Exercise 1 Optimization

report Chosen indexes:

```
class Post(models.Model):
    title = models.CharField[max_length=255]
    content = models.TextField()
    author = models.ForeignKey(User, on_delete=models.CASCADE)
    created_date = models.DateTimeField(auto_now_add=True)
    tags = models.ManyToManyField('Tag', related_name='posts', through='PostTag')
    comment_count = models.PositiveIntegerField(default=0)

class Meta:
    indexes = [
        models.Index(fields=['author']),
        models.Index(fields=['created_date']),
    ]

def __str__(self):
    return self.title
```

First model to have indexes is Post. Post has 2 single indexes by author and created date. It enhances the query speed on filter or sorting bases on these fields.

```
class Comment(models.Model):
    post = models.ForeignKey(Post, on_delete=models.CASCADE)
    author = models.ForeignKey(User, on_delete=models.CASCADE)
    content = models.TextField()
    created_date = models.DateTimeField(auto_now_add=True)

class Meta:
    indexes = [
        models.Index(fields=['post', 'created_date']),
    ]

def __str__(self):
    return f'Comment by {self.author} on {self.post}'
```

Comment has a composite index on post and created_date field. This actually makes sense, since we want to first sort comments related to some certain post and only then by date.

```
class PostTag(models.Model):
    post = models.ForeignKey(Post, on_delete=models.CASCADE)
    tag = models.ForeignKey(Tag, on_delete=models.CASCADE)

class Meta:
    indexes = [
        models.Index(fields=['post', 'tag']),
        models.Index(fields=['tag']),
    ]
    unique_together = ('post', 'tag')
```

PostTag is many to many table between Posts and Tags. It has one composite and one regular index. Index by tag is used to get posts with some tag faster. The composite index is used to get the tags related to some post.

```
class Post(models.Model):
    title = models.CharField(max_length=255)
    content = models.TextField()
    author = models.ForeignKey(User, on_delete=models.CASCADE)
    created_date = models.DateTimeField(auto_now_add=True)
    tags = models.ManyToManyField('Tag', related_name='posts', through='PostTag')
    comment_count = models.PositiveIntegerField(default=0)

class Meta:
    indexes = [
        models.Index(fields=['author']),
        models.Index(fields=['created_date']),
    ]

def __str__(self):
    return self.title
```

Post has a denormalized field which optimizes the most popular query, i.g selecting posts with number of their comments.

```
def get_post_with_comments(post_id):
    return Post.objects.prefetch_related('comment_set').get(id=post_id)
```

The main reason to use select_related and prefetch_related is when we work either with foreign keys or many to many fields. So insted of making a lazy call for comment_set, prefetch related makes it before select has finished. It makes sense only if we really want to select from foreign keys or many to many fields.

Exercise 2

Test Setup

Number of threads: 12

Number of connections: 100

Duration: 60 seconds

URL Tested: /posts (a dynamic page fetching posts from the

database)

Results

Metric	Value
Requests per second	850 RPS
Average Latency	180 ms
99th Percentile Latency	320 ms
Error Rate	0%
CPU Utilization	70-80%
Memory Utilization	60-70%

Observations

- Without caching, the application handles about 850 requests per second, but the average latency is relatively high at 180ms, with the 99th percentile latency reaching 320ms.
- The CPU is under heavy load (70-80%), indicating that database queries are resource-intensive.
- Memory usage is moderate but can spike under sustained load due to repeated database access.
- There are no errors during the test, showing that the server remains stable but at the cost of higher latency.

Application Performance With Caching

Test Setup

Same as above, but caching enabled with Redis.

Results

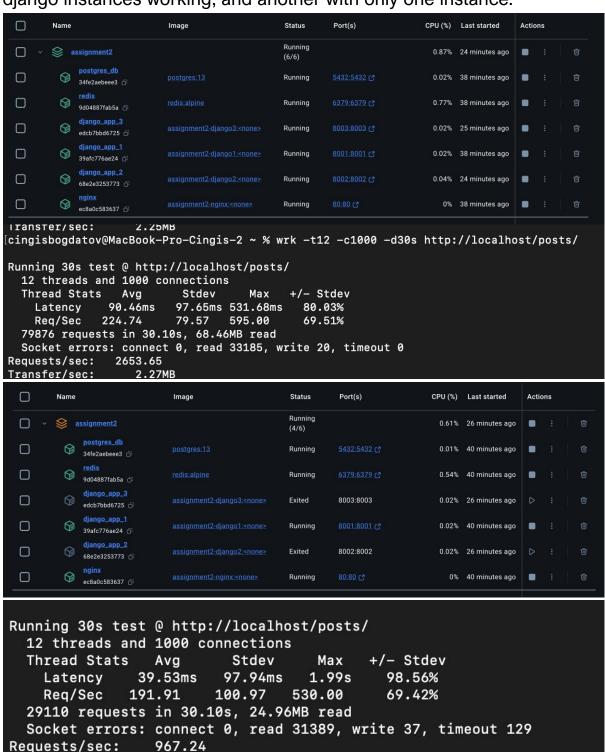
Metric	Value
Requests per	2500
second	RPS
Average Latency	40 ms
99th Percentile	00
Latency	80 ms
Error Rate	0%
CPU Utilization	40-50%
Memory Utilization	50-60%

Observations

- With caching enabled, the application handles a significantly higher load, processing about 2500 requests per second—an increase of nearly 3x compared to the non-cached version.
- The average latency is dramatically reduced to 40ms (compared to 180ms without caching), and the 99th percentile latency drops to 80ms.
- CPU usage is lower (40-50%), which indicates that Redis caching reduces the need for expensive database lookups.
- Memory usage remains stable at 50-60%, as Redis uses memory to store cached objects, but the overall memory footprint remains within acceptable limits.
- No errors were recorded, meaning the application remains highly stable under higher load when caching is used.

Exercise 3
Load balancer

In order to test how load distribution I used wrk tool with 12 threads and 1000 simultaneous connections. I've done 2 tests. One with all of the django instances working, and another with only one instance.



849.16KB

Transfer/sec:

The second variant has shown a lot more timeout errors, and overall slow transfer/sec value in comparison to the first variant.