

Let's design a URL shortening service like TinyURL. This service will provide short aliases redirecting to long URLs. Similar services: bit.ly, goo.gl, 2020.fm etc. Difficulty Level: Easy


## 1. Why do we need URL shortening?

URL shortening is used to create shorter aliases for long URLs. Users are redirected to the original URL when they hit these aliases. A shorter version of any URL would save a lot of space whenever we use it e.g., when printing or tweeting as tweets have a character limit.

The shortened URL is nearly 1/3rd of the size of the actual URL.

URL shortening is used for optimizing links across devices, tracking individual links to analyze audience and campaign performance, and hiding affiliated original URLs, etc. If you haven't used [tinyurl.com](https://tinyurl.com) before, please try creating a new shortened URL and spend some time going through different options their service offers. This will help you a lot in understanding this chapter better.

## 2. Requirements and Goals of the System

 *You should always clarify requirements at the beginning of the interview and should ask questions to find the exact scope of the system that the interviewer has in mind.*

Our URL shortener system should meet the following requirements:

### Functional Requirements:

1. Given a URL, our service should generate a shorter and unique alias of it.
2. When users access a shorter URL, our service should redirect them to the original link.
3. Users should optionally be able to pick a custom alias for their URL.
4. Links will expire after a specific timespan automatically; users should also be able to specify expiration time.

### Non-Functional Requirements:

1. The system should be highly available. This is required because if our service is down, all the URL redirections will start failing.
2. URL redirection should happen in real-time with minimum latency.
3. Shortened links should not be guessable (not predictable).

## Extended Requirements:

1. Analytics, e.g., how many times a redirection happened?
2. Our service should also be accessible through REST APIs by other services.

### 3. Capacity Estimation and Constraints

Our system would be read-heavy; there would be lots of redirection requests compared to new URL shortenings. Let's assume 100:1 ratio between read and write.

**Traffic estimates:** If we assume that we would have 500M new URLs shortenings per month, we can expect ( $100 * 500M \Rightarrow 50B$ ) redirections during the same time. What would be Queries Per Second (QPS) for our system?

New URLs shortenings per second:

$500 \text{ million} / (30 \text{ days} * 24 \text{ hours} * 3600 \text{ seconds}) \sim 200 \text{ URLs/s}$

URLs redirections per second:

$50 \text{ billion} / (30 \text{ days} * 24 \text{ hours} * 3600 \text{ sec}) \sim 19K/s$

**Storage estimates:** Since we expect to have 500M new URLs every month and if we would be keeping these objects for five years; total number of objects we will be storing would be 30 billion.

$500 \text{ million} * 5 \text{ years} * 12 \text{ months} = 30 \text{ billion}$

Let's assume that each object we are storing can be of 500 bytes (just a ballpark, we will dig into it later); we would need 15TB of total storage:

$30 \text{ billion} * 500 \text{ bytes} = 15 \text{ TB}$

URL Shortenings per month	500	million
Total years	5	
URL object size	500	Bytes
<hr/>		
Total Files	30	billion
Total Storage	15	TB

**Bandwidth estimates:** For write requests, since every second we expect 200 new URLs, total incoming data for our service would be 100KB per second.

$$200 * 500 \text{ bytes} = 100 \text{ KB/s}$$

For read requests, since every second we expect ~19K URLs redirections, total outgoing data for our service would be 9MB per second.

$$19K * 500 \text{ bytes} \approx 9 \text{ MB/s}$$

**Memory estimates:** If we want to cache some of the hot URLs that are frequently accessed, how much memory would we need to store them? If we follow the 80-20 rule, meaning 20% of URLs generating 80% of traffic, we would like to cache these 20% hot URLs.

Since we have 19K requests per second, we would be getting 1.7billion requests per day.

$$19K * 3600 \text{ seconds} * 24 \text{ hours} \approx 1.7 \text{ billion}$$

To cache 20% of these requests, we would need 170GB of memory.

$$0.2 * 1.7 \text{ billion} * 500 \text{ bytes} \approx 170GB$$

**High level estimates:** Assuming 500 million new URLs per month and 100:1 [read:write](#) ratio, following is the summary of the high level estimates for our service:

New URLs	200/s
URL redirections	19K/s
Incoming data	100KB/s
Outgoing data	9MB/s
Storage for 5 years	15TB
Memory for cache	170GB

#### 4. System APIs

💡 *Once we've finalized the requirements, it's always a good idea to define the system APIs. This would explicitly state what is expected from the system.*

We can have SOAP or REST APIs to expose the functionality of our service. Following could be the definitions of the APIs for creating and deleting URLs:

```
creatURL(api_dev_key, original_url, custom_alias=None user_name=None, expire_date=None)
```

**Parameters:**

api\_dev\_key (string): The API developer key of a registered account. This will be used to, among other things, throttle users based on their allocated quota.

original\_url (string): Original URL to be shortened.

custom\_alias (string): Optional custom key for the URL.

user\_name (string): Optional user name to be used in encoding.

expire\_date (string): Optional expiration date for the shortened URL.

**Returns:** (string)

A successful insertion returns the shortened URL, otherwise, returns an error code.

```
deleteURL(api_dev_key, url_key)
```

Where “url\_key” is a string representing the shortened URL to be retrieved. A successful deletion returns ‘URL Removed’.

**How do we detect and prevent abuse?** For instance, any service can put us out of business by consuming all our keys in the current design. To prevent abuse, we can limit users through their api\_dev\_key, how many URL they can create or access in a certain time.

## 5. Database Design

💡 *Defining the DB schema in the early stages of the interview would help to understand the data flow among various components and later would guide towards the data partitioning.*

A few observations about nature of the data we are going to store:

1. We need to store billions of records.
2. Each object we are going to store is small (less than 1K).
3. There are no relationships between records, except if we want to store which user created what URL.
4. Our service is read-heavy.

**Database Schema:**

We would need two tables, one for storing information about the URL mappings and the other for users' data.

URL	
PK	<b><u>Hash: varchar(16)</u></b>
	OriginalURL: varchar(512) CreationDate: datetime ExpirationDate: datetime UserID: int

User	
PK	<b><u>UserID: int</u></b>
	Name: varchar(20) Email: varchar(32) CreationDate: datetime LastLogin: datetime

**What kind of database should we use?** Since we are likely going to store billions of rows and we don't need to use relationships between objects – a NoSQL key-value store like Dynamo or Cassandra is a better choice, which would also be easier to scale. If we choose NoSQL, we cannot store UserID in the URL table (as there are no foreign keys in NoSQL), for that we would need a third table which will store the mapping between URL and the user.

## 6. Basic System Design and Algorithm

The problem we are solving here is to generate a short and unique key for the given URL. For example, the shortened URL can be: “<http://tinyurl.com/abcdef>”, the last six characters of this URL is the short key we want to generate. We'll explore two solutions here:

### a. Encoding actual URL

We can compute a unique hash (e.g., [MD5](#) or [SHA256](#), etc.) of the given URL. The hash can then be encoded for displaying. This encoding could be base36 ([a-z, 0-9]) or base62 ([A-Z, a-z, 0-9]) and if we add '-' and '.', we can use base64 encoding. A reasonable question would be; what should be the length of the short key? 6, 8 or 10 characters?

Using base64 encoding, a 6 letter long key would result in  $64^6 \approx 68.7$  billion possible strings

Using base64 encoding, an 8 letter long key would result in  $64^8 \approx 281$  trillion possible strings

With 68.7B unique strings, let's assume for our system six letters keys would suffice.

If we use the MD5 algorithm as our hash function, it'll produce a 128-bit hash value. After base64 encoding we'll get a string having more than 20 characters, how will we choose our key then? We can take the first 6 (or 8) letters for the key. This could result in key duplication though, upon which we can choose some other characters out of the encoding string or swap some characters.

**What are different issues with our solution?** We have the following couple of problems with our encoding scheme:

1. If multiple users enter the same URL, they can get the same shortened URL, which is not acceptable.
2. What if parts of the URL are URL-encoded?

**Workaround for the issues:** We can append an increasing sequence number to each input URL to make it unique and then generate a hash of it. We don't need to store this sequence number in the databases, though. Possible problems with this approach could be how big this sequence number would be, can it overflow? Appending an increasing sequence number will impact the performance of the service too.

Another solution could be, to append user id (which should be unique) to the input URL. However, if the user has not signed in, we can ask the user to choose a uniqueness key. Even after this if we have a conflict, we have to keep generating a key until we get a unique one.

## **b. Generating keys offline**

We can have a standalone Key Generation Service (KGS) that generates random six letter strings beforehand and stores them in a database (let's call it key-DB). Whenever we want to shorten a URL, we will just take one of the already generated keys and use it. This approach will make things quite simple and fast since we will not be encoding the URL or worrying about duplications or collisions. KGS will make sure all the keys inserted in key-DB are unique.

**Can concurrency cause problems?** As soon as a key is used, it should be marked in the database so that it doesn't get used again. If there are multiple servers reading keys concurrently, we might get a scenario where two or more servers try to read the same key from the database. How can we solve this concurrency problem?

Servers can use KGS to read/mark keys in the database. KGS can use two tables to store keys, one for keys that are not used yet and one for all the used keys. As soon as KGS

gives keys to one of the servers, it can move them to the used keys table. KGS can always keep some keys in memory so that whenever a server needs them, it can quickly provide them. For simplicity, as soon as KGS loads some keys in memory, it can move them to used keys table. This way we can make sure each server gets unique keys.

If KGS dies before assigning all the loaded keys to some server, we will be wasting those keys, which we can ignore given a huge number of keys we have. KGS also has to make sure not to give the same key to multiple servers. For that, it must synchronize (or get a lock to) the data structure holding the keys before removing keys from it and giving them to a server.

**What would be the key-DB size?** With base64 encoding, we can generate 68.7B unique six letters keys. If we need one byte to store one alpha-numeric character, we can store all these keys in:

$6 \text{ (characters per key)} * 68.7\text{B (unique keys)} \Rightarrow 412 \text{ GB.}$

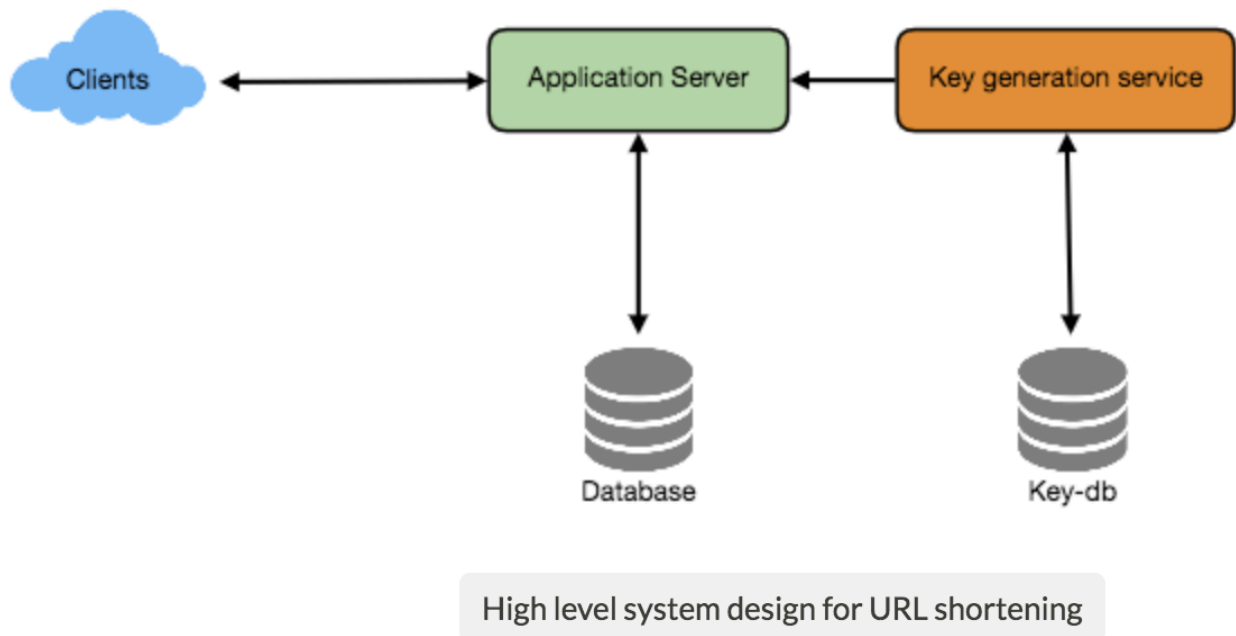
**Isn't KGS the single point of failure?** Yes, it is. To solve this, we can have a standby replica of KGS, and whenever the primary server dies, it can take over to generate and provide keys.

**Can each app server cache some keys from key-DB?** Yes, this can surely speed things up. Although in this case, if the application server dies before consuming all the keys, we will end up losing those keys. This could be acceptable since we have 68B unique six letters keys.

**How would we perform a key lookup?** We can look up the key in our database or key-value store to get the full URL. If it's present, issue a "HTTP 302 Redirect" status back to the browser, passing the stored URL in the "Location" field of the request. If that key is not present in our system, issue a "HTTP 404 Not Found" status, or redirect the user back to the homepage.

**Should we impose size limits on custom aliases?** Since our service supports custom aliases, users can pick any 'key' they like, but providing a custom alias is not mandatory. However, it is reasonable (and often desirable) to impose a size limit on a custom alias, so that we have a consistent URL database. Let's assume users can specify maximum 16 characters long customer key (as reflected in the above database schema).





## 7. Data Partitioning and Replication

To scale out our DB, we need to partition it so that it can store information about billions of URL. We need to come up with a partitioning scheme that would divide and store our data to different DB servers.

**a. Range Based Partitioning:** We can store URLs in separate partitions based on the first letter of the URL or the hash key. Hence we save all the URLs starting with letter 'A' in one partition and those that start with letter 'B' into another partition and so on. This approach is called range based partitioning. We can even combine certain less frequently occurring letters into one database partition. We should come up with this partitioning scheme statically so that we can always store/find a file in a predictable manner.

The main problem with this approach is that it can lead to unbalanced servers, for instance; if we decide to put all URLs starting with letter 'E' into a DB partition, but later we realize that we have too many URLs that start with letter 'E', which we can't fit into one DB partition.

**b. Hash-Based Partitioning:** In this scheme, we take a hash of the object we are storing, and based on this hash we figure out the DB partition to which this object should go. In our case, we can take the hash of the 'key' or the actual URL to determine the partition to store the file. Our hashing function will randomly distribute URLs into different partitions, e.g., our hashing function can always map any key to a number between  $[1 \dots 256]$ , and this number would represent the partition to store our object.

This approach can still lead to overloaded partitions, which can be solved by using Consistent Hashing.

## 8. Cache

We can cache URLs that are frequently accessed. We can use some off-the-shelf solution like Memcache, that can store full URLs with their respective hashes. The application servers, before hitting backend storage, can quickly check if the cache has desired URL.

**How much cache should we have?** We can start with 20% of daily traffic and based on clients' usage pattern we can adjust how many cache servers we need. As estimated above we need 170GB memory to cache 20% of daily traffic since a modern day server can have 256GB memory, we can easily fit all the cache into one machine, or we can choose to use a couple of smaller servers to store all these hot URLs.

**Which cache eviction policy would best fit our needs?** When the cache is full, and we want to replace a link with a newer/hotter URL, how would we choose? Least Recently Used (LRU) can be a reasonable policy for our system. Under this policy, we discard the least recently used URL first. We can use a [Linked Hash Map](#) or a similar data structure to store our URLs and Hashes, which will also keep track of which URLs are accessed recently.

To further increase the efficiency, we can replicate our caching servers to distribute load between them.

**How can each cache replica be updated?** Whenever there is a cache miss, our servers would be hitting backend database. Whenever this happens, we can update the cache and pass the new entry to all the cache replicas. Each replica can update their cache by adding the new entry. If a replica already has that entry, it can simply ignore it.