (1 -5.3)	Х	X	X	Х	X	x
	Portal – patient flow (5.3-1)	Х	Х	Х	Х	Х	x
	Researcher – portal flow (2-5.4)	Х	Х	Х	X	Х	x
	Portal – researcher flow (5.4-2)	Х	Х	Χ	Χ	Х	x
	Nurse – portal flow (3-5.5)	Х	Х	Х	Х	Х	x
	Portal – nurse flow (5.5-3)	Х	Х	Х	Χ	Х	x
	Disease service – browse diseases flow (4-5.6)	Х	Х	Х	X	Х	x
	Browse disease – disease service flow (5.6-4)	Х	Х	Х	Χ	Х	x
	Patient portal - browse diseases flow (5.3-5.6)	Х	Х	Х	X	Х	x
	Browse diseases – patient portal flow (5.6-5.3)	Х	Х	Х	Χ	Х	x
	Patient portal – manage PHR flow (5.3-5.7)	Х	Х	Х	Х	Х	x
	Manage PHR flow (5.7-5.3)	Х	Х	Х	Χ	Х	x
	Patient portal – consult group PHR (5.3-5.8)	Х	Х	Х	Х	Х	x
	Consult group PHR – patient portal flow (5.8-5.3)	Х	Х	Х	Χ	Х	x
	Researcher portal – PHR data flow (5.4-5.1)	Х	Х	Х	Χ	Х	x
	PHR data – researcher portal flow (5.1-5.4)	Х	Х	Х	Χ	Х	х
	Nurse portal – user data flow (5.5-5.2)	Х	Х	Х	Х	Х	x
	User data – nurse portal flow(5.2-5.5)	Х	Х	Х	Х	Х	x
	Manage PHR – PHR data (5.7-5.1)	Х	Х	Х	Х	Х	X
	PHR data – manage PHR (5.1-5.7)	Х	Х	Х	Х	Х	91 x

1. DFD	2. Mapping	3. threat scenarios	4. Priorities	5. privacy regs
	тиаррия	secilarios	Thomas	1342

	L	I	N	D	D	U	N
Consult group PHR – PHR data flow (5.8-5.1)	Χ	Χ	Χ	Χ	Χ		Х
PHR data – consult group PHR flow (5.1-5.8)	Χ	Χ	X	Χ	Х		Х
Consult group PHR – user data (5.8-5.2)	Χ	Χ	X	Χ	Х		х
User data – consult group PHR (5.2-5.8)	Χ	Χ	Χ	Χ	Х		х
Patient – authN flow (1-5.9)	Χ	Χ	X	Χ	Х		х
authN - patient flow (5.9-1)	Χ	Χ	Χ	Χ	Χ		х
Research – authN flow (2-5.9)	Χ	Χ	X	Χ	Х		х
authN flow – researcher (5.9-2)	Χ	Χ	Χ	Χ	Χ		х
Nurse – authN flow (3-5.9)	Χ	Χ	X	Χ	Х		х
authN – nurse (3-5.9)	Χ	Χ	Χ	Χ	Χ		х
User data – authN flow (5.2-5.9)	Χ	Χ	X	Χ	Х		х
authN – user data flow (5.9-5.2)	Χ	Χ	Χ	Χ	Χ		х
Patient portal – authN (5.3-5.9)	Χ	Χ	Χ	Χ	Χ		х
authN – patient portal (5.9-5.3)	Χ	Χ	Χ	Χ	Χ		х
Researcher portal – authN (5.4-5.9)	Χ	Χ	Χ	Χ	Χ		х
authN – researcher portal (5.9-5.4)	Χ	Χ	Χ	Χ	Χ		х
Nurse portal – authN (5.5-5.9)	Χ	Χ	Χ	Χ	Χ		х
authN – nurse portal (5.9-5.5)	Χ	Х	Χ	Χ	Χ		х



		L	1	N	D	D	U	N
process	Patient portal (5.3)	Χ	Χ	Χ	Χ	Χ		х
	Researcher portal (5.4)	Χ	Χ	Χ	Χ	Χ		х
	Nurse portal (5.5)	Х	Х	Х	Χ	Х		х
	Browse disease (5.6)	Χ	Χ	Χ	Χ	Χ		х
	Manage PHR (5.7)	Х	Х	Х	Χ	Х		х
	Consult group PHR (5.8)	Х	Χ	Χ	Χ	Х		х
	authN (5.9)	Χ	Х	Х	X	Х		х
Entity	Patient (1)	Χ	Χ				Χ	
	Researcher (2)	Х	Х				Х	
	Nurse (3)	Χ	Χ				Χ	
	External disease service (4)	Χ	X				Х	



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General assumptions

- all internal processes are only susceptible to insider threats, as we consider the back-end sufficiently protected against outsider threats. We will therefore combine the process threats and examine only one, as the threats apply to all of them
- 2. all data flows between internal processes and between internal processes and internal data stores are only susceptible to insider threats, as we consider the back-end sufficiently protected against outsider threats. We will therefore combine the data flow threats and examine only one, as the threats apply to all of them
- 3. data flows between an entity and a process are not considered trusted (as it involves transactions of an external entity to and from a trusted process over an insecure communication line)
- 4. data stores are not considered confidential, as no access control system is present

Positive assumptions can help understand reasoning

		L	- 1	N	1. DFD	2. 3. thr	
Data store	PHR data (5.1)	X	X	X		Mapping scena	rios Priorities reqs
	User data (5.2)	Χ	Х	Х	Х	Х	x
flow	Patient – portal flow (1 -5.3)	X	Χ	X	Х	X	Х
	Portal – patient flow (5.3-1)	X	Χ	Х	Х	X	Х
	Researcher – portal flow (2-5.4)	X	X	Х	Х	X	Х
	Portal – researcher flow (5.4-2)	X	Χ	Х	Х	X	Х
	Nurse – portal flow (3-5.5)	X	X	Х	Х	Х	Х
	Portal – nurse flow (5.5-3)	X	Χ	Х	Х	X	Х
	Disease service – browse diseases flow (4-5.6)	X	X	Х	Х	X	Х
	Browse disease – disease service flow (5.6-4)	Х	Х	Х	Х	Х	Х
\rightarrow	Patient portal - browse diseases flow (5.3-5.6)	X	Х	Х	Х	Х	х
	Browse diseases – patient portal flow (5.6-5.3)	X	Χ	Х	Х	Х	Х
	Patient portal – manage PHR flow (5.3-5.7)	X	X	X	X	X	X
	Manage PHR flow (5.7-5.3)	X	Χ	Х	Χ	Χ	Х
nternal data	Patient portal – consult group PHR (5.3-5.8)	X	Χ	Х	X	Χ	Х
flows	Consult group PHR – patient portal flow (5.8-5.3)	X	Х	Х	Χ	X	Х
combined	Researcher portal – PHR data flow (5.4-5.1)	X	Χ	Х	Х	X	Х
	PHR data – researcher portal flow (5.1-5.4)	Χ	Χ	Х	Х	Χ	Х
	Nurse portal – user data flow (5.5-5.2)	X	Χ	X	Х	Х	Х
	User data – nurse portal flow(5.2-5.5)	X	Χ	Х	Х	X	Х
	Manage PHR – PHR data (5.7-5.1)	X	Х	Х	Х	Х	X
	PHR data – manage PHR (5.1-5.7)	Х	Χ	Χ	X	Х	96 X

1. DFD	2. Mapping	3. threat scenarios	4. Priorities	5. privacy reqs

		L	- 1	N	D	D	U	N
	Consult group PHR – PHR data flow (5.8-5.1)	Х	Χ	Χ	Х	Х		Х
Internal data	PHR data – consult group PHR flow (5.1-5.8)	X	Χ	X	Х	Х		х
flows	Consult group PHR – user data (5.8-5.2)	X	Х	Х	Х	Х		х
combined	User data – consult group PHR (5.2-5.8)	X	Х	Х	Х	Х		Х
	Patient – authN flow (1-5.9)	Х	Х	Х	Х	Х		Х
	authN - patient flow (5.9-1)	Х	Х	Х	Χ	Х		х
	Research – authN flow (2-5.9)	Х	Х	Х	Х	Х		х
	authN flow – researcher (5.9-2)	Х	Х	Х	Х	Х		Х
	Nurse – authN flow (3-5.9)	Х	Х	Х	Х	Х		х
	authN – nurse (3-5.9)	Х	Х	Х	Х	Χ		х
	User data – authN flow (5.2-5.9)	Х	Χ	Χ	Χ	Χ		Х
	authN – user data flow (5.9-5.2)	X	Χ	X	Х	Х		Х
	Patient portal – authN (5.3-5.9)	Х	Χ	X	Χ	Х		Х
Internal data	authN – patient portal (5.9-5.3)	X	Χ	X	Χ	Х		Х
flows	Researcher portal – authN (5.4-5.9)	X	Х	Х	Х	Х		Х
combined	authN – researcher portal (5.9-5.4)	X	Х	Х	Х	Х		Х
	Nurse portal – authN (5.5-5.9)	X	Х	Х	Х	Х		Х
	authN – nurse portal (5.9-5.5)	X	Х	Х	Х	Х		Х
								97

(



		L	ı	N	D	D	U	N
process	Patient portal (5.3)	Х	Х	Х	Х	Х		х
	Researcher portal (5.4)	Х	Χ	Х	Χ	Х		Х
	Nurse portal (5.5)	Χ	Χ	X	Х	Х		Х
	Browse disease (5.6)	X	Χ	X	X	Х		Х
Internal	Manage PHR (5.7)	X	Χ	X	Χ	Х		Х
processes	Consult group PHR (5.8)	Χ	Χ	X	Χ	X		Х
combined	authN (5.9)	Х	Х	Х	Х	Х		х
Entity	Patient (1)	Χ	Х				Х	
	Researcher (2)	Х	Х				X	
	Nurse (3)	Χ	Χ				Χ	
	External disease service (4)	Χ	Χ				X	

			L	- 1	N	1. DFD	2. 3. thr Mapping scena		4. 5. privac
	Data store	PHR data (5.1)	Χ	Χ	х	٨	^		×
		User data (5.2)	Х	Х	Х	Χ	Х		Х
	flow	Patient – portal flow (1 -5.3)	Х	Х	Х	Χ	Х		X
		Portal – patient flow (5.3-1)	Х	Х	Х	Χ	Х		Х
		Researcher – portal flow (2-5.4)	Х	Х	Χ	Χ	Х		X
		Portal – researcher flow (5.4-2)	Х	Х	Χ	Х	Х		X
		Nurse – portal flow (3-5.5)	Х	Х	Χ	Χ	Х		X
		Portal – nurse flow (5.5-3)	Х	Х	Χ	Χ	Х		X
Gen	eral	Disease service – browse diseases flow (4-5.6)	Х	Х	Х	Х	Х		X
inte	rnal	Browse disease – disease service flow (5.6-4)	Х	Х	Χ	Χ	Х		X
DF		Patient portal - browse diseases flow (5.3-5.6)	Х	Х	Х	Х	X		Х
1		Patient – authN flow (1-5.9)	Х	Х	Х	Х	Х		Х
		authN - patient flow (5.9-1)	Х	X	Χ	Х	X		х
		Research – authN flow (2-5.9)	Х	Х	Х	Х	X		Х
		authN flow – researcher (5.9-2)	Х	Х	Х	Х	X		Х
Gen	eral	Nurse – authN flow (3-5.9)	Х	Х	Х	Χ	X		X
inte	rnal	authN – nurse (3-5.9)	Х	Х	Х	Х	Х		Х
Р	process	Patient portal (5.3)	Х	Х	Х	Χ	Х		Х
	Entity	Patient (1)	Х	Х				Χ	
		Researcher (2)	Х	Х				Χ	
		Nurse (3)	Х	Х				Χ	99
		External disease service (4)	Х	Х				Χ	



General assumptions

- 5. No non-repudiation threats exist in the system, as the data flows, processes and data stores do not require plausible deniability
- 6. detectability is not considered a threat for this specific system. The privacy concerns of this system are all focused on the data itself, not on the detectability of it
- 7. non-compliance is an important threat, however, it is not specific to one part of the system, but poses to the system as a whole. We will therefore not make a distinction between the different DFD elements for this threat.

		L	1	N	D	D	U	N
Data store	PHR data (5.1)	Х	Х	Х	Χ	Х		Х
	User data (5.2)	Х	Х	X	X	Х		Х
flow	Patient – portal flow (1 -5.3)	Х	Х	X	X	Х		Х
	Portal – patient flow (5.3-1)	Х	Х	X	X	Х		Х
	Researcher – portal flow (2-5.4)	Х	Х	X	X	Х		Х
	Portal – researcher flow (5.4-2)	Χ	Х	X	Χ	Х		Х
	Nurse – portal flow (3-5.5)	Х	Х	X	Χ	Х		Х
	Portal – nurse flow (5.5-3)	Х	Х	X	Χ	Х		Х
	Disease service – browse diseases flow (4-5.6)	Х	Х	X	X	Х		Х
	Browse disease – disease service flow (5.6-4)	Х	X	X	X	Х		Х
	Patient portal - browse diseases flow (5.3-5.6)	Х	Х	Х	X	Х		Х
	Patient – authN flow (1-5.9)	Х	Х	Х	X	X		Х
	authN - patient flow (5.9-1)	Х	X	X	X	Х		Х
	Research – authN flow (2-5.9)	Х	Х	X	X	Х		Х
	authN flow – researcher (5.9-2)	Х	Х	X	X	Х		Х
	Nurse – authN flow (3-5.9)	Х	Χ	X	X	Х		Х
	authN – nurse (3-5.9)	X	Х	X	X	Х		Х
process	Patient portal (5.3)	Х	Х	Х	X	X		Х
Entity	Patient (1)	Х	Х	No n			Χ	
	Researcher (2)	Х	Х	_	diatio		Χ	
	Nurse (3)	Х	Х	x detectability threat			Χ	101
	External disease service (4)	Х	Χ				Χ	



Assumptions

- 8. Identifiability of entities (researchers, nurses, patients or the external service) is not considered a threat, as all entities should have their own unique (long-term) identifier and there is no need to hide the entity's identity. Knowing that an entity is using the community service is not considered an issue.
- 11. Linkability of entities (sensors, cardiologists, nurses, or patients) is not considered a threat, as all entities should have their own unique (long-term) identifier and there is no need to hide the entity's identity. Knowing that an entity is using the community service is not considered an issue.

		L	1	N	D	D	U	N
Data store	PHR data (5.1)	Χ	X	Х	Χ	Х		Х
	User data (5.2)	Х	X	Х	X	Х		Х
flow	Patient – portal flow (1 -5.3)	Х	X	X	X	Х		Х
	Portal – patient flow (5.3-1)	Χ	Χ	Х	X	Х		Х
	Researcher – portal flow (2-5.4)	Х	X	X	X	Х		Х
	Portal – researcher flow (5.4-2)	Χ	Χ	X	X	Х		Х
	Nurse – portal flow (3-5.5)	Х	X	X	X	Х		Х
	Portal – nurse flow (5.5-3)	Χ	Χ	Х	X	Х		Х
	Disease service – browse diseases flow (4-5.6)	Х	X	X	X	Х		Х
	Browse disease – disease service flow (5.6-4)	Χ	X	X	X	Х		Х
	Patient portal - browse diseases flow (5.3-5.6)	X	х	Х	X	- x -		Х
	Patient – authN flow (1-5.9)	Х	Х	Х	X	_ x		Х
	authN - patient flow (5.9-1)	Χ	Х	Х	X	Х		Х
	Research – authN flow (2-5.9)	Χ	X	X	X	Х		Х
	authN flow – researcher (5.9-2)	Х	X	Х	X	Х		Х
	Nurse – authN flow (3-5.9)	Χ	X	Х	X	Х		Х
	authN – nurse (3-5.9)	_ X	Х	Х	X	X		Х
process	Patient portal (5.3)	Х	Х	Х	X	X		Х
Entity	Patient (1)	Х	Х				Х	
	Researcher (2)	Х	X				Х	
	Nurse (3)	Х	X				Х	103
	External disease service (4)	Х	Χ				Χ	



Assumptions

- 14. Linkability and identifiability do not pose a threat to the data flows between entities (patient, nurse, and researcher) and (portal) processes because of assumptions 8 and 11
- 9. Identifiability of the data flow only poses a threat to one specific data flow: 5.6 ->4 (browse diseases to external disease services), as the external service should not be able to identify the patient that is using this disease browsing service.
- 10. Linkability of the data flow to the external disease service (5.6 -> 4) is the only linkability threat to data flows in the patient community system. Although less likely, when the patient identifiers are replaces by pseudonyms, linking the different symptoms (of different searches) together can still result in an identifiability threat
- 15. Linkability and identifiability do not apply to internal data flows as knowing that 2 requests belong to the same user, or knowing who made a request does not violate the patient's privacy. The patient's privacy is only violated when the content of the communication is revealed (information disclosure threat)
- 16. Linkability and identifiability do not apply to internal processes as knowing that 2 actions belong to the same user does not violate the patient's privacy. The patient's privacy is only violated when the content of the action is revealed (information disclosure threat)

		L	1	N	D	D	U	N
Data store	PHR data (5.1)	Х	Χ	Х	Х	Х		Х
	User data (5.2)	Х	Χ	X	Χ	Х		Х
flow	Patient – portal flow (1 -5.3)	X	Х	Х	X	Х		Х
	Portal – patient flow (5.3-1)	Х	X	Х	X	Х		Х
	Researcher – portal flow (2-5.4)	X	X	Х	X	Х		Х
	Portal – researcher flow (5.4-2)	X	X	X	X	Х		Х
	Nurse – portal flow (3-5.5)	X	X	X	X	Х		Х
	Portal – nurse flow (5.5-3)	Х	X	X	X	Х		Х
	Disease service – browse diseases flow (4-5.6)	X	X	X	X	Х		Х
	Browse disease – disease service flow (5.6-4)	Х	Х	Х	X	Х		Х
	Patient portal - browse diseases flow (5.3-5.6)	Х	Х	X	Х	Х		Х
	Patient – authN flow (1-5.9)	Х	Х	Х	X	Х		Х
	authN - patient flow (5.9-1)	X	Χ	X	X	Х		Х
	Research – authN flow (2-5.9)	X	X	X	X	Х		Х
	authN flow – researcher (5.9-2)	X	X	X	X	Х		Х
	Nurse – authN flow (3-5.9)	X	X	X	X	Х		Х
	authN – nurse (3-5.9)	X	Х	Х	Х	X		Х
process	Patient portal (5.3)	Х	Х	Х	X	X		Х
Entity	Patient (1)	Х	X				Х	
	Researcher (2)	Х	Χ				Χ	
	Nurse (3)	Х	Х				Х	105
	External disease service (4)	Х	Χ				Χ	



Assumptions

- 12. Information disclosure between the external disease service and the browse disease process does not pose a threat, as it does not contain any sensitive or personal information
- 19. Content unawareness only applies to the patient, as the researcher does not add any information, a nurse only registers patients, and the external disease service does not directly input any data
- 17. Identifiability and linkability are applicable to both data stores, and will therefore be examined in a combined fashion

		L	I	N	D	D	U	N
Data store	PHR data (5.1)	Х	Х	Х	X	Х		Х
	User data (5.2)	Х	Х	Х	X	Х		Х
flow	Patient – portal flow (1 -5.3)	X	X	X	X	Х		Х
	Portal – patient flow (5.3-1)	Х	X	X	X	Х		Х
	Researcher – portal flow (2-5.4)	X	X	Х	X	Х		Х
	Portal – researcher flow (5.4-2)	Х	X	Х	X	Х		Х
	Nurse – portal flow (3-5.5)	Х	X	X	X	Х		Х
	Portal – nurse flow (5.5-3)	X	X	X	X	Х		Х
	Disease service – browse diseases flow (4-5.6)	Х	X	X	X	Х		Х
	Browse disease – disease service flow (5.6-4)	Х	Х	Х	X	Х		Х
	Patient portal - browse diseases flow (5.3-5.6)	Х	Х	Х	Х	Х		Х
	Patient – authN flow (1-5.9)	Х	Х	Х	X	_ x		Х
	authN - patient flow (5.9-1)	Х	Χ	Х	X	Х		Х
	Research – authN flow (2-5.9)	Х	X	Х	X	Х		Х
	authN flow – researcher (5.9-2)	Х	X	Х	X	Х		Х
	Nurse – authN flow (3-5.9)	X	X	Х	X	Х		Х
	authN – nurse (3-5.9)	X	Х	Х	Х	Х		Х
process	Patient portal (5.3)	Х	Х	Х	X	Х		Х
Entity	Patient (1)	X	X				Х	
	Researcher (2)	Х	X				X	
	Nurse (3)	Х	X				X	107
	External disease service (4)	X	Χ				X	

		L	1	N	D	D	U	N
Data store	PHR data (5.1)	Х	Х			Χ		Х
	User data (5.2)	Х	Х			Χ		Х
flow	Patient – portal flow (1 -5.3)					Χ		Х
	Portal – patient flow (5.3-1)					Χ		Х
	Researcher – portal flow (2-5.4)					Х		Х
	Portal – researcher flow (5.4-2)					Χ		Х
	Nurse – portal flow (3-5.5)					Χ		Х
	Portal – nurse flow (5.5-3)					Χ		Х
	Disease service – browse diseases flow (4-5.6)					Х		Х
	Browse disease – disease service flow (5.6-4)	Х	X					Х
	Patient portal - browse diseases flow (5.3-5.6)					Х		Х
	Patient – authN flow (1-5.9)					Χ		Х
	authN - patient flow (5.9-1)					Χ		Х
	Research – authN flow (2-5.9)					Χ		Х
	authN flow – researcher (5.9-2)					Χ		Х
	Nurse – authN flow (3-5.9)					Χ		Х
	authN – nurse (3-5.9)					Х		Х
process	Patient portal (5.3)					X		Х
Entity	Patient (1)						X	
	Researcher (2)							
	Nurse (3)							
	External disease service (4)							108



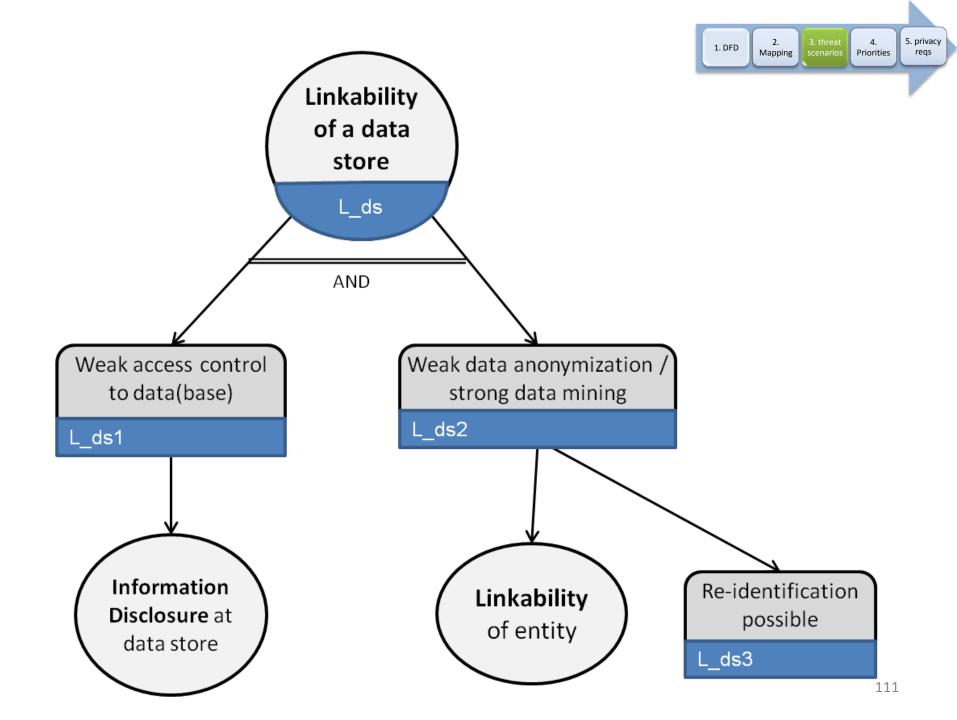
LINDDUN Methodology

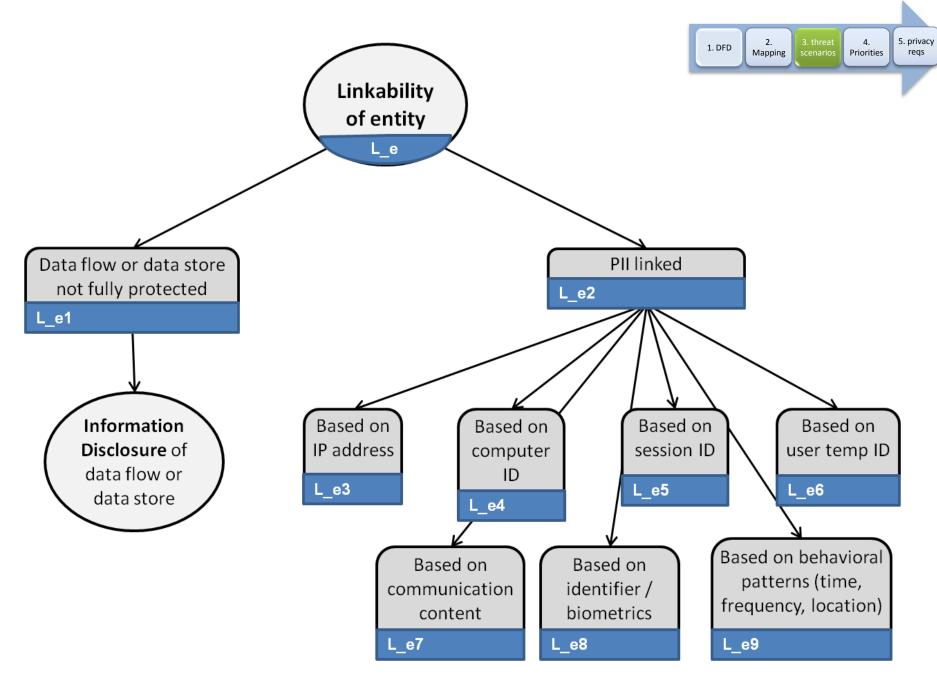
- Step 1
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- Step 5
 - Extract privacy requirements



Linking community data

- Assumption: Identifiability and linkability are applicable to both data stores, and will therefore be examined in a combined fashion
- linking PHR data
 - Applies also to user data





T01 - Profiling PHR data (linking)

Summary: A researcher or other insider with malicious intent links PHR data (or user data)

Primary mis-actor: unskilled insider (authenticated user, e.g. researcher)

Basic path:

- bf1. The misactor performs a set of targeted queries on the PHR data or user data store and retrieves very detailed results
- bf2. The misactor links the results of the queries together (e.g. based on medication which is usually combined, medical conditions which occur together, or pseudo-identifiers like street and age)

Consequence: By combining the query results, the misactor has access to more information about the patient than anticipated

Reference to threat tree node(s): L_ds2, L_e2

Parent threat tree(s): L_ds, I_ds

DFD element(s): 5.1 PHR data, 5.2 user data

- r1. This threat can be used as precondition for the identifiability threat at the data store (T03 Identifying a patient from his PHR data)
- r2. This threat was inspired by L_ds2 and L_e2, however none of L_e2's leaf nodes matched
- r3. The (weak) access requirement (L ds1) is fulfilled because the misactor is an insider who has access to the database
- r4. Although this threat mainly describes the PHR data case, it also applies to the user data store (assumption 4)



Linking community data

- Assumption: Identifiability and linkability are applicable to both data stores, and will therefore be examined in a combined fashion
- linking PHR data
 - Applies also to user data
- Linking PHR data to user data

1. DFD 2. Mapping 3. threat scenarios 4. Priorities 5. privacy reqs

T02 - Linking PHR data to user data

Summary: The administrator or other insider with access to both the PHR data store and user data store is able to link the data from both databases (and sell this information to advertisers, insurance companies, etc.)

Primary mis-actor: unskilled insider with access to both data stores

Basic path:

bf1. The misactor retrieves information from both the PHR data store and the user data store

bf2. The misactor links both sets of data (e.g. based on a shared foreign key)

Consequence: The combined set of data contains (possibly sensitive) personal identifiable information and especially poses a privacy threat when the misactor sells the information (e.g. to a company selling medication, to the patient's insurance company, etc.)

Reference to threat tree node(s): L ds2, L e6

Parent threat tree(s): L_ds, I_ds

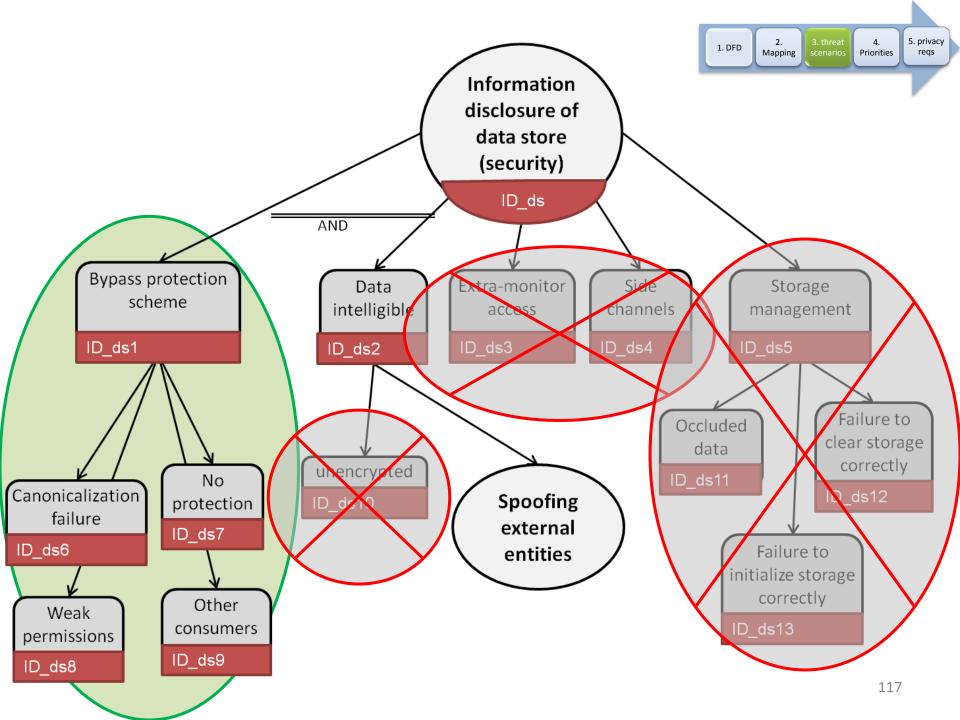
DFD element(s): 5.1 PHR data, 5.2 user data

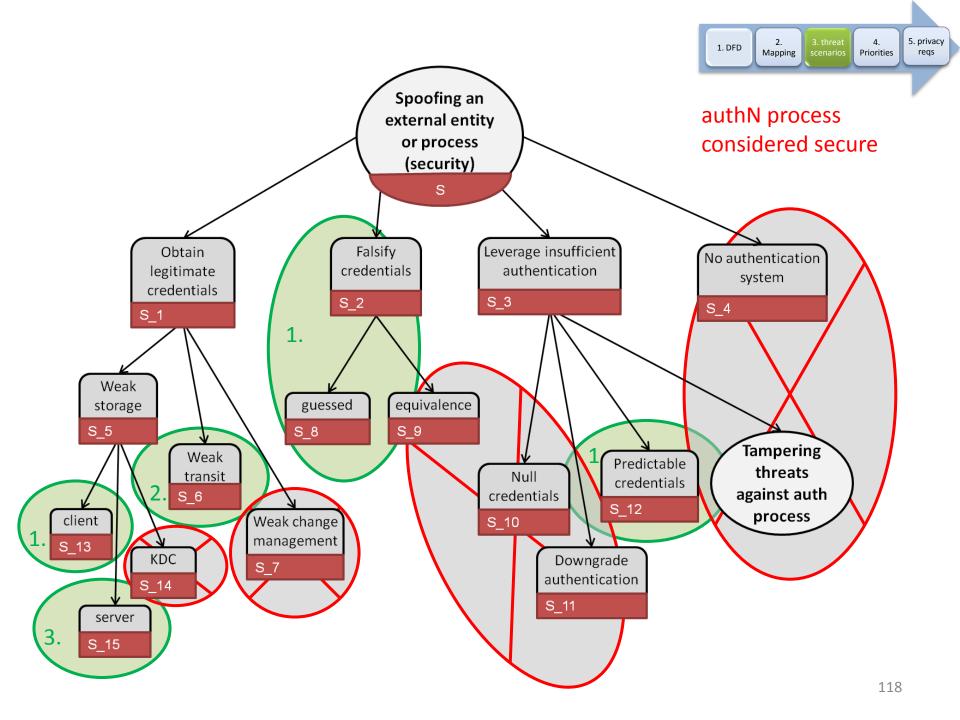
- r1. The L_ds1 requirement of (weak) access is fulfilled, as this threat only involves insiders who have access to the data stores
- r2. The linkability of entity leaf node L_e6, indicating linkability based on the user's temporary ID inspired to this data store linkability threat



Information disclosure of community data

- no access control system is present (assumption 4)
- We assume that the data stores are sufficiently protected and that side-channel attacks, extra-monitor and bad storage management are not possible (assumption 20)



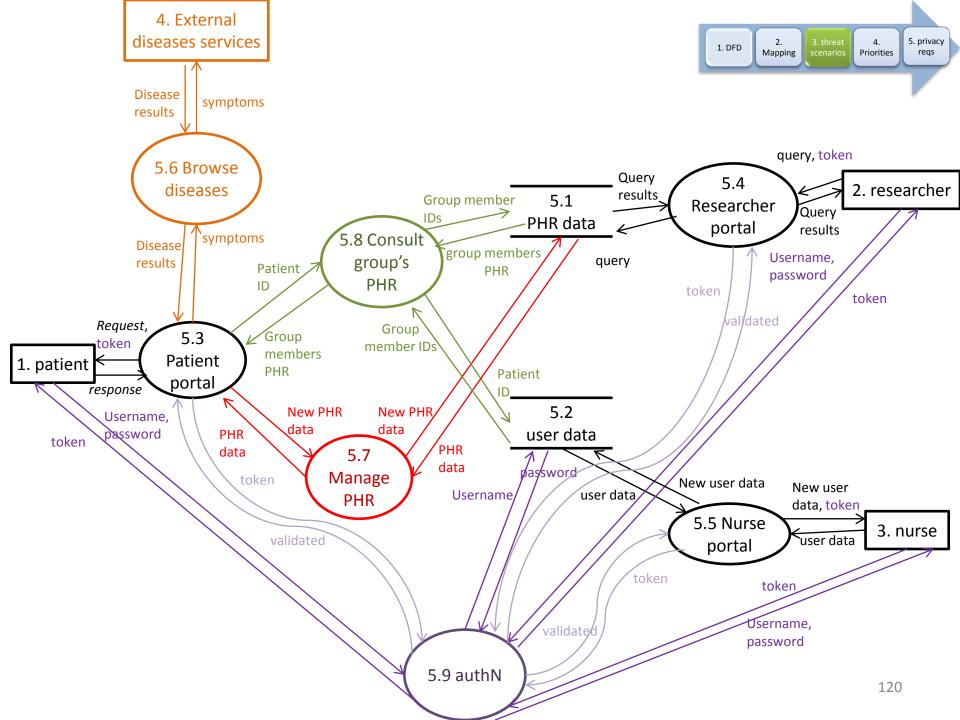


Spoofing users (patients, researchers, nurses)

- Spoofing by falsifying credentials
- Spoofing by eavesdropping communication
 - Information disclosure of transmitted credentials
 - Information disclosure of transmitted session token
- Spoofing because of weak credential storage
 - Information disclosure of community data

5. privacy

Priorities





spoofing external disease service

Spoofing external disease service

Summary: The external disease service is spoofed (e.g. by a competitor or a advertising company for medication)

Primary mis-actor: Skilled outsider

Basic path:

bf1. The misactor pretends to be the disease service

bf2. The community browse service contacts the spoofed disease service with symptoms

bf3. The misactor returns false information

Consequence: The patient community system returns false disease information to the patient which has an impact on the system's reputation (as one of the benefits of the provided service is the trustworthiness of the information)

Reference to threat tree node(s): S 4

Parent threat tree(s): S

DFD element(s): 4. external disease service

This is **NOT** a **privacy** threat.

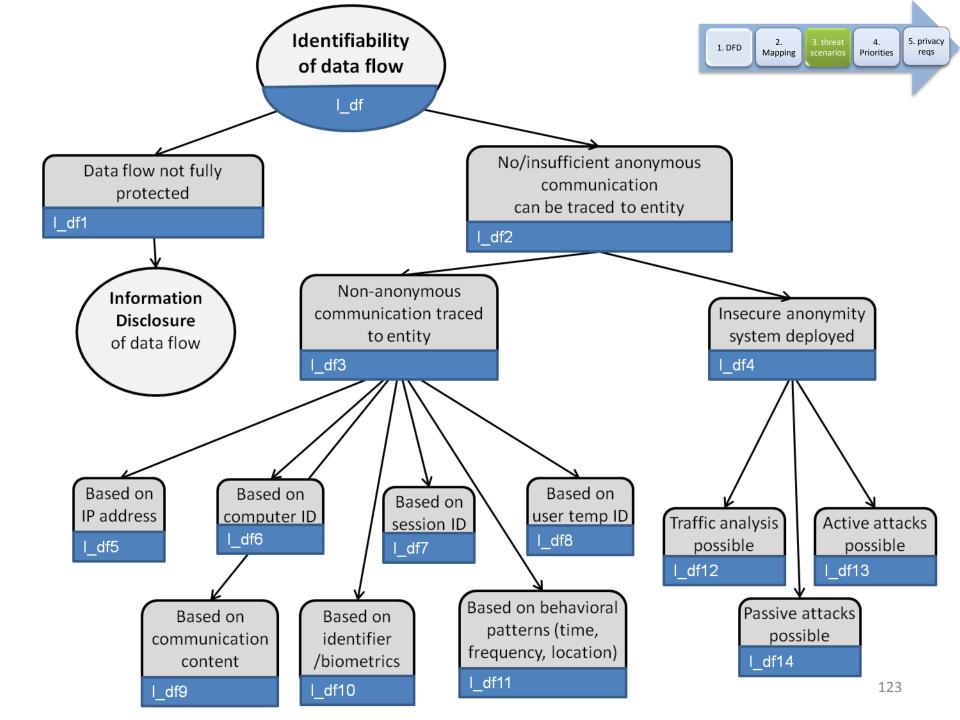
It is a security threat (against integrity)

And should not be included



External disease service

- Linkability & Identifiability of data flow
 - NOT during transit
 - When arrived at external disease service
 - Always information disclosure





T12 - Identifiability of data sent to external disease service

Summary: The misactor extracts the patient's identity from the request and links it to the symptoms

Primary mis-actor: unskilled insider/skilled outsider

Basic path:

- bf1. The patient searches diseases by providing his symptoms to the patient portal, which forwards the request (include the patient's identifiable information (e.g. SSN, address, etc.) to the external disease service
- bf2. The misactor intercepts the data flow (threat T10 Information disclosure of transmitted medical or personal data) or is (or has access to) the external disease service

Consequence: The misactor knows which patient has which symptoms

Reference to threat tree node(s): I_df1, I_df8

Parent threat tree(s): | df

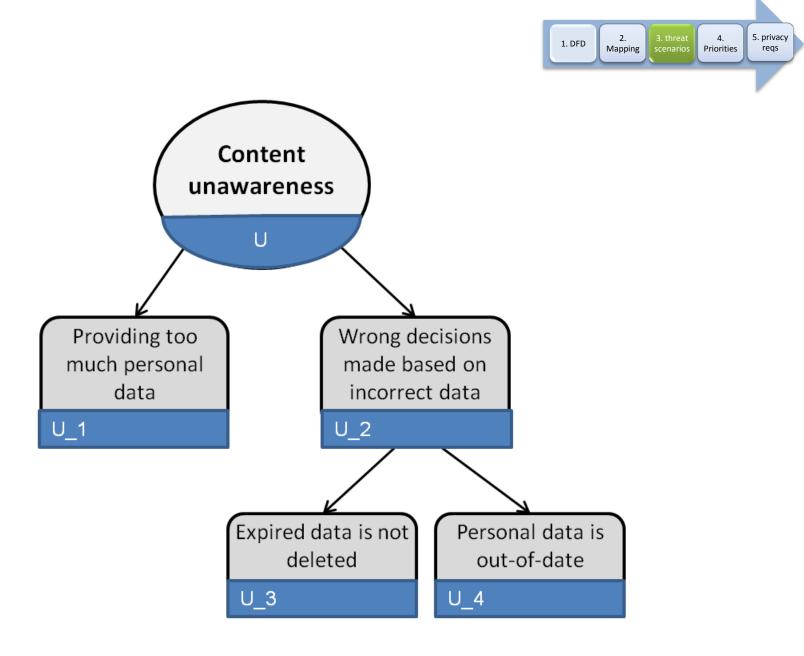
DFD element(s): data flow from browse service to external disease service (5.6 -> 4)

- r1. I_df1 requires an unprotected data flow, which is currently present (assumption 3) and misactor is receiver, thus assumption always applies
- r2. The different requests are traced back based on the transmitted (temporary/internal) user ID (I_df8)
- r3. The right branch of the tree (insecure anonymity system (I_df4)) and the other leaf nodes of the non-anonymous communication branch (I_df3) are not considered, as it is not the sender (browse service) whose identity should be protected, but the patient, who is not directly part of the data flow



Soft privacy

- Non-compliance of employees
- Non-compliance of management
- Missing consent system
- Patient unawareness
- Content inaccuracy





T19 - User unwareness

Summary: The user is unaware of the consequences of sharing information (e.g. by sharing too much information even anonymized data can reveal the user's identity)

Primary mis-actor: Management

Basic path:

- bf1. The management fails to add as requirement the need of notifications and warnings when the patients intends to upload sensitive and/or identifiable content (e.g. picture of his broken arm which also shows his face)
- bf2. The user adds information to the system which can easily identify him (e.g. a picture of himself) as he is unaware of the consequences
- bf3. Group members retrieve information and can still identify the pseudonymized user **Consequence:** When group members retrieve information, the identifiable information. The user's privacy is violated as he assumes that his information stays confidential and his identity will not be revealed

Reference to threat tree node(s): \cup 1

Parent threat tree(s): U

DFD element(s): 1 patient

- r1. This threat only applies to the patient (assumption 19)
- r2. The threat concerning inaccurate user information is described in T20 content inaccuracy

		L	I	N	D	D	U	N
Data store	PHR data (5.1)	T01	T03			T04, T05, T06, T07		
	User data (5.2)	T02						
flow	Patient – portal flow (1 -5.3)			J		Т09		
	Portal – patient flow (5.3-1)					T10		
	Researcher – portal flow (2-5.4)							
	Portal – researcher flow (5.4-2)							
	Nurse – portal flow (3-5.5)							
	Portal – nurse flow (5.5-3)							
	Disease service – browse diseases flow (4-5.6)							T16, T17,
	Browse disease – disease service flow (5.6-4)	T11	T12					T18
	Patient portal - browse diseases flow (5.3-5.6)					T13		
	Patient – authN flow (1-5.9)					T08		
	authN - patient flow (5.9-1)							
	Research – authN flow (2-5.9)							
	authN flow – researcher (5.9-2)							
	Nurse – authN flow (3-5.9)							
	authN – nurse (3-5.9)							
process	Patient portal (5.3)					T14, T15		
Entity	Patient (1)						T19, T20	
	Researcher (2)							
	Nurse (3)						1	28
	External disease service (4)							



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Information disclosure and identifiability of stored data violate privacy more than disclosure of "partial" transmitted data

Priorities



High

- T04 Information disclosure of patient community data
- T03 Identifying a patient from his PHR data
- T08 Disclosure of the transmitted log-in credentials
- T09 Disclosure of the transmitted session token
- T10 Disclosure of transmitted medical/personal information
- T05 Spoofing a user of the social network system by falsifying credentials
- T07 Spoofing a user of the social network system because of weak credential storage
- T06 Spoofing a user of the social network system by eavesdropping communication

Spoofing leads to information disclosure which is a high risk threat

Data in system is purely informative, and not used for important decisions, thus impact of threat is low

Low

- T16 Non-compliance of employees
- T20 content inaccuracy
- T14 Information disclosure internal process
- T13 Disclosure of internal transmitted medical/personal information
- T15 Side channel information disclosure internal process

There is a trust relationship with the employees, thus likelihood of threats is low

Medium

- T12 Identifiability of data sent to external disease service
- T11 Linkability of symptoms sent to external disease services
- T01 Profiling PHR data (linking)
- T02 Linking PHR data to user data
- T18 Non-compliance management
- T17 Missing user consents
- T19 User unwareness

Only partial data and patient deniability

Linkability can lead to identifiability

Non-compliance can still have "part" in place + reputation 130



LINDDUN Methodology

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Threats to requirements T01 - Profiling PHR data

- A researcher or other insider with malicious intent links PHR data or user data
 - e.g. based on medication which is usually combined, medical conditions which occur together, or pseudo-identifiers like street and age

Threats (misuse cases)	Caused by (leaf nodes)	Mitigated by (requirements)
Profiling PHR	Weak data anonymity (in the data store)	Apply strong data anonymization techniques in the database (for storage)
data	PII linked (after retrieval)	Apply data anonymization techniques on query results (for information retrieval)

3. threat

scenarios

Priorities

Mapping